

```

name: <unnamed>
log: C:\Users\ljzig\Desktop\buyaker et al 2020\LJZ buyaker et al 2020.smcl
log type: smcl
opened on: 31 Dec 2020, 12:18:21

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```
1 . do "C:\Users\ljzig\AppData\Local\Temp\STD22e0_000000.tmp"
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2 . *** Outcome
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3 .
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4 . tab V162062x votechoice, mi
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2016 PRE-POST VOTE SUMMARY: 2016 Presidential vote	votechoice			Total
	0	1	.	
-9. Refused	0	0	31	31
-8. Don't know (FTF o	0	0	2	2
-2. Missing, no vote	0	0	1,426	1,426
1. Hillary Clinton	1,364	0	0	1,364
2. Donald Trump	0	1,245	0	1,245
3. Gary Johnson	0	0	118	118
4. Jill Stein	0	0	32	32
5. Other candidate SP	0	0	52	52
Total	1,364	1,245	1,661	4,270

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5 .
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6 . *** Immigration items
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7 .
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8 . tab1 V161192 V161194x V161195x V161196x V162157 V162158 V162268 V162269 V162270 V162313
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-> tabulation of V161192

PRE: U.S. government policy toward unauthorized immigrants	Freq.	Percent	Cum.
1. Make all unauthorized immigrants fel	703	16.69	16.69
2. Have a guest worker program in order	644	15.29	31.99
3. Allow to remain and eventually quali	2,447	58.11	90.10
4. Allow to remain and eventually quali	417	9.90	100.00
Total	4,211	100.00	

-> tabulation of V161194x

PRE: SUMMARY - birthright citizenship	Freq.	Percent	Cum.
1. Favor a great deal	772	18.18	18.18
2. Favor a moderate amount	475	11.19	29.37
3. Favor a little	103	2.43	31.79
4. Neither favor nor oppose	1,210	28.50	60.29
5. Oppose a little	135	3.18	63.47
6. Oppose a moderate amount	557	13.12	76.59
7. Oppose a great deal	994	23.41	100.00
Total	4,246	100.00	

-> tabulation of V161195x

PRE: SUMMARY - Children brought illegally	Freq.	Percent	Cum.
1. Should send back - favor a great dea	329	7.84	7.84
2. Should send back - favor a moderate	332	7.92	15.76
3. Should send back - favor a little	112	2.67	18.43
4. Should allow to stay - favor a littl	435	10.37	28.80
5. Should allow to stay - favor a moder	1,437	34.26	63.07
6. Should allow to stay - favor a great	1,549	36.93	100.00
Total	4,194	100.00	

-> tabulation of V161196x

PRE: SUMMARY - Build wall with Mexico	Freq.	Percent	Cum.
1. Favor a great deal	880	20.72	20.72
2. Favor a moderate amount	407	9.58	30.30
3. Favor a little	82	1.93	32.23
4. Neither favor nor oppose	934	21.99	54.21
5. Oppose a little	87	2.05	56.26
6. Oppose a moderate amount	347	8.17	64.43
7. Oppose a great deal	1,511	35.57	100.00
Total	4,248	100.00	

-> tabulation of V162157

POST: What should immigration levels be	Freq.	Percent	Cum.
1. Increased a lot	203	5.61	5.61
2. Increased a little	382	10.55	16.16
3. Left the same as it is now	1,442	39.82	55.98
4. Decreased a little	688	19.00	74.98
5. Decreased a lot	906	25.02	100.00
Total	3,621	100.00	

-> tabulation of V162158

POST: How likely immigration will take away jobs	Freq.	Percent	Cum.
1. Extremely likely	553	15.24	15.24
2. Very likely	737	20.31	35.55
3. Somewhat likely	1,474	40.62	76.16
4. Not at all likely	865	23.84	100.00
Total	3,629	100.00	

-> tabulation of V162268

POST: Immigrants are generally good for America's economy	Freq.	Percent	Cum.
1. Agree strongly	603	16.69	16.69
2. Agree somewhat	1,337	37.00	53.68
3. Neither agree nor disagree	933	25.82	79.50
4. Disagree somewhat	518	14.33	93.83
5. Disagree strongly	223	6.17	100.00
Total	3,614	100.00	

-> tabulation of V162269

POST: America's culture is generally harmed by immigrants	Freq.	Percent	Cum.
1. Agree strongly	170	4.70	4.70
2. Agree somewhat	507	14.01	18.70
3. Neither agree nor disagree	861	23.78	42.49
4. Disagree somewhat	1,048	28.95	71.44
5. Disagree strongly	1,034	28.56	100.00
Total	3,620	100.00	

-> tabulation of V162270

POST: Immigrants increase crime rates in the U.S.	Freq.	Percent	Cum.
1. Agree strongly	214	5.92	5.92
2. Agree somewhat	779	21.56	27.48
3. Neither agree nor disagree	1,040	28.78	56.25
4. Disagree somewhat	776	21.47	77.73
5. Disagree strongly	805	22.27	100.00
Total	3,614	100.00	

-> tabulation of V162313

POST: FTF CASI/WEB: Feeling thermometer: ILLEGAL IMMIGRANTS	Freq.	Percent	Cum.
0	440	12.29	12.29
1	44	1.23	13.52
2	25	0.70	14.22
3	18	0.50	14.72
4	8	0.22	14.94
5	10	0.28	15.22
6	4	0.11	15.34
7	5	0.14	15.47
8	2	0.06	15.53
9	3	0.08	15.61
10	21	0.59	16.20
11	5	0.14	16.34
12	2	0.06	16.40
13	3	0.08	16.48
14	5	0.14	16.62
15	292	8.16	24.78
16	42	1.17	25.95
17	20	0.56	26.51
18	16	0.45	26.96
19	4	0.11	27.07
20	24	0.67	27.74
21	4	0.11	27.85
22	5	0.14	27.99
23	10	0.28	28.27
24	9	0.25	28.52
25	10	0.28	28.80
26	1	0.03	28.83
27	4	0.11	28.94
28	3	0.08	29.02
29	8	0.22	29.25
30	300	8.38	37.63
31	27	0.75	38.38
32	24	0.67	39.05
33	6	0.17	39.22
34	1	0.03	39.25
35	16	0.45	39.69
36	3	0.08	39.78
37	7	0.20	39.97

38	8	0.22	40.20
39	12	0.34	40.53
40	379	10.59	51.12
41	42	1.17	52.29
42	17	0.47	52.77
43	6	0.17	52.93
44	5	0.14	53.07
45	17	0.47	53.55
46	7	0.20	53.74
47	5	0.14	53.88
48	3	0.08	53.97
49	12	0.34	54.30
50	492	13.74	68.04
51	65	1.82	69.86
52	23	0.64	70.50
53	6	0.17	70.67
54	5	0.14	70.81
55	7	0.20	71.01
56	4	0.11	71.12
57	2	0.06	71.17
58	5	0.14	71.31
59	10	0.28	71.59
60	241	6.73	78.32
61	26	0.73	79.05
62	14	0.39	79.44
63	4	0.11	79.55
64	5	0.14	79.69
65	15	0.42	80.11
66	5	0.14	80.25
67	3	0.08	80.34
68	5	0.14	80.47
69	7	0.20	80.67
70	254	7.09	87.77
71	13	0.36	88.13
72	10	0.28	88.41
73	8	0.22	88.63
74	2	0.06	88.69
75	27	0.75	89.44
76	2	0.06	89.50
77	1	0.03	89.53
78	3	0.08	89.61
79	3	0.08	89.69
80	12	0.34	90.03
81	5	0.14	90.17
82	5	0.14	90.31
83	2	0.06	90.36
84	6	0.17	90.53
85	165	4.61	95.14
86	6	0.17	95.31
87	3	0.08	95.39
88	3	0.08	95.47
89	3	0.08	95.56
90	15	0.42	95.98
92	5	0.14	96.12
93	1	0.03	96.15
95	7	0.20	96.34
96	2	0.06	96.40
97	2	0.06	96.45
98	4	0.11	96.56
99	4	0.11	96.68
100	119	3.32	100.00
Total	3,580	100.00	

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9 .
10 . *** Comparison of the xenophobia predictor and the racial resentment predictor
11 .
12 . codebook rrf immf if white==1

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rrf

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      type:  numeric (float)

      range:  [0,1]                units:  1.000e-08
unique values: 22                missing .:  410/3,038

      mean:   .579829
      std. dev: .282653

percentiles:      10%      25%      50%      75%      90%
                  .1875    .375    .625    .8125    .9375

```

immf

```

      type:  numeric (float)

      range:  [0,1]                units:  1.000e-09
unique values: 1,096            missing .:  1/3,038

      mean:   .465359
      std. dev: .224861

percentiles:      10%      25%      50%      75%      90%
                  .158333  .291667  .471667  .633333  .756333

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13 . tab1 rrf immf if white==1

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-> tabulation of rrf if white==1

rrf	Freq.	Percent	Cum.
0	116	4.41	4.41
.0625	74	2.82	7.23
.125	63	2.40	9.63
.1666667	1	0.04	9.67
.1875	77	2.93	12.60
.25	125	4.76	17.35
.3125	111	4.22	21.58
.3333333	1	0.04	21.61
.375	135	5.14	26.75
.4375	149	5.67	32.42
.5	259	9.86	42.28
.5625	178	6.77	49.05
.5833333	2	0.08	49.12
.625	211	8.03	57.15
.6875	199	7.57	64.73
.75	209	7.95	72.68
.8125	145	5.52	78.20
.8333333	1	0.04	78.23
.875	186	7.08	85.31
.9166667	1	0.04	85.35
.9375	143	5.44	90.79
1	242	9.21	100.00
Total	2,628	100.00	

-> tabulation of immf if white==1

immf	Freq.	Percent	Cum.
0	23	0.76	0.76
.005	1	0.03	0.79
.01	1	0.03	0.82
.015	5	0.16	0.99
.025	1	0.03	1.02
.03	3	0.10	1.12
.0316667	1	0.03	1.15
.0333333	1	0.03	1.19
.038	1	0.03	1.22
.0383333	1	0.03	1.25
.04	5	0.16	1.42
.041	1	0.03	1.45
.0416667	6	0.20	1.65
.0433333	1	0.03	1.68
.045	1	0.03	1.71
.0473333	1	0.03	1.75
.0483333	5	0.16	1.91
.049	1	0.03	1.94
.05	10	0.33	2.27
.0543333	1	0.03	2.30
.055	3	0.10	2.40
.056	1	0.03	2.44
.0566667	3	0.10	2.54
.0583333	1	0.03	2.57
.06	1	0.03	2.60
.0633333	1	0.03	2.63
.065	2	0.07	2.70
.065	1	0.03	2.73
.0663333	1	0.03	2.77
.0713333	1	0.03	2.80
.0716667	3	0.10	2.90
.0733333	4	0.13	3.03
.0740741	1	0.03	3.06
.0743333	1	0.03	3.10
.075	3	0.10	3.19
.0783333	1	0.03	3.23
.08	6	0.20	3.42
.0813333	1	0.03	3.46
.0816667	3	0.10	3.56
.083	1	0.03	3.59
.0833333	23	0.76	4.35
.0866667	1	0.03	4.38
.0883333	3	0.10	4.48
.09	3	0.10	4.58
.0916667	4	0.13	4.71
.0933333	1	0.03	4.74
.0963333	1	0.03	4.77
.0966667	2	0.07	4.84
.0983333	4	0.13	4.97
.1003333	1	0.03	5.00
.1013333	1	0.03	5.04
.1018519	1	0.03	5.07
.1033333	2	0.07	5.14
.1036667	1	0.03	5.17
.104	1	0.03	5.20
.105	3	0.10	5.30
.1053333	1	0.03	5.33
.1066667	2	0.07	5.40
.1073333	2	0.07	5.47
.1083333	4	0.13	5.60
.1093333	1	0.03	5.63
.11	4	0.13	5.76
.1111111	1	0.03	5.80

.1116667	1	0.03	5.83
.1133333	5	0.16	5.99
.114	1	0.03	6.03
.115	4	0.13	6.16
.1163333	1	0.03	6.19
.1166667	2	0.07	6.26
.118	1	0.03	6.29
.1183333	1	0.03	6.32
.1183333	1	0.03	6.35
.1203704	1	0.03	6.39
.1206667	1	0.03	6.42
.1216667	4	0.13	6.55
.1222222	2	0.07	6.62
.1233333	15	0.49	7.11
.125	13	0.43	7.54
.1266667	1	0.03	7.57
.1276667	1	0.03	7.61
.1283333	1	0.03	7.64
.1291667	1	0.03	7.67
.13	3	0.10	7.77
.1316667	3	0.10	7.87
.133	1	0.03	7.90
.1333333	10	0.33	8.23
.135	2	0.07	8.30
.1353333	1	0.03	8.33
.1373333	1	0.03	8.36
.1383333	5	0.16	8.53
.139	1	0.03	8.56
.14	3	0.10	8.66
.1416667	4	0.13	8.79
.1433333	2	0.07	8.86
.1443333	1	0.03	8.89
.145	2	0.07	8.96
.1466667	2	0.07	9.02
.1473333	3	0.10	9.12
.1476667	1	0.03	9.15
.1483333	4	0.13	9.29
.15	2	0.07	9.35
.1503333	1	0.03	9.38
.1516667	3	0.10	9.48
.1533333	4	0.13	9.61
.155	5	0.16	9.78
.1566667	3	0.10	9.88
.1583333	4	0.13	10.01
.159	1	0.03	10.04
.1593333	1	0.03	10.08
.16	5	0.16	10.24
.1603333	1	0.03	10.27
.1623333	1	0.03	10.31
.1633333	5	0.16	10.47
.164	2	0.07	10.54
.165	2	0.07	10.60
.1656667	1	0.03	10.64
.1663333	1	0.03	10.67
.1666667	13	0.43	11.10
.1683333	1	0.03	11.13
.1696667	2	0.07	11.20
.1703704	1	0.03	11.23
.1716667	3	0.10	11.33
.1733333	8	0.26	11.59
.175	9	0.30	11.89
.175	6	0.20	12.08
.176	2	0.07	12.15
.1766667	4	0.13	12.28
.177	1	0.03	12.31
.1776667	1	0.03	12.35
.1783333	1	0.03	12.38

.1783333	1	0.03	12.41
.18	4	0.13	12.55
.1813333	1	0.03	12.58
.1816667	4	0.13	12.71
.1823333	2	0.07	12.78
.1833333	8	0.26	13.04
.185	3	0.10	13.14
.1866667	2	0.07	13.20
.1873333	1	0.03	13.24
.1883333	3	0.10	13.34
.19	3	0.10	13.43
.1913333	1	0.03	13.47
.1916667	4	0.13	13.60
.1933333	3	0.10	13.70
.1933333	3	0.10	13.80
.195	2	0.07	13.86
.1966667	4	0.13	13.99
.1973333	1	0.03	14.03
.1983333	4	0.13	14.16
.2	2	0.07	14.22
.2006667	1	0.03	14.26
.2016667	1	0.03	14.29
.2023333	1	0.03	14.32
.2033333	3	0.10	14.42
.204	1	0.03	14.46
.205	2	0.07	14.52
.2056667	1	0.03	14.55
.2066667	1	0.03	14.59
.2073333	2	0.07	14.65
.2083333	3	0.10	14.75
.2083333	17	0.56	15.31
.21	1	0.03	15.34
.2106667	1	0.03	15.38
.2116667	2	0.07	15.44
.2133333	5	0.16	15.61
.215	2	0.07	15.67
.2166667	12	0.40	16.07
.2181482	1	0.03	16.10
.2183333	6	0.20	16.30
.219	1	0.03	16.33
.22	2	0.07	16.40
.2213333	1	0.03	16.43
.2216667	1	0.03	16.46
.223	1	0.03	16.50
.2233333	4	0.13	16.63
.2236667	1	0.03	16.66
.2243333	1	0.03	16.69
.2246667	2	0.07	16.76
.225	1	0.03	16.79
.225	3	0.10	16.89
.2263333	1	0.03	16.92
.2266667	4	0.13	17.06
.227	1	0.03	17.09
.2273333	3	0.10	17.19
.2283333	8	0.26	17.45
.2296667	1	0.03	17.48
.23	3	0.10	17.58
.2316667	4	0.13	17.71
.233	1	0.03	17.75
.2333333	8	0.26	18.01
.2343333	1	0.03	18.04
.235	6	0.20	18.24
.2366667	3	0.10	18.34
.237	1	0.03	18.37
.237037	1	0.03	18.41
.2383333	4	0.13	18.54
.24	2	0.07	18.60

.2406667	1	0.03	18.64
.2407407	1	0.03	18.67
.2413333	2	0.07	18.74
.2416667	7	0.23	18.97
.2433333	4	0.13	19.10
.245	4	0.13	19.23
.2453333	1	0.03	19.26
.2456667	1	0.03	19.30
.2462963	1	0.03	19.33
.2463333	1	0.03	19.36
.2466667	5	0.16	19.53
.2473333	1	0.03	19.56
.2481482	1	0.03	19.59
.2483333	6	0.20	19.79
.2496667	1	0.03	19.82
.25	6	0.20	20.02
.2513333	3	0.10	20.12
.252	1	0.03	20.15
.2523333	2	0.07	20.22
.2526667	1	0.03	20.25
.2533333	4	0.13	20.38
.2537037	1	0.03	20.41
.254	1	0.03	20.45
.255	4	0.13	20.58
.2553333	1	0.03	20.61
.2563334	1	0.03	20.65
.2566667	6	0.20	20.84
.2583333	11	0.36	21.21
.26	2	0.07	21.27
.2603333	1	0.03	21.30
.2606667	2	0.07	21.37
.2616667	5	0.16	21.53
.262963	1	0.03	21.57
.2633333	1	0.03	21.60
.2633334	5	0.16	21.76
.265	1	0.03	21.80
.2656667	1	0.03	21.83
.2666667	1	0.03	21.86
.2666667	6	0.20	22.06
.2683333	5	0.16	22.23
.2696667	1	0.03	22.26
.27	2	0.07	22.32
.2706667	1	0.03	22.36
.2716667	3	0.10	22.46
.2722222	1	0.03	22.49
.2723333	1	0.03	22.52
.2733333	3	0.10	22.62
.2738095	1	0.03	22.65
.275	5	0.16	22.82
.2756667	1	0.03	22.85
.2763333	1	0.03	22.88
.2766667	8	0.26	23.15
.2773333	1	0.03	23.18
.278	1	0.03	23.21
.2783333	4	0.13	23.35
.2796667	1	0.03	23.38
.28	3	0.10	23.48
.2806667	1	0.03	23.51
.2816667	5	0.16	23.67
.2826667	1	0.03	23.71
.2833333	9	0.30	24.00
.2843333	1	0.03	24.04
.285	7	0.23	24.27
.2853333	1	0.03	24.30
.2856667	1	0.03	24.33
.286	2	0.07	24.40
.2866667	4	0.13	24.53

.2873333	1	0.03	24.56
.2883333	4	0.13	24.70
.29	2	0.07	24.76
.2906667	1	0.03	24.79
.2907407	1	0.03	24.83
.2916667	9	0.30	25.12
.2916667	2	0.07	25.19
.2933333	1	0.03	25.22
.2933334	2	0.07	25.29
.2946667	1	0.03	25.32
.295	1	0.03	25.35
.296	2	0.07	25.42
.2966667	11	0.36	25.78
.2976667	1	0.03	25.81
.2981482	1	0.03	25.85
.2983333	1	0.03	25.88
.3	17	0.56	26.44
.3006667	1	0.03	26.47
.3016667	9	0.30	26.77
.3023333	1	0.03	26.80
.3033333	3	0.10	26.90
.305	2	0.07	26.97
.3055556	1	0.03	27.00
.3056667	1	0.03	27.03
.3066667	3	0.10	27.13
.308	2	0.07	27.20
.3083333	8	0.26	27.46
.3092593	1	0.03	27.49
.31	5	0.16	27.66
.3106667	3	0.10	27.76
.3116667	2	0.07	27.82
.313	1	0.03	27.86
.3133333	3	0.10	27.96
.3153333	1	0.03	27.99
.3156667	2	0.07	28.05
.3166667	6	0.20	28.25
.3173333	1	0.03	28.28
.3183333	3	0.10	28.38
.3186667	1	0.03	28.42
.319	1	0.03	28.45
.32	3	0.10	28.55
.3206667	1	0.03	28.58
.3213333	1	0.03	28.61
.3216667	10	0.33	28.94
.3233333	5	0.16	29.11
.3243333	1	0.03	29.14
.3246667	1	0.03	29.17
.325	3	0.10	29.27
.3259259	1	0.03	29.31
.3266667	1	0.03	29.34
.3266667	2	0.07	29.40
.3283333	3	0.10	29.50
.3293333	1	0.03	29.54
.33	1	0.03	29.57
.3314815	1	0.03	29.60
.3316667	7	0.23	29.83
.3333333	1	0.03	29.86
.3333333	17	0.56	30.42
.3346667	1	0.03	30.46
.335	4	0.13	30.59
.3356667	1	0.03	30.62
.3366667	5	0.16	30.79
.3373333	2	0.07	30.85
.3376667	2	0.07	30.92
.338	1	0.03	30.95
.3383333	3	0.10	31.05
.339	1	0.03	31.08

.3393333	1	0.03	31.12
.34	3	0.10	31.22
.3403333	1	0.03	31.25
.3407407	1	0.03	31.28
.3416667	13	0.43	31.71
.342	1	0.03	31.74
.3423333	1	0.03	31.77
.3433333	2	0.07	31.84
.345	2	0.07	31.91
.3456667	2	0.07	31.97
.3466667	11	0.36	32.33
.3483333	3	0.10	32.43
.35	5	0.16	32.60
.35	5	0.16	32.76
.3516667	4	0.13	32.89
.352	1	0.03	32.93
.3533333	3	0.10	33.03
.3543333	1	0.03	33.06
.355	1	0.03	33.09
.355	3	0.10	33.19
.3553333	1	0.03	33.22
.3555556	1	0.03	33.26
.3566667	2	0.07	33.32
.3566667	2	0.07	33.39
.357	1	0.03	33.42
.3573333	1	0.03	33.45
.3583333	1	0.03	33.49
.3583333	6	0.20	33.68
.359	2	0.07	33.75
.36	4	0.13	33.88
.3606667	1	0.03	33.92
.3616667	6	0.20	34.11
.3633333	5	0.16	34.28
.365	4	0.13	34.41
.3666667	6	0.20	34.61
.367	1	0.03	34.64
.3676667	1	0.03	34.67
.368	1	0.03	34.71
.3683333	3	0.10	34.80
.3685185	1	0.03	34.84
.3686667	1	0.03	34.87
.37	2	0.07	34.94
.3706667	1	0.03	34.97
.3716667	11	0.36	35.33
.372	1	0.03	35.36
.3733333	7	0.23	35.59
.3736667	1	0.03	35.63
.375	2	0.07	35.69
.3756667	1	0.03	35.73
.3759259	1	0.03	35.76
.376	1	0.03	35.79
.3766667	5	0.16	35.96
.3773333	1	0.03	35.99
.3783333	4	0.13	36.12
.379	1	0.03	36.15
.3793333	1	0.03	36.19
.38	6	0.20	36.38
.3816667	1	0.03	36.42
.383	1	0.03	36.45
.3833333	26	0.86	37.31
.384	2	0.07	37.37
.385	4	0.13	37.50
.3853333	1	0.03	37.54
.3863333	1	0.03	37.57
.3866667	7	0.23	37.80
.3866667	2	0.07	37.87
.3883334	6	0.20	38.06

.3886667	1	0.03	38.10
.389	2	0.07	38.16
.3893333	1	0.03	38.20
.39	2	0.07	38.26
.39	4	0.13	38.39
.3916667	2	0.07	38.46
.3916667	9	0.30	38.76
.3923333	1	0.03	38.79
.3926667	1	0.03	38.82
.3933333	5	0.16	38.99
.394	2	0.07	39.05
.395	5	0.16	39.22
.3956667	1	0.03	39.25
.3962963	2	0.07	39.32
.3966667	6	0.20	39.51
.3983333	5	0.16	39.68
.3986667	1	0.03	39.71
.399	1	0.03	39.74
.4	6	0.20	39.94
.401	1	0.03	39.97
.4013333	2	0.07	40.04
.4016667	8	0.26	40.30
.402	1	0.03	40.34
.4023333	1	0.03	40.37
.4023333	2	0.07	40.43
.403	1	0.03	40.47
.4033333	8	0.26	40.73
.404	2	0.07	40.80
.4046667	1	0.03	40.83
.405	7	0.23	41.06
.4066667	7	0.23	41.29
.4074074	1	0.03	41.32
.4083333	4	0.13	41.46
.4093333	1	0.03	41.49
.4096667	3	0.10	41.59
.41	7	0.23	41.82
.411	1	0.03	41.85
.4116667	6	0.20	42.05
.4123333	1	0.03	42.08
.4126667	1	0.03	42.11
.4133333	3	0.10	42.21
.4136667	1	0.03	42.25
.4143333	1	0.03	42.28
.415	4	0.13	42.41
.4156667	3	0.10	42.51
.4166667	5	0.16	42.67
.4166667	4	0.13	42.81
.4173333	1	0.03	42.84
.4183333	1	0.03	42.87
.4183334	1	0.03	42.90
.4196667	2	0.07	42.97
.42	3	0.10	43.07
.42	5	0.16	43.23
.4206667	1	0.03	43.27
.421	1	0.03	43.30
.4216667	4	0.13	43.43
.4233333	5	0.16	43.60
.4240741	1	0.03	43.63
.4246667	1	0.03	43.66
.425	8	0.26	43.92
.4256667	1	0.03	43.96
.4259259	1	0.03	43.99
.426	1	0.03	44.02
.4263333	1	0.03	44.06
.4266667	3	0.10	44.16
.4273333	1	0.03	44.19
.4283333	3	0.10	44.29

.4296667	1	0.03	44.32
.43	8	0.26	44.58
.4316667	6	0.20	44.78
.4323333	1	0.03	44.81
.4333333	4	0.13	44.95
.4346667	1	0.03	44.98
.435	3	0.10	45.08
.4356667	2	0.07	45.14
.4366667	6	0.20	45.34
.437	1	0.03	45.37
.4373333	1	0.03	45.41
.4380952	1	0.03	45.44
.4383333	4	0.13	45.57
.4388889	1	0.03	45.60
.439	1	0.03	45.64
.44	2	0.07	45.70
.4403333	2	0.07	45.77
.441	1	0.03	45.80
.4416667	7	0.23	46.03
.4423333	1	0.03	46.07
.4426667	1	0.03	46.10
.443	1	0.03	46.13
.4433333	2	0.07	46.20
.4444444	2	0.07	46.26
.445	1	0.03	46.30
.4466667	6	0.20	46.49
.4483333	7	0.23	46.72
.449	2	0.07	46.79
.45	4	0.13	46.92
.45	1	0.03	46.95
.4516667	2	0.07	47.02
.4516667	3	0.10	47.12
.4533333	3	0.10	47.22
.4543333	1	0.03	47.25
.455	4	0.13	47.38
.4556667	1	0.03	47.42
.4563333	1	0.03	47.45
.4566667	11	0.36	47.81
.4574074	1	0.03	47.84
.4583333	11	0.36	48.21
.4586667	1	0.03	48.24
.46	2	0.07	48.30
.4616667	4	0.13	48.44
.4623333	1	0.03	48.47
.463	1	0.03	48.50
.4633333	1	0.03	48.53
.4643333	1	0.03	48.57
.465	5	0.16	48.73
.466	1	0.03	48.77
.4666667	21	0.69	49.46
.467	1	0.03	49.49
.4683333	5	0.16	49.65
.47	7	0.23	49.88
.4703704	1	0.03	49.92
.471	1	0.03	49.95
.4713333	1	0.03	49.98
.4716667	7	0.23	50.21
.472	1	0.03	50.25
.473	1	0.03	50.28
.4733333	7	0.23	50.51
.4743333	1	0.03	50.54
.4746667	1	0.03	50.58
.475	7	0.23	50.81
.475	2	0.07	50.87
.4753333	1	0.03	50.91
.4763333	1	0.03	50.94
.4766667	8	0.26	51.20

.4773333	1	0.03	51.23
.4783333	5	0.16	51.40
.4786667	1	0.03	51.43
.4791667	1	0.03	51.47
.4796667	1	0.03	51.50
.48	2	0.07	51.56
.48	4	0.13	51.70
.4806667	2	0.07	51.76
.4813333	1	0.03	51.79
.4816667	3	0.10	51.89
.4816667	5	0.16	52.06
.482	1	0.03	52.09
.4823333	1	0.03	52.12
.4826667	1	0.03	52.16
.4833333	4	0.13	52.29
.484	2	0.07	52.35
.485	3	0.10	52.45
.4853333	1	0.03	52.49
.4866667	9	0.30	52.78
.4873333	1	0.03	52.82
.488	1	0.03	52.85
.4883333	1	0.03	52.88
.4886667	1	0.03	52.91
.4888889	1	0.03	52.95
.4895833	1	0.03	52.98
.4896667	1	0.03	53.01
.49	7	0.23	53.24
.4906667	1	0.03	53.28
.4907407	1	0.03	53.31
.4916667	7	0.23	53.54
.4933333	3	0.10	53.64
.4936667	1	0.03	53.67
.4946667	2	0.07	53.74
.495	9	0.30	54.03
.4956667	1	0.03	54.07
.4958333	1	0.03	54.10
.496	1	0.03	54.13
.4966667	8	0.26	54.40
.498	1	0.03	54.43
.4983333	5	0.16	54.59
.499	1	0.03	54.63
.5	9	0.30	54.92
.5007408	1	0.03	54.96
.501	1	0.03	54.99
.5013334	1	0.03	55.02
.5016667	5	0.16	55.19
.5018519	1	0.03	55.22
.5026667	1	0.03	55.25
.5033333	6	0.20	55.45
.5037037	1	0.03	55.48
.5040741	1	0.03	55.52
.505	5	0.16	55.68
.506	1	0.03	55.71
.5066667	6	0.20	55.91
.5083333	10	0.33	56.24
.5086667	1	0.03	56.27
.51	4	0.13	56.40
.5116667	4	0.13	56.54
.512	1	0.03	56.57
.512963	1	0.03	56.60
.5133333	4	0.13	56.73
.5142857	1	0.03	56.77
.515	6	0.20	56.96
.516	1	0.03	57.00
.5163333	2	0.07	57.06
.5166667	16	0.53	57.59
.5176667	1	0.03	57.62

.5183333	5	0.16	57.79
.5185185	1	0.03	57.82
.52	3	0.10	57.92
.5206667	1	0.03	57.95
.5216666	3	0.10	58.05
.5233333	5	0.16	58.22
.5244445	1	0.03	58.25
.525	2	0.07	58.31
.525	2	0.07	58.38
.5253333	1	0.03	58.41
.5263333	1	0.03	58.45
.5266666	6	0.20	58.64
.5266667	1	0.03	58.68
.5277778	1	0.03	58.71
.5281482	1	0.03	58.74
.5283333	6	0.20	58.94
.5283334	1	0.03	58.97
.529	2	0.07	59.04
.5296296	1	0.03	59.07
.53	1	0.03	59.10
.53	8	0.26	59.37
.5303333	1	0.03	59.40
.5306666	3	0.10	59.50
.531	1	0.03	59.53
.5313333	1	0.03	59.57
.5316666	1	0.03	59.60
.5316667	4	0.13	59.73
.5323333	1	0.03	59.76
.5326667	2	0.07	59.83
.5333333	1	0.03	59.86
.5333334	9	0.30	60.16
.5343333	1	0.03	60.19
.535	5	0.16	60.36
.536	1	0.03	60.39
.5366667	4	0.13	60.52
.537	1	0.03	60.55
.5373333	2	0.07	60.62
.5383334	3	0.10	60.72
.5388889	1	0.03	60.75
.5393333	1	0.03	60.78
.5396667	1	0.03	60.82
.54	3	0.10	60.92
.5416667	8	0.26	61.18
.5433334	3	0.10	61.28
.5436667	1	0.03	61.31
.544	3	0.10	61.41
.5446666	1	0.03	61.44
.545	7	0.23	61.67
.5453333	1	0.03	61.71
.5456667	1	0.03	61.74
.5466667	5	0.16	61.90
.547	2	0.07	61.97
.5473334	1	0.03	62.00
.548	1	0.03	62.03
.5483333	4	0.13	62.17
.5496666	2	0.07	62.23
.55	24	0.79	63.02
.5533333	2	0.07	63.09
.5543333	1	0.03	63.12
.555	5	0.16	63.29
.5551852	1	0.03	63.32
.5553333	1	0.03	63.35
.5555556	3	0.10	63.45
.5566667	3	0.10	63.55
.5573334	1	0.03	63.58
.5574074	1	0.03	63.62
.5576667	1	0.03	63.65

.558	1	0.03	63.68
.5583333	8	0.26	63.94
.559	1	0.03	63.98
.5592592	1	0.03	64.01
.5596667	1	0.03	64.04
.56	4	0.13	64.18
.561	1	0.03	64.21
.5611111	1	0.03	64.24
.5616667	7	0.23	64.47
.563	1	0.03	64.50
.5633333	8	0.26	64.77
.564	3	0.10	64.87
.5648148	1	0.03	64.90
.565	4	0.13	65.03
.5666667	6	0.20	65.23
.568	1	0.03	65.26
.5683333	4	0.13	65.39
.5693333	1	0.03	65.43
.5696667	2	0.07	65.49
.57	3	0.10	65.59
.5706667	1	0.03	65.62
.5716667	3	0.10	65.72
.572	2	0.07	65.79
.5723333	1	0.03	65.82
.5729167	1	0.03	65.85
.5733333	3	0.10	65.95
.5743333	1	0.03	65.99
.575	5	0.16	66.15
.5756667	1	0.03	66.18
.5766667	1	0.03	66.22
.5770833	1	0.03	66.25
.5773333	1	0.03	66.28
.578	1	0.03	66.32
.5783333	6	0.20	66.51
.5786667	1	0.03	66.55
.579	1	0.03	66.58
.58	11	0.36	66.94
.5806667	1	0.03	66.97
.5816666	7	0.23	67.20
.5833333	15	0.49	67.70
.584	1	0.03	67.73
.585	2	0.07	67.80
.5851852	1	0.03	67.83
.5866666	5	0.16	67.99
.587	1	0.03	68.03
.587037	1	0.03	68.06
.588	1	0.03	68.09
.5883333	7	0.23	68.32
.5883334	4	0.13	68.46
.5888889	1	0.03	68.49
.5893334	1	0.03	68.52
.59	3	0.10	68.62
.59	1	0.03	68.65
.5903333	1	0.03	68.69
.5906667	1	0.03	68.72
.5907407	1	0.03	68.75
.5916666	6	0.20	68.95
.5916667	10	0.33	69.28
.5923333	1	0.03	69.31
.5925926	1	0.03	69.34
.593	1	0.03	69.38
.5933334	3	0.10	69.48
.594	1	0.03	69.51
.5946667	1	0.03	69.54
.595	5	0.16	69.71
.5956666	1	0.03	69.74
.5966667	8	0.26	70.00

.5973333	1	0.03	70.04
.5974074	1	0.03	70.07
.5977778	1	0.03	70.10
.5983334	4	0.13	70.23
.5996667	1	0.03	70.27
.6	4	0.13	70.40
.6016667	7	0.23	70.63
.6018519	1	0.03	70.66
.6023333	1	0.03	70.69
.6033334	9	0.30	70.99
.604	1	0.03	71.02
.605	3	0.10	71.12
.6053333	1	0.03	71.16
.6055555	1	0.03	71.19
.6056666	1	0.03	71.22
.6066667	4	0.13	71.35
.607	1	0.03	71.39
.6073333	1	0.03	71.42
.6074074	1	0.03	71.45
.6083333	7	0.23	71.68
.6086667	1	0.03	71.72
.61	2	0.07	71.78
.6103333	1	0.03	71.81
.6111111	1	0.03	71.85
.6116667	4	0.13	71.98
.6133333	4	0.13	72.11
.614	4	0.13	72.24
.6143333	1	0.03	72.28
.615	4	0.13	72.41
.6166667	8	0.26	72.67
.6176667	1	0.03	72.70
.6183333	6	0.20	72.90
.619	2	0.07	72.97
.62	5	0.16	73.13
.6206667	1	0.03	73.16
.6216667	6	0.20	73.36
.622	1	0.03	73.39
.6226667	1	0.03	73.43
.6233333	6	0.20	73.63
.624	1	0.03	73.66
.6246667	1	0.03	73.69
.625	5	0.16	73.86
.6266667	5	0.16	74.02
.6273333	1	0.03	74.05
.628	1	0.03	74.09
.6283333	6	0.20	74.28
.6296296	1	0.03	74.32
.63	7	0.23	74.55
.6306667	1	0.03	74.58
.6314815	1	0.03	74.61
.6316667	5	0.16	74.78
.6323333	1	0.03	74.81
.6333333	25	0.82	75.63
.6343333	1	0.03	75.67
.635	2	0.07	75.73
.6356667	1	0.03	75.77
.6366667	9	0.30	76.06
.6383333	8	0.26	76.33
.6386667	1	0.03	76.36
.639	3	0.10	76.46
.6396667	1	0.03	76.49
.64	3	0.10	76.59
.641	1	0.03	76.62
.6416667	9	0.30	76.92
.6433333	7	0.23	77.15
.645	1	0.03	77.18
.6456667	1	0.03	77.21

.646	1	0.03	77.25
.6463333	1	0.03	77.28
.6466666	6	0.20	77.48
.647	1	0.03	77.51
.6481482	1	0.03	77.54
.6483333	5	0.16	77.71
.6486667	2	0.07	77.77
.649	1	0.03	77.81
.65	3	0.10	77.91
.6506667	1	0.03	77.94
.651	1	0.03	77.97
.6516666	3	0.10	78.07
.6518518	1	0.03	78.10
.6523333	2	0.07	78.17
.6533333	6	0.20	78.37
.6533334	2	0.07	78.43
.655	5	0.16	78.60
.656	1	0.03	78.63
.6563333	1	0.03	78.66
.6566666	1	0.03	78.70
.6566667	2	0.07	78.76
.6574074	1	0.03	78.79
.6583334	4	0.13	78.93
.6592593	1	0.03	78.96
.6596296	1	0.03	78.99
.6596667	1	0.03	79.03
.66	3	0.10	79.12
.6603333	2	0.07	79.19
.6616667	7	0.23	79.42
.6633334	3	0.10	79.52
.665	5	0.16	79.68
.6656667	1	0.03	79.72
.6666667	9	0.30	80.01
.667	1	0.03	80.05
.668	1	0.03	80.08
.6683334	1	0.03	80.11
.6696666	2	0.07	80.18
.67	4	0.13	80.31
.6716667	2	0.07	80.38
.672	1	0.03	80.41
.6733333	7	0.23	80.64
.6736667	1	0.03	80.67
.6746666	2	0.07	80.74
.675	14	0.46	81.20
.6766667	3	0.10	81.30
.677	1	0.03	81.33
.6773334	1	0.03	81.36
.678	1	0.03	81.40
.6783333	5	0.16	81.56
.679	1	0.03	81.59
.6796667	1	0.03	81.63
.68	6	0.20	81.82
.6813334	1	0.03	81.86
.6816667	3	0.10	81.96
.6818519	1	0.03	81.99
.682	1	0.03	82.02
.6833333	10	0.33	82.35
.684	1	0.03	82.38
.685	2	0.07	82.45
.6856667	1	0.03	82.48
.6866667	7	0.23	82.71
.6876667	2	0.07	82.78
.6883333	4	0.13	82.91
.69	5	0.16	83.08
.6916667	3	0.10	83.17
.694	2	0.07	83.24
.6951852	1	0.03	83.27

.6966667	6	0.20	83.47
.697	1	0.03	83.50
.6973333	2	0.07	83.57
.6983333	8	0.26	83.83
.6986667	1	0.03	83.87
.6988889	1	0.03	83.90
.7	15	0.49	84.39
.7006667	1	0.03	84.43
.701	1	0.03	84.46
.7016667	1	0.03	84.49
.7023333	2	0.07	84.56
.7033333	5	0.16	84.72
.7037037	1	0.03	84.75
.704	2	0.07	84.82
.705	6	0.20	85.02
.7063333	1	0.03	85.05
.7066666	3	0.10	85.15
.7074074	1	0.03	85.18
.7083333	3	0.10	85.28
.71	4	0.13	85.41
.7113333	3	0.10	85.51
.713	1	0.03	85.54
.7133333	3	0.10	85.64
.715	3	0.10	85.74
.715	2	0.07	85.81
.7166667	4	0.13	85.94
.7183333	1	0.03	85.97
.7196667	1	0.03	86.01
.72	7	0.23	86.24
.7206666	1	0.03	86.27
.7213333	1	0.03	86.30
.7216667	2	0.07	86.37
.7223333	1	0.03	86.40
.7233334	2	0.07	86.47
.7246667	1	0.03	86.50
.725	5	0.16	86.66
.7266667	2	0.07	86.73
.7276667	1	0.03	86.76
.7283334	1	0.03	86.80
.729	1	0.03	86.83
.73	6	0.20	87.03
.731	1	0.03	87.06
.7313333	1	0.03	87.09
.7314815	1	0.03	87.13
.7316667	5	0.16	87.29
.732	2	0.07	87.36
.7323333	1	0.03	87.39
.7323334	2	0.07	87.45
.7326667	1	0.03	87.49
.7333333	6	0.20	87.69
.734	1	0.03	87.72
.735	1	0.03	87.75
.7363334	1	0.03	87.78
.7366667	2	0.07	87.85
.7373334	1	0.03	87.88
.7383333	5	0.16	88.05
.739	3	0.10	88.15
.7396667	1	0.03	88.18
.74	4	0.13	88.31
.7416667	3	0.10	88.41
.7433333	4	0.13	88.54
.745	2	0.07	88.61
.7462963	1	0.03	88.64
.7466667	8	0.26	88.90
.7473333	1	0.03	88.94
.7483333	3	0.10	89.04
.7486666	1	0.03	89.07

.749	2	0.07	89.13
.7493333	1	0.03	89.17
.75	15	0.49	89.66
.7516667	3	0.10	89.76
.7518519	1	0.03	89.79
.752	1	0.03	89.83
.7523333	1	0.03	89.86
.7526667	1	0.03	89.89
.7533333	1	0.03	89.92
.755	1	0.03	89.96
.7556667	1	0.03	89.99
.7563334	1	0.03	90.02
.7566667	3	0.10	90.12
.7573333	1	0.03	90.15
.7583333	4	0.13	90.29
.759	1	0.03	90.32
.76	1	0.03	90.35
.7603334	1	0.03	90.39
.7606667	1	0.03	90.42
.7616667	5	0.16	90.58
.7623333	1	0.03	90.62
.7633333	2	0.07	90.68
.765	4	0.13	90.81
.7666667	7	0.23	91.04
.7673333	1	0.03	91.08
.7683333	1	0.03	91.11
.7685185	1	0.03	91.14
.77	5	0.16	91.31
.7706667	1	0.03	91.34
.7716666	4	0.13	91.47
.7722222	1	0.03	91.50
.7733333	1	0.03	91.54
.775	1	0.03	91.57
.7756667	1	0.03	91.60
.7766666	1	0.03	91.64
.7777778	1	0.03	91.67
.778	1	0.03	91.70
.779	1	0.03	91.74
.78	1	0.03	91.77
.7806666	1	0.03	91.80
.7816666	1	0.03	91.83
.7816667	2	0.07	91.90
.7823333	1	0.03	91.93
.783	1	0.03	91.97
.7833333	1	0.03	92.00
.7833334	2	0.07	92.06
.784	1	0.03	92.10
.785	3	0.10	92.20
.7866667	5	0.16	92.36
.7876667	1	0.03	92.39
.788	1	0.03	92.43
.79	2	0.07	92.49
.7907407	1	0.03	92.53
.7913333	1	0.03	92.56
.7916667	4	0.13	92.69
.7933334	1	0.03	92.72
.795	3	0.10	92.82
.7966667	6	0.20	93.02
.797	1	0.03	93.05
.7983333	4	0.13	93.18
.7996666	1	0.03	93.22
.8	7	0.23	93.45
.8016667	3	0.10	93.55
.804	1	0.03	93.58
.8055556	1	0.03	93.61
.8056667	1	0.03	93.65
.8066667	2	0.07	93.71

.8073334	1	0.03	93.74
.8081481	1	0.03	93.78
.8083333	4	0.13	93.91
.81	1	0.03	93.94
.8106667	1	0.03	93.97
.8116667	1	0.03	94.01
.8123333	1	0.03	94.04
.8125	1	0.03	94.07
.8133333	2	0.07	94.14
.815	1	0.03	94.17
.8166667	2	0.07	94.24
.82	1	0.03	94.27
.8206667	1	0.03	94.30
.8233333	1	0.03	94.34
.8240741	1	0.03	94.37
.825	4	0.13	94.50
.8259259	1	0.03	94.53
.8283333	1	0.03	94.57
.829	1	0.03	94.60
.83	5	0.16	94.76
.8316666	2	0.07	94.83
.8333333	7	0.23	95.06
.8366666	1	0.03	95.09
.8383333	2	0.07	95.16
.8395833	1	0.03	95.19
.84	1	0.03	95.23
.8416667	2	0.07	95.29
.8456666	1	0.03	95.32
.8466667	1	0.03	95.36
.848	1	0.03	95.39
.8483334	1	0.03	95.42
.849	1	0.03	95.46
.85	6	0.20	95.65
.8507407	1	0.03	95.69
.8518519	1	0.03	95.72
.8533334	1	0.03	95.75
.855	1	0.03	95.79
.8566667	1	0.03	95.82
.8573334	1	0.03	95.85
.8583333	4	0.13	95.98
.8596666	1	0.03	96.02
.86	1	0.03	96.05
.8616667	1	0.03	96.08
.862963	1	0.03	96.11
.8633333	1	0.03	96.15
.8666667	7	0.23	96.38
.87	1	0.03	96.41
.8703704	1	0.03	96.44
.871	1	0.03	96.48
.8716667	3	0.10	96.58
.8726667	1	0.03	96.61
.8733333	1	0.03	96.64
.8746667	1	0.03	96.67
.875	9	0.30	96.97
.8756667	1	0.03	97.00
.8766667	1	0.03	97.04
.88	4	0.13	97.17
.8833333	2	0.07	97.23
.8846667	1	0.03	97.27
.8866667	1	0.03	97.30
.8883333	1	0.03	97.33
.8888889	2	0.07	97.40
.8896667	1	0.03	97.43
.8916667	2	0.07	97.50
.892	1	0.03	97.53
.8933333	1	0.03	97.56
.8946667	1	0.03	97.60

.895	1	0.03	97.63
.898	1	0.03	97.66
.9	6	0.20	97.86
.9033333	1	0.03	97.89
.9066667	1	0.03	97.93
.9133334	1	0.03	97.96
.9156667	1	0.03	97.99
.9166666	1	0.03	98.02
.9166667	6	0.20	98.22
.922	1	0.03	98.25
.925	3	0.10	98.35
.9323334	1	0.03	98.39
.9333333	3	0.10	98.49
.937	1	0.03	98.52
.945	1	0.03	98.55
.9466667	2	0.07	98.62
.95	3	0.10	98.72
.955	2	0.07	98.78
.9573333	1	0.03	98.81
.9583333	2	0.07	98.88
.96	2	0.07	98.95
.9636667	1	0.03	98.98
.972	1	0.03	99.01
.974	1	0.03	99.05
.975	3	0.10	99.14
.98	1	0.03	99.18
.998	1	0.03	99.21
1	24	0.79	100.00
Total	3,037	100.00	

```

14 .
15 . *** Buyaker et al 2020 model (Table 2 Model 3 I think)
16 .
17 . logit votechoice jardina3 rrf immf sexismf econ1 age3044 age4564 age65p female prot south ba incl
    > 1 ideoll1 if white==1 [pw=wtg]

```

```

Iteration 0:  log pseudolikelihood = -1196.262
Iteration 1:  log pseudolikelihood = -361.62569
Iteration 2:  log pseudolikelihood = -339.63013
Iteration 3:  log pseudolikelihood = -338.14372
Iteration 4:  log pseudolikelihood = -338.14164
Iteration 5:  log pseudolikelihood = -338.14164

```

```

Logistic regression                                Number of obs      =      1,879
                                                    Wald chi2(16)      =      308.75
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -338.14164                Pseudo R2         =      0.7173

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-1.304398	.5488355	-2.38	0.017	-2.380096	-.2287003
rrf	2.674991	.6802597	3.93	0.000	1.341707	4.008276
immf	6.099171	.9129822	6.68	0.000	4.309759	7.888583
sexismf	3.153369	.8475656	3.72	0.000	1.492171	4.814567
econ1	-1.351709	.6215366	-2.17	0.030	-2.569898	-.1335195
age3044	.2834575	.5601598	0.51	0.613	-.8144355	1.38135
age4564	.0733347	.5857693	0.13	0.900	-1.074752	1.221421
age65p	.3614768	.5463438	0.66	0.508	-.7093374	1.432291
female	.3349215	.2317952	1.44	0.148	-.1193887	.7892317
prot	-.1406833	.2575992	-0.55	0.585	-.6455686	.3642019
south	.7495102	.2724147	2.75	0.006	.2155871	1.283433
ba	-.0516494	.253691	-0.20	0.839	-.5488746	.4455759
incl	-.4771696	.5611904	-0.85	0.395	-1.577083	.6227435
incdk	-.3118771	.6404472	-0.49	0.626	-1.567131	.9433764

pid1	4.564902	.4311042	10.59	0.000	3.719953	5.409851
ideoll	2.936626	.6419583	4.57	0.000	1.678411	4.194841
_cons	-8.19265	.8649088	-9.47	0.000	-9.88784	-6.497459

```
18 . margins, atmeans at(rrf=(0 1))
```

Adjusted predictions	Number of obs	=	1,879
Model VCE : Robust			

Expression : `Pr(votechoice), predict()`

```
1._at      :  jardim3      =  .4270238 (mean)
            rrf           =  .0
            immf          =  .4689097 (mean)
            sexismf       =  .3719823 (mean)
            econ1         =  .4743608 (mean)
            age3044       =  .1959655 (mean)
            age4564       =  .4108636 (mean)
            age65p        =  .2584021 (mean)
            female        =  .5317754 (mean)
            prot          =  .3606696 (mean)
            south         =  .2612745 (mean)
            ba            =  .4006297 (mean)
            incl          =  .6104942 (mean)
            incdk         =  .0277694 (mean)
            pid1          =  .5391016 (mean)
            ideoll        =  .5575229 (mean)
```

```
2._at      :  jardim3      =  .4270238 (mean)
            rrf            =  1
            immf           =  .4689097 (mean)
            sexismf        =  .3719823 (mean)
            econ1          =  .4743608 (mean)
            age3044        =  .1959655 (mean)
            age4564        =  .4108636 (mean)
            age65p         =  .2584021 (mean)
            female         =  .5317754 (mean)
            prot           =  .3606696 (mean)
            south          =  .2612745 (mean)
            ba             =  .4006297 (mean)
            incl           =  .6104942 (mean)
            incdk          =  .0277694 (mean)
            pidl           =  .5391016 (mean)
            ideoll         =  .5575229 (mean)
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
at						
1	.2538653	.0821756	3.09	0.002	.0928041	.4149264
2	.8315831	.0406528	20.46	0.000	.751905	.9112611

```
19 . di 83.1-25.4 // 57.7
    57.7
```

```
20 . margins, atmeans at(immf=(0 1))
```

```
Adjusted predictions                                Number of obs    =      1,879
Model VCE      : Robust
```

```
Expression    : Pr(votechoice), predict()
```

```
1._at      : jardina3      =    .4270238 (mean)
              rrf           =    .5749568 (mean)
              immf          =           0
              sexismf       =    .3719823 (mean)
              econ1         =    .4743608 (mean)
              age3044       =    .1959655 (mean)
              age4564       =    .4108636 (mean)
              age65p        =    .2584021 (mean)
              female        =    .5317754 (mean)
              prot          =    .3606696 (mean)
              south         =    .2612745 (mean)
              ba            =    .4006297 (mean)
              incl          =    .6104942 (mean)
              incdk         =    .0277694 (mean)
              pid1          =    .5391016 (mean)
              ideoll        =    .5575229 (mean)
```

```
2._at      : jardina3      =    .4270238 (mean)
              rrf           =    .5749568 (mean)
              immf          =           1
              sexismf       =    .3719823 (mean)
              econ1         =    .4743608 (mean)
              age3044       =    .1959655 (mean)
              age4564       =    .4108636 (mean)
              age65p        =    .2584021 (mean)
              female        =    .5317754 (mean)
              prot          =    .3606696 (mean)
              south         =    .2612745 (mean)
              ba            =    .4006297 (mean)
              incl          =    .6104942 (mean)
              incdk         =    .0277694 (mean)
              pid1          =    .5391016 (mean)
              ideoll        =    .5575229 (mean)
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.0831682	.0321744	2.58	0.010	.0201075	.1462289
2	.9758521	.0123298	79.15	0.000	.951686	1.000018

```
21 . di 97.6-8.3 // 89.3
    89.3
```

```

22 .
23 . *** Reduced xenophobia predictors
24 .
25 . egen immf_5789=rowmean(antiimm5 antiimm7 antiimm8 antiimm9)
    (628 missing values generated)

26 . egen immf_5678=rowmean(antiimm5 antiimm6 antiimm7 antiimm8)
    (627 missing values generated)

27 . egen immf_6789=rowmean(antiimm6 antiimm7 antiimm8 antiimm9)
    (631 missing values generated)

28 .
29 . tab1      immf immf_5789 immf_5678 immf_6789 rrf if white==1

```

-> tabulation of immf if white==1

immf	Freq.	Percent	Cum.
0	23	0.76	0.76
.005	1	0.03	0.79
.01	1	0.03	0.82
.015	5	0.16	0.99
.025	1	0.03	1.02
.03	3	0.10	1.12
.0316667	1	0.03	1.15
.0333333	1	0.03	1.19
.038	1	0.03	1.22
.0383333	1	0.03	1.25
.04	5	0.16	1.42
.041	1	0.03	1.45
.0416667	6	0.20	1.65
.0433333	1	0.03	1.68
.045	1	0.03	1.71
.0473333	1	0.03	1.75
.0483333	5	0.16	1.91
.049	1	0.03	1.94
.05	10	0.33	2.27
.0543333	1	0.03	2.30
.055	3	0.10	2.40
.056	1	0.03	2.44
.0566667	3	0.10	2.54
.0583333	1	0.03	2.57
.06	1	0.03	2.60
.0633333	1	0.03	2.63
.065	2	0.07	2.70
.065	1	0.03	2.73
.0663333	1	0.03	2.77
.0713333	1	0.03	2.80
.0716667	3	0.10	2.90
.0733333	4	0.13	3.03
.0740741	1	0.03	3.06
.0743333	1	0.03	3.10
.075	3	0.10	3.19
.0783333	1	0.03	3.23
.08	6	0.20	3.42
.0813333	1	0.03	3.46
.0816667	3	0.10	3.56
.083	1	0.03	3.59
.0833333	23	0.76	4.35
.0866667	1	0.03	4.38
.0883333	3	0.10	4.48
.09	3	0.10	4.58
.0916667	4	0.13	4.71
.0933333	1	0.03	4.74
.0963333	1	0.03	4.77
.0966667	2	0.07	4.84

.0983333	4	0.13	4.97
.1003333	1	0.03	5.00
.1013333	1	0.03	5.04
.1018519	1	0.03	5.07
.1033333	2	0.07	5.14
.1036667	1	0.03	5.17
.104	1	0.03	5.20
.105	3	0.10	5.30
.1053333	1	0.03	5.33
.1066667	2	0.07	5.40
.1073333	2	0.07	5.47
.1083333	4	0.13	5.60
.1093333	1	0.03	5.63
.11	4	0.13	5.76
.1111111	1	0.03	5.80
.1116667	1	0.03	5.83
.1133333	5	0.16	5.99
.114	1	0.03	6.03
.115	4	0.13	6.16
.1163333	1	0.03	6.19
.1166667	2	0.07	6.26
.118	1	0.03	6.29
.1183333	1	0.03	6.32
.1183333	1	0.03	6.35
.1203704	1	0.03	6.39
.1206667	1	0.03	6.42
.1216667	4	0.13	6.55
.1222222	2	0.07	6.62
.1233333	15	0.49	7.11
.125	13	0.43	7.54
.1266667	1	0.03	7.57
.1276667	1	0.03	7.61
.1283333	1	0.03	7.64
.1291667	1	0.03	7.67
.13	3	0.10	7.77
.1316667	3	0.10	7.87
.133	1	0.03	7.90
.1333333	10	0.33	8.23
.135	2	0.07	8.30
.1353333	1	0.03	8.33
.1373333	1	0.03	8.36
.1383333	5	0.16	8.53
.139	1	0.03	8.56
.14	3	0.10	8.66
.1416667	4	0.13	8.79
.1433333	2	0.07	8.86
.1443333	1	0.03	8.89
.145	2	0.07	8.96
.1466667	2	0.07	9.02
.1473333	3	0.10	9.12
.1476667	1	0.03	9.15
.1483333	4	0.13	9.29
.15	2	0.07	9.35
.1503333	1	0.03	9.38
.1516667	3	0.10	9.48
.1533333	4	0.13	9.61
.155	5	0.16	9.78
.1566667	3	0.10	9.88
.1583333	4	0.13	10.01
.159	1	0.03	10.04
.1593333	1	0.03	10.08
.16	5	0.16	10.24
.1603333	1	0.03	10.27
.1623333	1	0.03	10.31
.1633333	5	0.16	10.47
.164	2	0.07	10.54
.165	2	0.07	10.60

.1656667	1	0.03	10.64
.1663333	1	0.03	10.67
.1666667	13	0.43	11.10
.1683333	1	0.03	11.13
.1696667	2	0.07	11.20
.1703704	1	0.03	11.23
.1716667	3	0.10	11.33
.1733333	8	0.26	11.59
.175	9	0.30	11.89
.175	6	0.20	12.08
.176	2	0.07	12.15
.1766667	4	0.13	12.28
.177	1	0.03	12.31
.1776667	1	0.03	12.35
.1783333	1	0.03	12.38
.1783333	1	0.03	12.41
.18	4	0.13	12.55
.1813333	1	0.03	12.58
.1816667	4	0.13	12.71
.1823333	2	0.07	12.78
.1833333	8	0.26	13.04
.185	3	0.10	13.14
.1866667	2	0.07	13.20
.1873333	1	0.03	13.24
.1883333	3	0.10	13.34
.19	3	0.10	13.43
.1913333	1	0.03	13.47
.1916667	4	0.13	13.60
.1933333	3	0.10	13.70
.1933333	3	0.10	13.80
.195	2	0.07	13.86
.1966667	4	0.13	13.99
.1973333	1	0.03	14.03
.1983333	4	0.13	14.16
.2	2	0.07	14.22
.2006667	1	0.03	14.26
.2016667	1	0.03	14.29
.2023333	1	0.03	14.32
.2033333	3	0.10	14.42
.204	1	0.03	14.46
.205	2	0.07	14.52
.2056667	1	0.03	14.55
.2066667	1	0.03	14.59
.2073333	2	0.07	14.65
.2083333	3	0.10	14.75
.2083333	17	0.56	15.31
.21	1	0.03	15.34
.2106667	1	0.03	15.38
.2116667	2	0.07	15.44
.2133333	5	0.16	15.61
.215	2	0.07	15.67
.2166667	12	0.40	16.07
.2181482	1	0.03	16.10
.2183333	6	0.20	16.30
.219	1	0.03	16.33
.22	2	0.07	16.40
.2213333	1	0.03	16.43
.2216667	1	0.03	16.46
.223	1	0.03	16.50
.2233333	4	0.13	16.63
.2236667	1	0.03	16.66
.2243333	1	0.03	16.69
.2246667	2	0.07	16.76
.225	1	0.03	16.79
.225	3	0.10	16.89
.2263333	1	0.03	16.92
.2266667	4	0.13	17.06

.227	1	0.03	17.09
.2273333	3	0.10	17.19
.2283333	8	0.26	17.45
.2296667	1	0.03	17.48
.23	3	0.10	17.58
.2316667	4	0.13	17.71
.233	1	0.03	17.75
.2333333	8	0.26	18.01
.2343333	1	0.03	18.04
.235	6	0.20	18.24
.2366667	3	0.10	18.34
.237	1	0.03	18.37
.237037	1	0.03	18.41
.2383333	4	0.13	18.54
.24	2	0.07	18.60
.2406667	1	0.03	18.64
.2407407	1	0.03	18.67
.2413333	2	0.07	18.74
.2416667	7	0.23	18.97
.2433333	4	0.13	19.10
.245	4	0.13	19.23
.2453333	1	0.03	19.26
.2456667	1	0.03	19.30
.2462963	1	0.03	19.33
.2463333	1	0.03	19.36
.2466667	5	0.16	19.53
.2473333	1	0.03	19.56
.2481482	1	0.03	19.59
.2483333	6	0.20	19.79
.2496667	1	0.03	19.82
.25	6	0.20	20.02
.2513333	3	0.10	20.12
.252	1	0.03	20.15
.2523333	2	0.07	20.22
.2526667	1	0.03	20.25
.2533333	4	0.13	20.38
.2537037	1	0.03	20.41
.254	1	0.03	20.45
.255	4	0.13	20.58
.2553333	1	0.03	20.61
.2563334	1	0.03	20.65
.2566667	6	0.20	20.84
.2583333	11	0.36	21.21
.26	2	0.07	21.27
.2603333	1	0.03	21.30
.2606667	2	0.07	21.37
.2616667	5	0.16	21.53
.262963	1	0.03	21.57
.2633333	1	0.03	21.60
.2633334	5	0.16	21.76
.265	1	0.03	21.80
.2656667	1	0.03	21.83
.2666667	1	0.03	21.86
.2666667	6	0.20	22.06
.2683333	5	0.16	22.23
.2696667	1	0.03	22.26
.27	2	0.07	22.32
.2706667	1	0.03	22.36
.2716667	3	0.10	22.46
.2722222	1	0.03	22.49
.2723333	1	0.03	22.52
.2733333	3	0.10	22.62
.2738095	1	0.03	22.65
.275	5	0.16	22.82
.2756667	1	0.03	22.85
.2763333	1	0.03	22.88
.2766667	8	0.26	23.15

.2773333	1	0.03	23.18
.278	1	0.03	23.21
.2783333	4	0.13	23.35
.2796667	1	0.03	23.38
.28	3	0.10	23.48
.2806667	1	0.03	23.51
.2816667	5	0.16	23.67
.2826667	1	0.03	23.71
.2833333	9	0.30	24.00
.2843333	1	0.03	24.04
.285	7	0.23	24.27
.2853333	1	0.03	24.30
.2856667	1	0.03	24.33
.286	2	0.07	24.40
.2866667	4	0.13	24.53
.2873333	1	0.03	24.56
.2883333	4	0.13	24.70
.29	2	0.07	24.76
.2906667	1	0.03	24.79
.2907407	1	0.03	24.83
.2916667	9	0.30	25.12
.2916667	2	0.07	25.19
.2933333	1	0.03	25.22
.2933334	2	0.07	25.29
.2946667	1	0.03	25.32
.295	1	0.03	25.35
.296	2	0.07	25.42
.2966667	11	0.36	25.78
.2976667	1	0.03	25.81
.2981482	1	0.03	25.85
.2983333	1	0.03	25.88
.3	17	0.56	26.44
.3006667	1	0.03	26.47
.3016667	9	0.30	26.77
.3023333	1	0.03	26.80
.3033333	3	0.10	26.90
.305	2	0.07	26.97
.3055556	1	0.03	27.00
.3056667	1	0.03	27.03
.3066667	3	0.10	27.13
.308	2	0.07	27.20
.3083333	8	0.26	27.46
.3092593	1	0.03	27.49
.31	5	0.16	27.66
.3106667	3	0.10	27.76
.3116667	2	0.07	27.82
.313	1	0.03	27.86
.3133333	3	0.10	27.96
.3153333	1	0.03	27.99
.3156667	2	0.07	28.05
.3166667	6	0.20	28.25
.3173333	1	0.03	28.28
.3183333	3	0.10	28.38
.3186667	1	0.03	28.42
.319	1	0.03	28.45
.32	3	0.10	28.55
.3206667	1	0.03	28.58
.3213333	1	0.03	28.61
.3216667	10	0.33	28.94
.3233333	5	0.16	29.11
.3243333	1	0.03	29.14
.3246667	1	0.03	29.17
.325	3	0.10	29.27
.3259259	1	0.03	29.31
.3266667	1	0.03	29.34
.3266667	2	0.07	29.40
.3283333	3	0.10	29.50

.3293333	1	0.03	29.54
.33	1	0.03	29.57
.3314815	1	0.03	29.60
.3316667	7	0.23	29.83
.3333333	1	0.03	29.86
.3333333	17	0.56	30.42
.3346667	1	0.03	30.46
.335	4	0.13	30.59
.3356667	1	0.03	30.62
.3366667	5	0.16	30.79
.3373333	2	0.07	30.85
.3376667	2	0.07	30.92
.338	1	0.03	30.95
.3383333	3	0.10	31.05
.339	1	0.03	31.08
.3393333	1	0.03	31.12
.34	3	0.10	31.22
.3403333	1	0.03	31.25
.3407407	1	0.03	31.28
.3416667	13	0.43	31.71
.342	1	0.03	31.74
.3423333	1	0.03	31.77
.3433333	2	0.07	31.84
.345	2	0.07	31.91
.3456667	2	0.07	31.97
.3466667	11	0.36	32.33
.3483333	3	0.10	32.43
.35	5	0.16	32.60
.35	5	0.16	32.76
.3516667	4	0.13	32.89
.352	1	0.03	32.93
.3533333	3	0.10	33.03
.3543333	1	0.03	33.06
.355	1	0.03	33.09
.355	3	0.10	33.19
.3553333	1	0.03	33.22
.3555556	1	0.03	33.26
.3566667	2	0.07	33.32
.3566667	2	0.07	33.39
.357	1	0.03	33.42
.3573333	1	0.03	33.45
.3583333	1	0.03	33.49
.3583333	6	0.20	33.68
.359	2	0.07	33.75
.36	4	0.13	33.88
.3606667	1	0.03	33.92
.3616667	6	0.20	34.11
.3633333	5	0.16	34.28
.365	4	0.13	34.41
.3666667	6	0.20	34.61
.367	1	0.03	34.64
.3676667	1	0.03	34.67
.368	1	0.03	34.71
.3683333	3	0.10	34.80
.3685185	1	0.03	34.84
.3686667	1	0.03	34.87
.37	2	0.07	34.94
.3706667	1	0.03	34.97
.3716667	11	0.36	35.33
.372	1	0.03	35.36
.3733333	7	0.23	35.59
.3736667	1	0.03	35.63
.375	2	0.07	35.69
.3756667	1	0.03	35.73
.3759259	1	0.03	35.76
.376	1	0.03	35.79
.3766667	5	0.16	35.96

.3773333	1	0.03	35.99
.3783333	4	0.13	36.12
.379	1	0.03	36.15
.3793333	1	0.03	36.19
.38	6	0.20	36.38
.3816667	1	0.03	36.42
.383	1	0.03	36.45
.3833333	26	0.86	37.31
.384	2	0.07	37.37
.385	4	0.13	37.50
.3853333	1	0.03	37.54
.3863333	1	0.03	37.57
.3866667	7	0.23	37.80
.3866667	2	0.07	37.87
.3883334	6	0.20	38.06
.3886667	1	0.03	38.10
.389	2	0.07	38.16
.3893333	1	0.03	38.20
.39	2	0.07	38.26
.39	4	0.13	38.39
.3916667	2	0.07	38.46
.3916667	9	0.30	38.76
.3923333	1	0.03	38.79
.3926667	1	0.03	38.82
.3933333	5	0.16	38.99
.394	2	0.07	39.05
.395	5	0.16	39.22
.3956667	1	0.03	39.25
.3962963	2	0.07	39.32
.3966667	6	0.20	39.51
.3983333	5	0.16	39.68
.3986667	1	0.03	39.71
.399	1	0.03	39.74
.4	6	0.20	39.94
.401	1	0.03	39.97
.4013333	2	0.07	40.04
.4016667	8	0.26	40.30
.402	1	0.03	40.34
.4023333	1	0.03	40.37
.4023333	2	0.07	40.43
.403	1	0.03	40.47
.4033333	8	0.26	40.73
.404	2	0.07	40.80
.4046667	1	0.03	40.83
.405	7	0.23	41.06
.4066667	7	0.23	41.29
.4074074	1	0.03	41.32
.4083333	4	0.13	41.46
.4093333	1	0.03	41.49
.4096667	3	0.10	41.59
.41	7	0.23	41.82
.411	1	0.03	41.85
.4116667	6	0.20	42.05
.4123333	1	0.03	42.08
.4126667	1	0.03	42.11
.4133333	3	0.10	42.21
.4136667	1	0.03	42.25
.4143333	1	0.03	42.28
.415	4	0.13	42.41
.4156667	3	0.10	42.51
.4166667	5	0.16	42.67
.4166667	4	0.13	42.81
.4173333	1	0.03	42.84
.4183333	1	0.03	42.87
.4183334	1	0.03	42.90
.4196667	2	0.07	42.97
.42	3	0.10	43.07

.42	5	0.16	43.23
.4206667	1	0.03	43.27
.421	1	0.03	43.30
.4216667	4	0.13	43.43
.4233333	5	0.16	43.60
.4240741	1	0.03	43.63
.4246667	1	0.03	43.66
.425	8	0.26	43.92
.4256667	1	0.03	43.96
.4259259	1	0.03	43.99
.426	1	0.03	44.02
.4263333	1	0.03	44.06
.4266667	3	0.10	44.16
.4273333	1	0.03	44.19
.4283333	3	0.10	44.29
.4296667	1	0.03	44.32
.43	8	0.26	44.58
.4316667	6	0.20	44.78
.4323333	1	0.03	44.81
.4333333	4	0.13	44.95
.4346667	1	0.03	44.98
.435	3	0.10	45.08
.4356667	2	0.07	45.14
.4366667	6	0.20	45.34
.437	1	0.03	45.37
.4373333	1	0.03	45.41
.4380952	1	0.03	45.44
.4383333	4	0.13	45.57
.4388889	1	0.03	45.60
.439	1	0.03	45.64
.44	2	0.07	45.70
.4403333	2	0.07	45.77
.441	1	0.03	45.80
.4416667	7	0.23	46.03
.4423333	1	0.03	46.07
.4426667	1	0.03	46.10
.443	1	0.03	46.13
.4433333	2	0.07	46.20
.4444444	2	0.07	46.26
.445	1	0.03	46.30
.4466667	6	0.20	46.49
.4483333	7	0.23	46.72
.449	2	0.07	46.79
.45	4	0.13	46.92
.45	1	0.03	46.95
.4516667	2	0.07	47.02
.4516667	3	0.10	47.12
.4533333	3	0.10	47.22
.4543333	1	0.03	47.25
.455	4	0.13	47.38
.4556667	1	0.03	47.42
.4563333	1	0.03	47.45
.4566667	11	0.36	47.81
.4574074	1	0.03	47.84
.4583333	11	0.36	48.21
.4586667	1	0.03	48.24
.46	2	0.07	48.30
.4616667	4	0.13	48.44
.4623333	1	0.03	48.47
.463	1	0.03	48.50
.4633333	1	0.03	48.53
.4643333	1	0.03	48.57
.465	5	0.16	48.73
.466	1	0.03	48.77
.4666667	21	0.69	49.46
.467	1	0.03	49.49
.4683333	5	0.16	49.65

.47	7	0.23	49.88
.4703704	1	0.03	49.92
.471	1	0.03	49.95
.4713333	1	0.03	49.98
.4716667	7	0.23	50.21
.472	1	0.03	50.25
.473	1	0.03	50.28
.4733333	7	0.23	50.51
.4743333	1	0.03	50.54
.4746667	1	0.03	50.58
.475	7	0.23	50.81
.475	2	0.07	50.87
.4753333	1	0.03	50.91
.4763333	1	0.03	50.94
.4766667	8	0.26	51.20
.4773333	1	0.03	51.23
.4783333	5	0.16	51.40
.4786667	1	0.03	51.43
.4791667	1	0.03	51.47
.4796667	1	0.03	51.50
.48	2	0.07	51.56
.48	4	0.13	51.70
.4806667	2	0.07	51.76
.4813333	1	0.03	51.79
.4816667	3	0.10	51.89
.4816667	5	0.16	52.06
.482	1	0.03	52.09
.4823333	1	0.03	52.12
.4826667	1	0.03	52.16
.4833333	4	0.13	52.29
.484	2	0.07	52.35
.485	3	0.10	52.45
.4853333	1	0.03	52.49
.4866667	9	0.30	52.78
.4873333	1	0.03	52.82
.488	1	0.03	52.85
.4883333	1	0.03	52.88
.4886667	1	0.03	52.91
.4888889	1	0.03	52.95
.4895833	1	0.03	52.98
.4896667	1	0.03	53.01
.49	7	0.23	53.24
.4906667	1	0.03	53.28
.4907407	1	0.03	53.31
.4916667	7	0.23	53.54
.4933333	3	0.10	53.64
.4936667	1	0.03	53.67
.4946667	2	0.07	53.74
.495	9	0.30	54.03
.4956667	1	0.03	54.07
.4958333	1	0.03	54.10
.496	1	0.03	54.13
.4966667	8	0.26	54.40
.498	1	0.03	54.43
.4983333	5	0.16	54.59
.499	1	0.03	54.63
.5	9	0.30	54.92
.5007408	1	0.03	54.96
.501	1	0.03	54.99
.5013334	1	0.03	55.02
.5016667	5	0.16	55.19
.5018519	1	0.03	55.22
.5026667	1	0.03	55.25
.5033333	6	0.20	55.45
.5037037	1	0.03	55.48
.5040741	1	0.03	55.52
.505	5	0.16	55.68

.506	1	0.03	55.71
.5066667	6	0.20	55.91
.5083333	10	0.33	56.24
.5086667	1	0.03	56.27
.51	4	0.13	56.40
.5116667	4	0.13	56.54
.512	1	0.03	56.57
.512963	1	0.03	56.60
.5133333	4	0.13	56.73
.5142857	1	0.03	56.77
.515	6	0.20	56.96
.516	1	0.03	57.00
.5163333	2	0.07	57.06
.5166667	16	0.53	57.59
.5176667	1	0.03	57.62
.5183333	5	0.16	57.79
.5185185	1	0.03	57.82
.52	3	0.10	57.92
.5206667	1	0.03	57.95
.5216666	3	0.10	58.05
.5233333	5	0.16	58.22
.5244445	1	0.03	58.25
.525	2	0.07	58.31
.525	2	0.07	58.38
.5253333	1	0.03	58.41
.5263333	1	0.03	58.45
.5266666	6	0.20	58.64
.5266667	1	0.03	58.68
.5277778	1	0.03	58.71
.5281482	1	0.03	58.74
.5283333	6	0.20	58.94
.5283334	1	0.03	58.97
.529	2	0.07	59.04
.5296296	1	0.03	59.07
.53	1	0.03	59.10
.53	8	0.26	59.37
.5303333	1	0.03	59.40
.5306666	3	0.10	59.50
.531	1	0.03	59.53
.5313333	1	0.03	59.57
.5316666	1	0.03	59.60
.5316667	4	0.13	59.73
.5323333	1	0.03	59.76
.5326667	2	0.07	59.83
.5333333	1	0.03	59.86
.5333334	9	0.30	60.16
.5343333	1	0.03	60.19
.535	5	0.16	60.36
.536	1	0.03	60.39
.5366667	4	0.13	60.52
.537	1	0.03	60.55
.5373333	2	0.07	60.62
.5383334	3	0.10	60.72
.5388889	1	0.03	60.75
.5393333	1	0.03	60.78
.5396667	1	0.03	60.82
.54	3	0.10	60.92
.5416667	8	0.26	61.18
.5433334	3	0.10	61.28
.5436667	1	0.03	61.31
.544	3	0.10	61.41
.5446666	1	0.03	61.44
.545	7	0.23	61.67
.5453333	1	0.03	61.71
.5456667	1	0.03	61.74
.5466667	5	0.16	61.90
.547	2	0.07	61.97

.5473334	1	0.03	62.00
.548	1	0.03	62.03
.5483333	4	0.13	62.17
.5496666	2	0.07	62.23
.55	24	0.79	63.02
.5533333	2	0.07	63.09
.5543333	1	0.03	63.12
.555	5	0.16	63.29
.5551852	1	0.03	63.32
.5553333	1	0.03	63.35
.5555556	3	0.10	63.45
.5566667	3	0.10	63.55
.5573334	1	0.03	63.58
.5574074	1	0.03	63.62
.5576667	1	0.03	63.65
.558	1	0.03	63.68
.5583333	8	0.26	63.94
.559	1	0.03	63.98
.5592592	1	0.03	64.01
.5596667	1	0.03	64.04
.56	4	0.13	64.18
.561	1	0.03	64.21
.5611111	1	0.03	64.24
.5616667	7	0.23	64.47
.563	1	0.03	64.50
.5633333	8	0.26	64.77
.564	3	0.10	64.87
.5648148	1	0.03	64.90
.565	4	0.13	65.03
.5666667	6	0.20	65.23
.568	1	0.03	65.26
.5683333	4	0.13	65.39
.5693333	1	0.03	65.43
.5696667	2	0.07	65.49
.57	3	0.10	65.59
.5706667	1	0.03	65.62
.5716667	3	0.10	65.72
.572	2	0.07	65.79
.5723333	1	0.03	65.82
.5729167	1	0.03	65.85
.5733333	3	0.10	65.95
.5743333	1	0.03	65.99
.575	5	0.16	66.15
.5756667	1	0.03	66.18
.5766667	1	0.03	66.22
.5770833	1	0.03	66.25
.5773333	1	0.03	66.28
.578	1	0.03	66.32
.5783333	6	0.20	66.51
.5786667	1	0.03	66.55
.579	1	0.03	66.58
.58	11	0.36	66.94
.5806667	1	0.03	66.97
.5816666	7	0.23	67.20
.5833333	15	0.49	67.70
.584	1	0.03	67.73
.585	2	0.07	67.80
.5851852	1	0.03	67.83
.5866666	5	0.16	67.99
.587	1	0.03	68.03
.587037	1	0.03	68.06
.588	1	0.03	68.09
.5883333	7	0.23	68.32
.5883334	4	0.13	68.46
.5888889	1	0.03	68.49
.5893334	1	0.03	68.52
.59	3	0.10	68.62

.59	1	0.03	68.65
.5903333	1	0.03	68.69
.5906667	1	0.03	68.72
.5907407	1	0.03	68.75
.5916666	6	0.20	68.95
.5916667	10	0.33	69.28
.5923333	1	0.03	69.31
.5925926	1	0.03	69.34
.593	1	0.03	69.38
.5933334	3	0.10	69.48
.594	1	0.03	69.51
.5946667	1	0.03	69.54
.595	5	0.16	69.71
.5956666	1	0.03	69.74
.5966667	8	0.26	70.00
.5973333	1	0.03	70.04
.5974074	1	0.03	70.07
.5977778	1	0.03	70.10
.5983334	4	0.13	70.23
.5996667	1	0.03	70.27
.6	4	0.13	70.40
.6016667	7	0.23	70.63
.6018519	1	0.03	70.66
.6023333	1	0.03	70.69
.6033334	9	0.30	70.99
.604	1	0.03	71.02
.605	3	0.10	71.12
.6053333	1	0.03	71.16
.6055555	1	0.03	71.19
.6056666	1	0.03	71.22
.6066667	4	0.13	71.35
.607	1	0.03	71.39
.6073333	1	0.03	71.42
.6074074	1	0.03	71.45
.6083333	7	0.23	71.68
.6086667	1	0.03	71.72
.61	2	0.07	71.78
.6103333	1	0.03	71.81
.6111111	1	0.03	71.85
.6116667	4	0.13	71.98
.6133333	4	0.13	72.11
.614	4	0.13	72.24
.6143333	1	0.03	72.28
.615	4	0.13	72.41
.6166667	8	0.26	72.67
.6176667	1	0.03	72.70
.6183333	6	0.20	72.90
.619	2	0.07	72.97
.62	5	0.16	73.13
.6206667	1	0.03	73.16
.6216667	6	0.20	73.36
.622	1	0.03	73.39
.6226667	1	0.03	73.43
.6233333	6	0.20	73.63
.624	1	0.03	73.66
.6246667	1	0.03	73.69
.625	5	0.16	73.86
.6266667	5	0.16	74.02
.6273333	1	0.03	74.05
.628	1	0.03	74.09
.6283333	6	0.20	74.28
.6296296	1	0.03	74.32
.63	7	0.23	74.55
.6306667	1	0.03	74.58
.6314815	1	0.03	74.61
.6316667	5	0.16	74.78
.6323333	1	0.03	74.81

.6333333	25	0.82	75.63
.6343333	1	0.03	75.67
.635	2	0.07	75.73
.6356667	1	0.03	75.77
.6366667	9	0.30	76.06
.6383333	8	0.26	76.33
.6386667	1	0.03	76.36
.639	3	0.10	76.46
.6396667	1	0.03	76.49
.64	3	0.10	76.59
.641	1	0.03	76.62
.6416667	9	0.30	76.92
.6433333	7	0.23	77.15
.645	1	0.03	77.18
.6456667	1	0.03	77.21
.646	1	0.03	77.25
.6463333	1	0.03	77.28
.6466666	6	0.20	77.48
.647	1	0.03	77.51
.6481482	1	0.03	77.54
.6483333	5	0.16	77.71
.6486667	2	0.07	77.77
.649	1	0.03	77.81
.65	3	0.10	77.91
.6506667	1	0.03	77.94
.651	1	0.03	77.97
.6516666	3	0.10	78.07
.6518518	1	0.03	78.10
.6523333	2	0.07	78.17
.6533333	6	0.20	78.37
.6533334	2	0.07	78.43
.655	5	0.16	78.60
.656	1	0.03	78.63
.6563333	1	0.03	78.66
.6566666	1	0.03	78.70
.6566667	2	0.07	78.76
.6574074	1	0.03	78.79
.6583334	4	0.13	78.93
.6592593	1	0.03	78.96
.6596296	1	0.03	78.99
.6596667	1	0.03	79.03
.66	3	0.10	79.12
.6603333	2	0.07	79.19
.6616667	7	0.23	79.42
.6633334	3	0.10	79.52
.665	5	0.16	79.68
.6656667	1	0.03	79.72
.6666667	9	0.30	80.01
.667	1	0.03	80.05
.668	1	0.03	80.08
.6683334	1	0.03	80.11
.6696666	2	0.07	80.18
.67	4	0.13	80.31
.6716667	2	0.07	80.38
.672	1	0.03	80.41
.6733333	7	0.23	80.64
.6736667	1	0.03	80.67
.6746666	2	0.07	80.74
.675	14	0.46	81.20
.6766667	3	0.10	81.30
.677	1	0.03	81.33
.6773334	1	0.03	81.36
.678	1	0.03	81.40
.6783333	5	0.16	81.56
.679	1	0.03	81.59
.6796667	1	0.03	81.63
.68	6	0.20	81.82

.6813334	1	0.03	81.86
.6816667	3	0.10	81.96
.6818519	1	0.03	81.99
.682	1	0.03	82.02
.6833333	10	0.33	82.35
.684	1	0.03	82.38
.685	2	0.07	82.45
.6856667	1	0.03	82.48
.6866667	7	0.23	82.71
.6876667	2	0.07	82.78
.6883333	4	0.13	82.91
.69	5	0.16	83.08
.6916667	3	0.10	83.17
.694	2	0.07	83.24
.6951852	1	0.03	83.27
.6966667	6	0.20	83.47
.697	1	0.03	83.50
.6973333	2	0.07	83.57
.6983333	8	0.26	83.83
.6986667	1	0.03	83.87
.6988889	1	0.03	83.90
.7	15	0.49	84.39
.7006667	1	0.03	84.43
.701	1	0.03	84.46
.7016667	1	0.03	84.49
.7023333	2	0.07	84.56
.7033333	5	0.16	84.72
.7037037	1	0.03	84.75
.704	2	0.07	84.82
.705	6	0.20	85.02
.7063333	1	0.03	85.05
.7066666	3	0.10	85.15
.7074074	1	0.03	85.18
.7083333	3	0.10	85.28
.71	4	0.13	85.41
.7113333	3	0.10	85.51
.713	1	0.03	85.54
.7133333	3	0.10	85.64
.715	3	0.10	85.74
.715	2	0.07	85.81
.7166667	4	0.13	85.94
.7183333	1	0.03	85.97
.7196667	1	0.03	86.01
.72	7	0.23	86.24
.7206666	1	0.03	86.27
.7213333	1	0.03	86.30
.7216667	2	0.07	86.37
.7223333	1	0.03	86.40
.7233334	2	0.07	86.47
.7246667	1	0.03	86.50
.725	5	0.16	86.66
.7266667	2	0.07	86.73
.7276667	1	0.03	86.76
.7283334	1	0.03	86.80
.729	1	0.03	86.83
.73	6	0.20	87.03
.731	1	0.03	87.06
.7313333	1	0.03	87.09
.7314815	1	0.03	87.13
.7316667	5	0.16	87.29
.732	2	0.07	87.36
.7323333	1	0.03	87.39
.7323334	2	0.07	87.45
.7326667	1	0.03	87.49
.7333333	6	0.20	87.69
.734	1	0.03	87.72
.735	1	0.03	87.75

.7363334	1	0.03	87.78
.7366667	2	0.07	87.85
.7373334	1	0.03	87.88
.7383333	5	0.16	88.05
.739	3	0.10	88.15
.7396667	1	0.03	88.18
.74	4	0.13	88.31
.7416667	3	0.10	88.41
.7433333	4	0.13	88.54
.745	2	0.07	88.61
.7462963	1	0.03	88.64
.7466667	8	0.26	88.90
.7473333	1	0.03	88.94
.7483333	3	0.10	89.04
.7486666	1	0.03	89.07
.749	2	0.07	89.13
.7493333	1	0.03	89.17
.75	15	0.49	89.66
.7516667	3	0.10	89.76
.7518519	1	0.03	89.79
.752	1	0.03	89.83
.7523333	1	0.03	89.86
.7526667	1	0.03	89.89
.7533333	1	0.03	89.92
.755	1	0.03	89.96
.7556667	1	0.03	89.99
.7563334	1	0.03	90.02
.7566667	3	0.10	90.12
.7573333	1	0.03	90.15
.7583333	4	0.13	90.29
.759	1	0.03	90.32
.76	1	0.03	90.35
.7603334	1	0.03	90.39
.7606667	1	0.03	90.42
.7616667	5	0.16	90.58
.7623333	1	0.03	90.62
.7633333	2	0.07	90.68
.765	4	0.13	90.81
.7666667	7	0.23	91.04
.7673333	1	0.03	91.08
.7683333	1	0.03	91.11
.7685185	1	0.03	91.14
.77	5	0.16	91.31
.7706667	1	0.03	91.34
.7716666	4	0.13	91.47
.7722222	1	0.03	91.50
.7733333	1	0.03	91.54
.775	1	0.03	91.57
.7756667	1	0.03	91.60
.7766666	1	0.03	91.64
.7777778	1	0.03	91.67
.778	1	0.03	91.70
.779	1	0.03	91.74
.78	1	0.03	91.77
.7806666	1	0.03	91.80
.7816666	1	0.03	91.83
.7816667	2	0.07	91.90
.7823333	1	0.03	91.93
.783	1	0.03	91.97
.7833333	1	0.03	92.00
.7833334	2	0.07	92.06
.784	1	0.03	92.10
.785	3	0.10	92.20
.7866667	5	0.16	92.36
.7876667	1	0.03	92.39
.788	1	0.03	92.43
.79	2	0.07	92.49

.7907407	1	0.03	92.53
.7913333	1	0.03	92.56
.7916667	4	0.13	92.69
.7933334	1	0.03	92.72
.795	3	0.10	92.82
.7966667	6	0.20	93.02
.797	1	0.03	93.05
.7983333	4	0.13	93.18
.7996666	1	0.03	93.22
.8	7	0.23	93.45
.8016667	3	0.10	93.55
.804	1	0.03	93.58
.8055556	1	0.03	93.61
.8056667	1	0.03	93.65
.8066667	2	0.07	93.71
.8073334	1	0.03	93.74
.8081481	1	0.03	93.78
.8083333	4	0.13	93.91
.81	1	0.03	93.94
.8106667	1	0.03	93.97
.8116667	1	0.03	94.01
.8123333	1	0.03	94.04
.8125	1	0.03	94.07
.8133333	2	0.07	94.14
.815	1	0.03	94.17
.8166667	2	0.07	94.24
.82	1	0.03	94.27
.8206667	1	0.03	94.30
.8233333	1	0.03	94.34
.8240741	1	0.03	94.37
.825	4	0.13	94.50
.8259259	1	0.03	94.53
.8283333	1	0.03	94.57
.829	1	0.03	94.60
.83	5	0.16	94.76
.8316666	2	0.07	94.83
.8333333	7	0.23	95.06
.8366666	1	0.03	95.09
.8383333	2	0.07	95.16
.8395833	1	0.03	95.19
.84	1	0.03	95.23
.8416667	2	0.07	95.29
.8456666	1	0.03	95.32
.8466667	1	0.03	95.36
.848	1	0.03	95.39
.8483334	1	0.03	95.42
.849	1	0.03	95.46
.85	6	0.20	95.65
.8507407	1	0.03	95.69
.8518519	1	0.03	95.72
.8533334	1	0.03	95.75
.855	1	0.03	95.79
.8566667	1	0.03	95.82
.8573334	1	0.03	95.85
.8583333	4	0.13	95.98
.8596666	1	0.03	96.02
.86	1	0.03	96.05
.8616667	1	0.03	96.08
.862963	1	0.03	96.11
.8633333	1	0.03	96.15
.8666667	7	0.23	96.38
.87	1	0.03	96.41
.8703704	1	0.03	96.44
.871	1	0.03	96.48
.8716667	3	0.10	96.58
.8726667	1	0.03	96.61
.8733333	1	0.03	96.64

.8746667	1	0.03	96.67
.875	9	0.30	96.97
.8756667	1	0.03	97.00
.8766667	1	0.03	97.04
.88	4	0.13	97.17
.8833333	2	0.07	97.23
.8846667	1	0.03	97.27
.8866667	1	0.03	97.30
.8883333	1	0.03	97.33
.8888889	2	0.07	97.40
.8896667	1	0.03	97.43
.8916667	2	0.07	97.50
.892	1	0.03	97.53
.8933333	1	0.03	97.56
.8946667	1	0.03	97.60
.895	1	0.03	97.63
.898	1	0.03	97.66
.9	6	0.20	97.86
.9033333	1	0.03	97.89
.9066667	1	0.03	97.93
.9133334	1	0.03	97.96
.9156667	1	0.03	97.99
.9166666	1	0.03	98.02
.9166667	6	0.20	98.22
.922	1	0.03	98.25
.925	3	0.10	98.35
.9323334	1	0.03	98.39
.9333333	3	0.10	98.49
.937	1	0.03	98.52
.945	1	0.03	98.55
.9466667	2	0.07	98.62
.95	3	0.10	98.72
.955	2	0.07	98.78
.9573333	1	0.03	98.81
.9583333	2	0.07	98.88
.96	2	0.07	98.95
.9636667	1	0.03	98.98
.972	1	0.03	99.01
.974	1	0.03	99.05
.975	3	0.10	99.14
.98	1	0.03	99.18
.998	1	0.03	99.21
1	24	0.79	100.00
Total	3,037	100.00	

-> tabulation of immf_5789 if white==1

immf_5789	Freq.	Percent	Cum.
0	59	2.25	2.25
.0625	98	3.73	5.97
.125	152	5.78	11.76
.1666667	5	0.19	11.95
.1875	146	5.56	17.50
.25	180	6.85	24.35
.3125	238	9.06	33.41
.3333333	5	0.19	33.60
.375	235	8.94	42.54
.4166667	2	0.08	42.62
.4375	204	7.76	50.38
.5	249	9.47	59.86
.5625	254	9.67	69.52
.625	222	8.45	77.97
.6666667	3	0.11	78.08
.6875	178	6.77	84.86
.75	145	5.52	90.37

.8125	111	4.22	94.60
.8333333	1	0.04	94.63
.875	55	2.09	96.73
.9375	36	1.37	98.10
1	50	1.90	100.00
Total	2,628	100.00	

-> tabulation of immf_5678 if white==1

immf_5678	Freq.	Percent	Cum.
0	51	1.94	1.94
.0625	82	3.12	5.06
.0833333	15	0.57	5.63
.125	100	3.81	9.44
.1458333	37	1.41	10.84
.1666667	2	0.08	10.92
.1875	101	3.84	14.76
.2083333	87	3.31	18.07
.2291667	1	0.04	18.11
.25	96	3.65	21.77
.2708333	141	5.37	27.13
.2777778	4	0.15	27.28
.2916667	8	0.30	27.59
.3125	44	1.67	29.26
.3333333	188	7.15	36.42
.3541667	25	0.95	37.37
.3611111	1	0.04	37.40
.375	39	1.48	38.89
.3958333	154	5.86	44.75
.4166667	47	1.79	46.54
.4375	30	1.14	47.68
.4444444	2	0.08	47.75
.4583333	166	6.32	54.07
.4791667	64	2.44	56.51
.5	29	1.10	57.61
.5208333	111	4.22	61.83
.5416667	84	3.20	65.03
.5555556	2	0.08	65.11
.5625	22	0.84	65.94
.5833333	85	3.23	69.18
.6041667	93	3.54	72.72
.6111111	1	0.04	72.75
.625	35	1.33	74.09
.6458333	49	1.86	75.95
.6666667	105	4.00	79.95
.6875	60	2.28	82.23
.6944444	1	0.04	82.27
.7083333	29	1.10	83.37
.7083334	1	0.04	83.41
.7222222	2	0.08	83.49
.7291667	64	2.44	85.92
.75	88	3.35	89.27
.7708333	8	0.30	89.57
.7916667	42	1.60	91.17
.8125	45	1.71	92.88
.8333333	5	0.19	93.07
.8333334	2	0.08	93.15
.8541667	16	0.61	93.76
.875	67	2.55	96.31
.9166667	10	0.38	96.69
.9375	38	1.45	98.14
1	49	1.86	100.00
Total	2,628	100.00	

-> tabulation of immf_6789 if white==1

immf_6789	Freq.	Percent	Cum.
0	157	5.98	5.98
.0625	98	3.73	9.71
.0833333	59	2.25	11.96
.1111111	3	0.11	12.07
.125	78	2.97	15.04
.1458333	102	3.88	18.93
.1666667	3	0.11	19.04
.1875	77	2.93	21.97
.1944444	1	0.04	22.01
.2083333	92	3.50	25.51
.2291667	12	0.46	25.97
.25	55	2.09	28.07
.2708333	150	5.71	33.78
.2916667	17	0.65	34.42
.3125	42	1.60	36.02
.3333333	176	6.70	42.73
.3541667	38	1.45	44.17
.3611111	1	0.04	44.21
.375	32	1.22	45.43
.3888889	1	0.04	45.47
.3958333	140	5.33	50.80
.4166667	51	1.94	52.74
.4375	32	1.22	53.96
.4444444	2	0.08	54.04
.4583333	151	5.75	59.79
.4791667	95	3.62	63.40
.5	29	1.10	64.51
.5208333	102	3.88	68.39
.5416667	101	3.85	72.24
.5625	25	0.95	73.19
.5833333	55	2.09	75.29
.6041667	85	3.24	78.52
.6111111	1	0.04	78.56
.625	72	2.74	81.30
.6388889	1	0.04	81.34
.6458333	32	1.22	82.56
.6666667	70	2.67	85.22
.6875	58	2.21	87.43
.7083333	10	0.38	87.81
.7083334	1	0.04	87.85
.7291667	49	1.87	89.72
.75	68	2.59	92.31
.7708333	6	0.23	92.54
.7916667	26	0.99	93.53
.8125	55	2.09	95.62
.8333333	3	0.11	95.73
.8541667	6	0.23	95.96
.875	32	1.22	97.18
.9166667	8	0.30	97.49
.9375	25	0.95	98.44
1	41	1.56	100.00
Total	2,626	100.00	

-> tabulation of rrf if white==1

rrf	Freq.	Percent	Cum.
0	116	4.41	4.41
.0625	74	2.82	7.23
.125	63	2.40	9.63
.1666667	1	0.04	9.67
.1875	77	2.93	12.60
.25	125	4.76	17.35
.3125	111	4.22	21.58
.3333333	1	0.04	21.61
.375	135	5.14	26.75
.4375	149	5.67	32.42
.5	259	9.86	42.28
.5625	178	6.77	49.05
.5833333	2	0.08	49.12
.625	211	8.03	57.15
.6875	199	7.57	64.73
.75	209	7.95	72.68
.8125	145	5.52	78.20
.8333333	1	0.04	78.23
.875	186	7.08	85.31
.9166667	1	0.04	85.35
.9375	143	5.44	90.79
1	242	9.21	100.00
Total	2,628	100.00	

30 . codebook immf immf_5789 immf_5678 immf_6789 rrf if white==1

immf

```

type:  numeric (float)

range:  [0,1]                      units:  1.000e-09
unique values:  1,096              missing .:  1/3,038

mean:    .465359
std. dev: .224861

percentiles:      10%      25%      50%      75%      90%
                  .158333  .291667  .471667  .633333  .756333

```

immf_5789

```

type:  numeric (float)

range:  [0,1]                      units:  1.000e-08
unique values:  22                 missing .:  410/3,038

mean:    .462226
std. dev: .237457

percentiles:      10%      25%      50%      75%      90%
                  .125     .3125   .4375   .625     .75

```

immf_5678

```

type:  numeric (float)

```

```

range: [0,1] units: 1.000e-09
unique values: 52 missing .: 410/3,038

mean: .462516
std. dev: .237973

percentiles: 10% 25% 50% 75% 90%
               .145833 .270833 .458333 .645833 .791667

```

immf_6789

```

type: numeric (float)

range: [0,1] units: 1.000e-09
unique values: 51 missing .: 412/3,038

mean: .411946
std. dev: .243468

percentiles: 10% 25% 50% 75% 90%
               .083333 .208333 .395833 .583333 .75

```

rrf

```

type: numeric (float)

range: [0,1] units: 1.000e-08
unique values: 22 missing .: 410/3,038

mean: .579829
std. dev: .282653

percentiles: 10% 25% 50% 75% 90%
               .1875 .375 .625 .8125 .9375

```

31 .

32 . logit votechoice jardina3 rrf immf_5789 sexismf econ1 age3044 age4564 age65p female prot south b
> k pid1 ideol1 if white==1 [pw=wgt]

```

Iteration 0: log pseudolikelihood = -1196.262
Iteration 1: log pseudolikelihood = -387.54741
Iteration 2: log pseudolikelihood = -360.02443
Iteration 3: log pseudolikelihood = -359.39375
Iteration 4: log pseudolikelihood = -359.3915
Iteration 5: log pseudolikelihood = -359.3915

```

```

Logistic regression                                Number of obs    =      1,879
                                                    Wald chi2(16)    =      334.37
                                                    Prob > chi2      =      0.0000
Log pseudolikelihood = -359.3915                Pseudo R2       =      0.6996

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	- .8934712	.5291583	-1.69	0.091	-1.930603	.1436601
rrf	3.200042	.6619346	4.83	0.000	1.902674	4.49741
immf_5789	3.33652	.7514072	4.44	0.000	1.863789	4.809251
sexismf	3.752767	.8388008	4.47	0.000	2.108748	5.396786
econ1	-1.558281	.5983106	-2.60	0.009	-2.730948	-.3856135
age3044	.4986321	.5395817	0.92	0.355	-.5589286	1.556193
age4564	.3746528	.548881	0.68	0.495	-.7011342	1.45044
age65p	.5220936	.5226132	1.00	0.318	-.5022095	1.546397

female	.2566126	.223	1.15	0.250	-.1804594	.6936845
prot	-.1651779	.2516207	-0.66	0.512	-.6583455	.3279896
south	.7348012	.2699526	2.72	0.006	.2057038	1.263899
ba	-.1369994	.2415311	-0.57	0.571	-.6103916	.3363928
incl	-.5713716	.5127834	-1.11	0.265	-1.576409	.4336653
incdk	-.3672242	.6006368	-0.61	0.541	-1.544451	.8100022
pid1	4.675405	.4133232	11.31	0.000	3.865306	5.485504
ideoll	3.150024	.6317908	4.99	0.000	1.911737	4.388311
_cons	-7.757936	.8423995	-9.21	0.000	-9.409008	-6.106863

```
33 . margins, atmeans at(rrf=(0 1))
```

Adjusted predictions	Number of obs	=	1,879
Model VCE : Robust			

Expression : `Pr(votechoice), predict()`

```
1._at      : jardina3      = .4270238 (mean)
            rrf            = 0
            immf_5789      = .4603547 (mean)
            sexismf        = .3719823 (mean)
            econ1          = .4743608 (mean)
            age3044        = .1959655 (mean)
            age4564        = .4108636 (mean)
            age65p         = .2584021 (mean)
            female         = .5317754 (mean)
            prot           = .3606696 (mean)
            south          = .2612745 (mean)
            ba             = .4006297 (mean)
            incl           = .6104942 (mean)
            incdk          = .0277694 (mean)
            pid1           = .5391016 (mean)
            ideol1         = .5575229 (mean)
```

```
2._at      : jardina3      =      .4270238 (mean)
            rrf            =              1
            immf_5789      =      .4603547 (mean)
            sexismf        =      .3719823 (mean)
            econ1          =      .4743608 (mean)
            age3044        =      .1959655 (mean)
            age4564        =      .4108636 (mean)
            age65p         =      .2584021 (mean)
            female         =      .5317754 (mean)
            prot           =      .3606696 (mean)
            south          =      .2612745 (mean)
            ba             =      .4006297 (mean)
            incl           =      .6104942 (mean)
            incdk          =      .0277694 (mean)
            pid1           =      .5391016 (mean)
            ideol1         =      .5575229 (mean)
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at_1	.193339	.064432	3.00	0.003	.0670547	.3196233
2	.8546544	.0360758	23.69	0.000	.7839472	.9253617

```
34 . di 85.5-19.3 // 66.2
66.2
```

```
35 . margins, atmeans at(immf_5789=(0 1))
```

```
Adjusted predictions                                Number of obs      =      1,879
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at      : jardina3      =      .4270238 (mean)
              rrf           =      .5749568 (mean)
              immf_5789     =      0
              sexismf       =      .3719823 (mean)
              econ1         =      .4743608 (mean)
              age3044       =      .1959655 (mean)
              age4564       =      .4108636 (mean)
              age65p        =      .2584021 (mean)
              female        =      .5317754 (mean)
              prot          =      .3606696 (mean)
              south         =      .2612745 (mean)
              ba            =      .4006297 (mean)
              incl          =      .6104942 (mean)
              incdk         =      .0277694 (mean)
              pid1          =      .5391016 (mean)
              ideoll        =      .5575229 (mean)
```

```
2._at      : jardina3      =      .4270238 (mean)
              rrf           =      .5749568 (mean)
              immf_5789     =      1
              sexismf       =      .3719823 (mean)
              econ1         =      .4743608 (mean)
              age3044       =      .1959655 (mean)
              age4564       =      .4108636 (mean)
              age65p        =      .2584021 (mean)
              female        =      .5317754 (mean)
              prot          =      .3606696 (mean)
              south         =      .2612745 (mean)
              ba            =      .4006297 (mean)
              incl          =      .6104942 (mean)
              incdk         =      .0277694 (mean)
              pid1          =      .5391016 (mean)
              ideoll        =      .5575229 (mean)
```

	Delta-method				
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]
_at					
1	.2451673	.0654577	3.75	0.000	.1168725 .373462
2	.9013188	.0386017	23.35	0.000	.8256608 .9769768

```
36 . di 90.1-24.5 // 65.6
65.6
```

```

37 .
38 . logit votechoice jardina3 rrf immf_5678 sexismf econ1 age3044 age4564 age65p female prot south b
> k pid1 ideoll if white==1 [pw=wt]

```

```

Iteration 0: log pseudolikelihood = -1196.262
Iteration 1: log pseudolikelihood = -388.01518
Iteration 2: log pseudolikelihood = -360.6606
Iteration 3: log pseudolikelihood = -360.08154
Iteration 4: log pseudolikelihood = -360.0795
Iteration 5: log pseudolikelihood = -360.0795

```

```

Logistic regression                                Number of obs      =      1,879
                                                    Wald chi2(16)      =      341.66
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -360.0795                Pseudo R2         =      0.6990

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-.8962571	.5263225	-1.70	0.089	-1.92783	.1353159
rrf	3.249783	.6591876	4.93	0.000	1.957799	4.541767
immf_5678	3.199017	.7607991	4.20	0.000	1.707879	4.690156
sexismf	3.69491	.8683772	4.25	0.000	1.992922	5.396898
econ1	-1.439674	.6039905	-2.38	0.017	-2.623474	-.2558744
age3044	.4607871	.5404199	0.85	0.394	-.5984165	1.519991
age4564	.2980099	.5463123	0.55	0.585	-.7727426	1.368762
age65p	.4850477	.5241259	0.93	0.355	-.5422202	1.512316
female	.2318564	.2232667	1.04	0.299	-.2057383	.669451
prot	-.1439861	.249836	-0.58	0.564	-.6336557	.3456835
south	.7048994	.2633202	2.68	0.007	.1888013	1.220997
ba	-.1825257	.2415895	-0.76	0.450	-.6560325	.2909811
incl	-.471506	.5137467	-0.92	0.359	-1.478431	.5354191
incdk	-.293999	.5818877	-0.51	0.613	-1.434478	.84648
pid1	4.730971	.4226847	11.19	0.000	3.902524	5.559418
ideoll	3.134367	.63018	4.97	0.000	1.899237	4.369497
_cons	-7.730128	.850739	-9.09	0.000	-9.397546	-6.06271

```

39 . margins, atmeans at(rrf=(0 1))

```

```

Adjusted predictions                                Number of obs      =      1,879
Model VCE      : Robust

```

```

Expression      : Pr(votechoice), predict()

```

```

1._at      : jardina3      =      .4270238 (mean)
              rrf          =      0
              immf_5678    =      .4605256 (mean)
              sexismf      =      .3719823 (mean)
              econ1        =      .4743608 (mean)
              age3044      =      .1959655 (mean)
              age4564      =      .4108636 (mean)
              age65p       =      .2584021 (mean)
              female       =      .5317754 (mean)
              prot         =      .3606696 (mean)
              south        =      .2612745 (mean)
              ba           =      .4006297 (mean)
              incl         =      .6104942 (mean)
              incdk        =      .0277694 (mean)
              pid1         =      .5391016 (mean)
              ideoll       =      .5575229 (mean)

```

```

2._at      : jardina3      = .4270238 (mean)
             rrf            = 1
             immf_5678      = .4605256 (mean)
             sexismf        = .3719823 (mean)
             econl          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)
             pidl           = .5391016 (mean)
             ideoll         = .5575229 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.1937731	.064659	3.00	0.003	.0670437	.3205024
2	.8610581	.0344107	25.02	0.000	.7936144	.9285017

```

40 . di 86.1-19.4 // 66.7
    66.7

```

```

41 . margins, atmeans at(immf_5678=(0 1))

```

```

Adjusted predictions      Number of obs      =      1,879
Model VCE      : Robust

```

```

Expression      : Pr(votechoice), predict()

```

```

1._at      : jardina3      = .4270238 (mean)
             rrf            = .5749568 (mean)
             immf_5678      = 0
             sexismf        = .3719823 (mean)
             econl          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)
             pidl           = .5391016 (mean)
             ideoll         = .5575229 (mean)

```

```

2._at      : jardina3      = .4270238 (mean)
             rrf            = .5749568 (mean)
             immf_5678      = 1
             sexismf        = .3719823 (mean)
             econl          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)

```

```

pid1          = .5391016 (mean)
ideoll        = .5575229 (mean)

```

	Delta-method				[95% Conf. Interval]	
	Margin	Std. Err.	z	P> z		
_at						
1	.2630023	.0677729	3.88	0.000	.1301699	.3958347
2	.8973937	.0411627	21.80	0.000	.8167163	.9780711

```

42 . di 89.7-26.3 // 63.4
63.4

```

```

43 .
44 . logit votechoice jardina3 rrf immf_6789 sexismf econ1 age3044 age4564 age65p female prot south b
> k pid1 ideoll if white==1 [pw=wgt]

```

```

Iteration 0:  log pseudolikelihood = -1196.262
Iteration 1:  log pseudolikelihood = -386.49053
Iteration 2:  log pseudolikelihood = -361.33153
Iteration 3:  log pseudolikelihood = -358.40524
Iteration 4:  log pseudolikelihood = -358.39975
Iteration 5:  log pseudolikelihood = -358.39974

```

```

Logistic regression                                Number of obs      =      1,879
                                                    Wald chi2(16)      =      331.93
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -358.39974                Pseudo R2         =      0.7004

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-.9613375	.5310955	-1.81	0.070	-2.002266	.0795907
rrf	3.293759	.6558286	5.02	0.000	2.008359	4.57916
immf_6789	3.323149	.7555902	4.40	0.000	1.84222	4.804079
sexismf	3.562853	.8468	4.21	0.000	1.903156	5.222551
econ1	-1.601575	.6010963	-2.66	0.008	-2.779702	-.4234477
age3044	.4736316	.5316801	0.89	0.373	-.5684421	1.515705
age4564	.3508962	.5403207	0.65	0.516	-.708113	1.409905
age65p	.5629439	.5124748	1.10	0.272	-.4414883	1.567376
female	.2964189	.22449	1.32	0.187	-.1435735	.7364113
prot	-.1491777	.2527007	-0.59	0.555	-.6444619	.3461065
south	.7042978	.2671634	2.64	0.008	.1806672	1.227928
ba	-.1906208	.2421134	-0.79	0.431	-.6651544	.2839128
incl	-.4106474	.5178472	-0.79	0.428	-1.425609	.6043144
incdk	-.298963	.5956766	-0.50	0.616	-1.466468	.8685416
pid1	4.738559	.4258537	11.13	0.000	3.903902	5.573217
ideoll	3.110998	.6301477	4.94	0.000	1.875931	4.346065
_cons	-7.605656	.8320034	-9.14	0.000	-9.236353	-5.974959

```

45 . margins, atmeans at(rrf=(0 1))

```

```

Adjusted predictions                                Number of obs      =      1,879
Model VCE      : Robust

Expression      : Pr(votechoice), predict()

```

```

1._at      : jardina3      = .4270238 (mean)
              rrf           =          0
              immf_6789     = .4099409 (mean)
              sexismf       = .3719823 (mean)
              econ1         = .4743608 (mean)
              age3044       = .1959655 (mean)
              age4564       = .4108636 (mean)
              age65p        = .2584021 (mean)
              female        = .5317754 (mean)
              prot          = .3606696 (mean)
              south         = .2612745 (mean)
              ba            = .4006297 (mean)
              incl          = .6104942 (mean)
              incdk         = .0277694 (mean)
              pidl          = .5391016 (mean)
              ideoll        = .5575229 (mean)

2._at      : jardina3      = .4270238 (mean)
              rrf           =          1
              immf_6789     = .4099409 (mean)
              sexismf       = .3719823 (mean)
              econ1         = .4743608 (mean)
              age3044       = .1959655 (mean)
              age4564       = .4108636 (mean)
              age65p        = .2584021 (mean)
              female        = .5317754 (mean)
              prot          = .3606696 (mean)
              south         = .2612745 (mean)
              ba            = .4006297 (mean)
              incl          = .6104942 (mean)
              incdk         = .0277694 (mean)
              pidl          = .5391016 (mean)
              ideoll        = .5575229 (mean)

```

	Delta-method				
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]
_at					
1	.1878044	.0627678	2.99	0.003	.0647816 .3108271
2	.8616925	.0342098	25.19	0.000	.7946425 .9287425

```

46 . di 86.2-18.8 // 67.4
    67.4

```

```

47 . margins, atmeans at(immf_6789=(0 1))

```

```

Adjusted predictions      Number of obs      =      1,879
Model VCE      : Robust

Expression      : Pr(votechoice), predict()

1._at      : jardina3      = .4270238 (mean)
              rrf           = .5749568 (mean)
              immf_6789     =          0
              sexismf       = .3719823 (mean)
              econ1         = .4743608 (mean)
              age3044       = .1959655 (mean)
              age4564       = .4108636 (mean)
              age65p        = .2584021 (mean)
              female        = .5317754 (mean)
              prot          = .3606696 (mean)
              south         = .2612745 (mean)
              ba            = .4006297 (mean)
              incl          = .6104942 (mean)
              incdk         = .0277694 (mean)

```

```

pidl = .5391016 (mean)
ideoll = .5575229 (mean)

2._at : jardina3 = .4270238 (mean)
      rrf = .5749568 (mean)
      immf_6789 = 1
      sexismf = .3719823 (mean)
      econl = .4743608 (mean)
      age3044 = .1959655 (mean)
      age4564 = .4108636 (mean)
      age65p = .2584021 (mean)
      female = .5317754 (mean)
      prot = .3606696 (mean)
      south = .2612745 (mean)
      ba = .4006297 (mean)
      incl = .6104942 (mean)
      incdk = .0277694 (mean)
      pidl = .5391016 (mean)
      ideoll = .5575229 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.2823437	.0630385	4.48	0.000	.1587904	.4058969
2	.9160834	.0370507	24.73	0.000	.8434653	.9887015

```

48 . di 91.6-28.2 // 63.4
63.4

```

```

49 .
50 . *** Reduced xenophobia predictor with the border wall item
51 .
52 . pwcorr votechoice V161192 V161194x V161195x V161196x V162157 V162158 V162268 V162269 V162270 V162313
> black2 black3 black4 if white==1

```

	votech~e	V161192	V161194x	V161195x	V161196x	V162157	V162158	
votechoice	1.0000							
V161192	-0.4609	1.0000						
V161194x	-0.4034	0.3069	1.0000					
V161195x	-0.3908	0.5247	0.3236	1.0000				
V161196x	-0.7212	0.4719	0.4249	0.4463	1.0000			
V162157	0.5282	-0.4041	-0.3721	-0.3801	-0.5335	1.0000		
V162158	-0.4828	0.3900	0.2982	0.3654	0.4891	-0.5226	1.0000	
V162268	0.3846	-0.3571	-0.2627	-0.3339	-0.3840	0.5109	-0.4659	
V162269	-0.5057	0.4187	0.3383	0.4011	0.4920	-0.5214	0.5283	
V162270	-0.5565	0.4015	0.3438	0.3791	0.5438	-0.5216	0.5152	
V162313	-0.4780	0.4377	0.3891	0.4255	0.4865	-0.4987	0.4269	
black1	0.5823	-0.3904	-0.3348	-0.3580	-0.5261	0.4929	-0.4686	
black2	0.5038	-0.3467	-0.2968	-0.3475	-0.4198	0.4266	-0.3462	
black3	0.5338	-0.3484	-0.3109	-0.3327	-0.4595	0.4652	-0.3796	
black4	0.5497	-0.3816	-0.2997	-0.3484	-0.4840	0.4565	-0.4414	
	V162268	V162269	V162270	V162313	black1	black2	black3	
V162268	1.0000							
V162269	-0.6034	1.0000						
V162270	-0.5011	0.6656	1.0000					
V162313	-0.4059	0.4491	0.4641	1.0000				
black1	0.4014	-0.4644	-0.5002	-0.4760	1.0000			
black2	0.3305	-0.3632	-0.3823	-0.4153	0.5723	1.0000		
black3	0.3565	-0.3924	-0.4074	-0.4477	0.6192	0.6819	1.0000	
black4	0.3849	-0.4816	-0.4929	-0.4379	0.6808	0.5466	0.5819	

	black4
black4	1.0000

```

53 .
54 . egen immf_4789=rowmean(antiimm4 antiimm7 antiimm8 antiimm9)
    (9 missing values generated)

55 . logit votechoice jardina3 rrf immf_4789 sexismf econ1 age3044 age4564 age65p female prot south ba
    > k pid1 ideol1 if white==1 [pw=wt]

```

```

Iteration 0:  log pseudolikelihood =  -1196.262
Iteration 1:  log pseudolikelihood =  -348.17859
Iteration 2:  log pseudolikelihood =  -327.95917
Iteration 3:  log pseudolikelihood =  -326.73859
Iteration 4:  log pseudolikelihood =  -326.73657
Iteration 5:  log pseudolikelihood =  -326.73657

```

```

Logistic regression                                Number of obs      =      1,879
                                                    Wald chi2(16)      =      279.64
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -326.73657                Pseudo R2         =      0.7269

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-1.424939	.5670508	-2.51	0.012	-2.536338	-.3135394
rrf	2.799917	.6721558	4.17	0.000	1.482516	4.117318
immf_4789	5.628904	.7523407	7.48	0.000	4.154344	7.103465
sexismf	3.314644	.796082	4.16	0.000	1.754352	4.874936
econ1	-1.430176	.60667	-2.36	0.018	-2.619227	-.2411242
age3044	.3677599	.5680851	0.65	0.517	-.7456664	1.481186
age4564	.1817098	.5878878	0.31	0.757	-.9705291	1.333949
age65p	.4043974	.5442859	0.74	0.457	-.6623834	1.471178
female	.3415656	.2320408	1.47	0.141	-.113226	.7963572
prot	-.1404493	.262382	-0.54	0.592	-.6547085	.3738099
south	.7191063	.2686165	2.68	0.007	.1926277	1.245585
ba	.0781092	.2543411	0.31	0.759	-.4203902	.5766086
incl	-.5180865	.5627986	-0.92	0.357	-1.621151	.5849785
incdk	-.4138144	.6570865	-0.63	0.529	-1.70168	.8740515
pid1	4.595435	.4332222	10.61	0.000	3.746335	5.444535
ideol1	3.110623	.666464	4.67	0.000	1.804378	4.416868
_cons	-7.941199	.8883959	-8.94	0.000	-9.682423	-6.199975

```

56 . margins, atmeans at(rrf=(0 1))

```

```

Adjusted predictions                                Number of obs      =      1,879
Model VCE      : Robust

```

```

Expression      : Pr(votechoice), predict()

```

```

1._at      : jardina3      =      .4270238 (mean)
              rrf          =              0
              immf_4789    =      .4223346 (mean)
              sexismf      =      .3719823 (mean)
              econ1        =      .4743608 (mean)
              age3044      =      .1959655 (mean)
              age4564      =      .4108636 (mean)
              age65p       =      .2584021 (mean)
              female       =      .5317754 (mean)
              prot         =      .3606696 (mean)
              south        =      .2612745 (mean)
              ba           =      .4006297 (mean)
              incl         =      .6104942 (mean)
              incdk        =      .0277694 (mean)

```

```

                pidl          =    .5391016 (mean)
                ideoll         =    .5575229 (mean)

2._at          :  jardimna3    =    .4270238 (mean)
                rrf            =          1
                immf_4789      =    .4223346 (mean)
                sexismf        =    .3719823 (mean)
                econl          =    .4743608 (mean)
                age3044        =    .1959655 (mean)
                age4564        =    .4108636 (mean)
                age65p         =    .2584021 (mean)
                female         =    .5317754 (mean)
                prot           =    .3606696 (mean)
                south          =    .2612745 (mean)
                ba              =    .4006297 (mean)
                incl           =    .6104942 (mean)
                incdk          =    .0277694 (mean)
                pidl           =    .5391016 (mean)
                ideoll         =    .5575229 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.2436853	.0783146	3.11	0.002	.0901915	.397179
2	.8412204	.0392228	21.45	0.000	.7643451	.9180958

```

57 . di 84.1-24.4 // 59.7
59.7

```

```

58 . margins, atmeans at(immf_4789=(0 1))

```

```

Adjusted predictions          Number of obs      =      1,879
Model VCE      : Robust

```

```

Expression      : Pr(votechoice), predict()

```

```

1._at          :  jardimna3    =    .4270238 (mean)
                rrf            =    .5749568 (mean)
                immf_4789      =          0
                sexismf        =    .3719823 (mean)
                econl          =    .4743608 (mean)
                age3044        =    .1959655 (mean)
                age4564        =    .4108636 (mean)
                age65p         =    .2584021 (mean)
                female         =    .5317754 (mean)
                prot           =    .3606696 (mean)
                south          =    .2612745 (mean)
                ba              =    .4006297 (mean)
                incl           =    .6104942 (mean)
                incdk          =    .0277694 (mean)
                pidl           =    .5391016 (mean)
                ideoll         =    .5575229 (mean)

```

```

2._at      : jardina3      = .4270238 (mean)
             rrf            = .5749568 (mean)
             immf_4789      = 1
             sexismf        = .3719823 (mean)
             econ1          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)
             pid1           = .5391016 (mean)
             ideoll         = .5575229 (mean)

```

	Delta-method				
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]
_at					
1	.1301051	.0353795	3.68	0.000	.0607626 .1994476
2	.9765436	.0110091	88.70	0.000	.954966 .9981211

```

59 . di 97.6-13.0 // 84.6
    84.6

```

```

60 .
61 . logit votechoice rrf immf_4789 if white==1 [pw=wgt]

```

```

Iteration 0:  log pseudolikelihood = -1233.4658
Iteration 1:  log pseudolikelihood = -627.78656
Iteration 2:  log pseudolikelihood = -624.23011
Iteration 3:  log pseudolikelihood = -624.22014
Iteration 4:  log pseudolikelihood = -624.22014

```

```

Logistic regression                                Number of obs      =      1,943
                                                    Wald chi2(2)       =      359.77
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -624.22014                Pseudo R2         =      0.4939

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rrf	3.857071	.3687433	10.46	0.000	3.134348	4.579795
immf_4789	6.772212	.4892839	13.84	0.000	5.813233	7.731191
_cons	-4.584009	.2513258	-18.24	0.000	-5.076599	-4.091419

```

62 . margins, atmeans at(rrf=(0 1))

```

```

Adjusted predictions                                Number of obs      =      1,943
Model VCE      : Robust

```

```

Expression    : Pr(votechoice), predict()

```

```

1._at      : rrf            = 0
             immf_4789      = .4222338 (mean)

2._at      : rrf            = 1
             immf_4789      = .4222338 (mean)

```

	Delta-method				[95% Conf. Interval]	
	Margin	Std. Err.	z	P> z		
_at						
1	.1512857	.029136	5.19	0.000	.0941802	.2083913
2	.8940239	.0171257	52.20	0.000	.8604582	.9275896

```
63 . di 89.4-15.1 // 74.3
74.3
```

```
64 . margins, atmeans at(immf_4789=(0 1))
```

```
Adjusted predictions      Number of obs      =      1,943
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at      : rrf      =      .5755283 (mean)
             immf_4789      =      0
```

```
2._at      : rrf      =      .5755283 (mean)
             immf_4789      =      1
```

	Delta-method				[95% Conf. Interval]	
	Margin	Std. Err.	z	P> z		
_at						
1	.0859472	.0144523	5.95	0.000	.0576212	.1142732
2	.9879677	.0039361	251.00	0.000	.9802531	.9956824

```
65 . di 98.8-8.6 // 90.2
90.2
```

```
66 .
67 . logit votechoice rrf immf_5789 if white==1 [pw=wgt]
```

```
Iteration 0:  log pseudolikelihood = -1233.4658
Iteration 1:  log pseudolikelihood = -727.90318
Iteration 2:  log pseudolikelihood = -722.77675
Iteration 3:  log pseudolikelihood = -722.67854
Iteration 4:  log pseudolikelihood = -722.6785
```

```
Logistic regression      Number of obs      =      1,943
                          Wald chi2(2)      =      329.11
                          Prob > chi2      =      0.0000
Log pseudolikelihood = -722.6785      Pseudo R2      =      0.4141
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
votechoice						
rrf	4.663314	.3451569	13.51	0.000	3.986819	5.339809
immf_5789	4.658715	.4938052	9.43	0.000	3.690874	5.626555
_cons	-4.407628	.2567745	-17.17	0.000	-4.910897	-3.904359

```
68 . margins, atmeans at(rrf=(0 1))
```

```
Adjusted predictions                                Number of obs      =       1,943
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at          : rrf              =           0
                  immf_5789        =   .4606657 (mean)
```

```
2._at          : rrf              =           1
                  immf_5789        =   .4606657 (mean)
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.0943606	.0177781	5.31	0.000	.0595162	.129205
2	.9169642	.0129764	70.66	0.000	.891531	.9423974

```
69 . di 91.7-9.4 // 82.3
82.3
```

```
70 . margins, atmeans at(immf_5789=(0 1))
```

```
Adjusted predictions                                Number of obs      =       1,943
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at          : rrf              =   .5755283 (mean)
                  immf_5789        =           0
```

```
2._at          : rrf              =   .5755283 (mean)
                  immf_5789        =           1
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.1513877	.0272975	5.55	0.000	.0978856	.2048898
2	.9495476	.0145026	65.47	0.000	.921123	.9779723

```
71 . di 94.9-15.1 // 79.8
79.8
```

```
72 .
```

```
73 . *** Nine-item xenophobia predictor without the border wall item
```

```
74 .
```

```
75 . egen immf_no4=rowmean(antiimm1 antiimm2 antiimm3          antiimm5 antiimm6 antiimm7 antiimm8 ant
> mm10)
(7 missing values generated)
```

```
76 . logit votechoice jardina3 rrf immf_no4 sexismf econ1 age3044 age4564 age65p female prot south ba
> pid1 ideoll1 if white==1 [pw=wgt]
```

```
Iteration 0:    log pseudolikelihood = -1196.262
Iteration 1:    log pseudolikelihood = -381.75318
Iteration 2:    log pseudolikelihood = -357.11177
Iteration 3:    log pseudolikelihood = -354.43177
Iteration 4:    log pseudolikelihood = -354.42677
Iteration 5:    log pseudolikelihood = -354.42677
```

Logistic regression	Number of obs	=	1,879
	Wald chi2(16)	=	332.02
	Prob > chi2	=	0.0000
Log pseudolikelihood = -354.42677	Pseudo R2	=	0.7037

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-1.024769	.5311783	-1.93	0.054	-2.065859	.0163215
rrf	2.983797	.6721202	4.44	0.000	1.666466	4.301129
immf_no4	4.64906	.9101557	5.11	0.000	2.865187	6.432932
sexismf	3.37733	.864502	3.91	0.000	1.682937	5.071723
econl	-1.539721	.6065758	-2.54	0.011	-2.728588	-.3508546
age3044	.3676652	.5444381	0.68	0.499	-.6994139	1.434744
age4564	.2219228	.5626229	0.39	0.693	-.8807978	1.324643
age65p	.4807561	.5325187	0.90	0.367	-.5629614	1.524474
female	.2955246	.2261723	1.31	0.191	-.1477649	.7388141
prot	-.1450963	.2537362	-0.57	0.567	-.6424102	.3522175
south	.7456667	.2716003	2.75	0.006	.21334	1.277993
ba	-.1608677	.2449896	-0.66	0.511	-.6410385	.3193032
incl	-.4835785	.5324389	-0.91	0.364	-1.52714	.5599827
incdk	-.2927694	.6242417	-0.47	0.639	-1.516261	.9307219
pidl	4.622367	.4200534	11.00	0.000	3.799077	5.445656
ideoll	2.979511	.6248297	4.77	0.000	1.754868	4.204155
_cons	-7.914088	.8271608	-9.57	0.000	-9.535293	-6.292882

```
77 . margins, atmeans at(rrf=(0 1))
```

Adjusted predictions	Number of obs	=	1,879
Model VCE : Robust			

Expression : `Pr(votechoice), predict()`

```
1._at      : jardina3      =      .4270238 (mean)
            rrf            =      0
            immf_no4       =      .4660991 (mean)
            sexismf        =      .3719823 (mean)
            econ1          =      .4743608 (mean)
            age3044        =      .1959655 (mean)
            age4564        =      .4108636 (mean)
            age65p         =      .2584021 (mean)
            female         =      .5317754 (mean)
            prot           =      .3606696 (mean)
            south          =      .2612745 (mean)
            ba             =      .4006297 (mean)
            incl           =      .6104942 (mean)
            incdk          =      .0277694 (mean)
            pid1           =      .5391016 (mean)
            ideol1         =      .5575229 (mean)
```

```

2._at      : jardina3      = .4270238 (mean)
             rrf            = 1
             immf_no4       = .4660991 (mean)
             sexismf        = .3719823 (mean)
             econ1          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)
             pid1           = .5391016 (mean)
             ideoll         = .5575229 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
at						
1	.2164511	.0720295	3.01	0.003	.0752759	.3576263
2	.8451855	.0378674	22.32	0.000	.7709668	.9194041

```

78 . di 84.5-21.6 // 62.9
    62.9

```

```

79 . margins, atmeans at(immf_no4=(0 1))

```

```

Adjusted predictions      Number of obs      =      1,879
Model VCE      : Robust

```

```

Expression      : Pr(votechoice), predict()

```

```

1._at      : jardina3      = .4270238 (mean)
             rrf            = .5749568 (mean)
             immf_no4       = 0
             sexismf        = .3719823 (mean)
             econ1          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)
             pid1           = .5391016 (mean)
             ideoll         = .5575229 (mean)

```

```

2._at      : jardina3      = .4270238 (mean)
             rrf            = .5749568 (mean)
             immf_no4       = 1
             sexismf        = .3719823 (mean)
             econ1          = .4743608 (mean)
             age3044        = .1959655 (mean)
             age4564        = .4108636 (mean)
             age65p         = .2584021 (mean)
             female         = .5317754 (mean)
             prot           = .3606696 (mean)
             south          = .2612745 (mean)
             ba             = .4006297 (mean)
             incl           = .6104942 (mean)
             incdk          = .0277694 (mean)

```

```
pid1          = .5391016 (mean)
ideoll        = .5575229 (mean)
```

	Delta-method				[95% Conf. Interval]	
	Margin	Std. Err.	z	P> z		
_at						
1	.1495879	.0534401	2.80	0.005	.0448472	.2543286
2	.9483984	.0254797	37.22	0.000	.898459	.9983378

```
80 . di 94.8-15.0 // 79.8
79.8
```

```
81 .
82 . *** Using indicators
83 .
84 . tab1 rrf immf_4789
```

-> tabulation of rrf

rrf	Freq.	Percent	Cum.
0	187	5.14	5.14
.0625	124	3.41	8.55
.0833333	1	0.03	8.57
.125	110	3.02	11.60
.1666667	1	0.03	11.62
.1875	119	3.27	14.89
.25	190	5.22	20.12
.3125	164	4.51	24.62
.3333333	2	0.05	24.68
.375	196	5.39	30.06
.4375	227	6.24	36.30
.5	422	11.60	47.90
.5625	257	7.06	54.96
.5833333	3	0.08	55.04
.625	273	7.50	62.54
.6875	254	6.98	69.52
.75	264	7.25	76.78
.8125	172	4.73	81.51
.8333333	1	0.03	81.53
.875	212	5.83	87.36
.9166667	1	0.03	87.39
.9375	167	4.59	91.98
1	292	8.02	100.00
Total	3,639	100.00	

-> tabulation of immf_4789

immf_4789	Freq.	Percent	Cum.
0	495	11.62	11.62
.0416667	17	0.40	12.02
.0625	241	5.66	17.67
.0833333	1	0.02	17.70
.1041667	21	0.49	18.19
.1111111	1	0.02	18.21
.125	190	4.46	22.67
.1458333	7	0.16	22.84
.1666667	81	1.90	24.74
.1875	226	5.30	30.04
.2083333	5	0.12	30.16
.2083333	3	0.07	30.23
.2222222	1	0.02	30.25

.2291667	53	1.24	31.49
.25	202	4.74	36.24
.2708333	7	0.16	36.40
.2708333	13	0.31	36.70
.2916667	41	0.96	37.67
.2916667	1	0.02	37.69
.3055556	1	0.02	37.71
.3125	166	3.90	41.61
.3333333	15	0.35	41.96
.3333333	22	0.52	42.48
.3541667	40	0.94	43.42
.3541667	9	0.21	43.63
.375	213	5.00	48.63
.3958333	26	0.61	49.24
.3958333	8	0.19	49.43
.4166667	48	1.13	50.55
.4166667	12	0.28	50.83
.4375	200	4.69	55.53
.4444444	1	0.02	55.55
.4583333	37	0.87	56.42
.4583333	13	0.31	56.72
.4791667	27	0.63	57.36
.4791667	11	0.26	57.62
.5	440	10.33	67.94
.5208333	53	1.24	69.19
.5277778	1	0.02	69.21
.5416667	25	0.59	69.80
.5555556	1	0.02	69.82
.5625	181	4.25	74.07
.5833333	72	1.69	75.76
.6041667	13	0.31	76.06
.6111111	1	0.02	76.09
.625	193	4.53	80.61
.6458333	60	1.41	82.02
.6666666	1	0.02	82.05
.6666667	26	0.61	82.66
.6875	145	3.40	86.06
.7083333	42	0.99	87.05
.7291667	4	0.09	87.14
.75	102	2.39	89.53
.7708333	34	0.80	90.33
.7777778	1	0.02	90.35
.7916666	1	0.02	90.38
.7916667	2	0.05	90.42
.8125	81	1.90	92.33
.8333333	65	1.53	93.85
.8541667	1	0.02	93.87
.875	53	1.24	95.12
.8958333	6	0.14	95.26
.9166667	1	0.02	95.28
.9375	32	0.75	96.03
.9583333	3	0.07	96.10
1	166	3.90	100.00
Total	4,261	100.00	

```

85 .
86 . gen rr16 = rrf * 16          // Get rr onto a scale in which all values are integers
    (631 missing values generated)

87 . gen rr16round = round(rr16)
    (631 missing values generated)

88 .
89 . gen imm24 = immf_4789 * 24 // Get imm onto a scale in which all values are integers
    (9 missing values generated)

90 . gen imm24round = round(imm24)
    (9 missing values generated)

91 .
92 . tab1 rr16 imm24

```

-> tabulation of rr16

rr16	Freq.	Percent	Cum.
0	187	5.14	5.14
1	124	3.41	8.55
1.333333	1	0.03	8.57
2	110	3.02	11.60
2.666667	1	0.03	11.62
3	119	3.27	14.89
4	190	5.22	20.12
5	164	4.51	24.62
5.333333	2	0.05	24.68
6	196	5.39	30.06
7	227	6.24	36.30
8	422	11.60	47.90
9	257	7.06	54.96
9.333333	3	0.08	55.04
10	273	7.50	62.54
11	254	6.98	69.52
12	264	7.25	76.78
13	172	4.73	81.51
13.33333	1	0.03	81.53
14	212	5.83	87.36
14.66667	1	0.03	87.39
15	167	4.59	91.98
16	292	8.02	100.00
Total	3,639	100.00	

-> tabulation of imm24

imm24	Freq.	Percent	Cum.
0	495	11.62	11.62
1	17	0.40	12.02
1.5	241	5.66	17.67
2	1	0.02	17.70
2.5	21	0.49	18.19
2.666667	1	0.02	18.21
3	190	4.46	22.67
3.5	7	0.16	22.84
4	81	1.90	24.74
4.5	226	5.30	30.04
5	8	0.19	30.23
5.333333	1	0.02	30.25
5.5	53	1.24	31.49
6	202	4.74	36.24
6.5	7	0.16	36.40
6.5	13	0.31	36.70

7	41	0.96	37.67
7	1	0.02	37.69
7.333333	1	0.02	37.71
7.5	166	3.90	41.61
8	15	0.35	41.96
8	22	0.52	42.48
8.5	49	1.15	43.63
9	213	5.00	48.63
9.5	34	0.80	49.43
10	60	1.41	50.83
10.5	200	4.69	55.53
10.66667	1	0.02	55.55
11	50	1.17	56.72
11.5	38	0.89	57.62
12	440	10.33	67.94
12.5	53	1.24	69.19
12.66667	1	0.02	69.21
13	25	0.59	69.80
13.33333	1	0.02	69.82
13.5	181	4.25	74.07
14	72	1.69	75.76
14.5	13	0.31	76.06
14.66667	1	0.02	76.09
15	193	4.53	80.61
15.5	60	1.41	82.02
16	1	0.02	82.05
16	26	0.61	82.66
16.5	145	3.40	86.06
17	42	0.99	87.05
17.5	4	0.09	87.14
18	102	2.39	89.53
18.5	34	0.80	90.33
18.66667	1	0.02	90.35
19	3	0.07	90.42
19.5	81	1.90	92.33
20	65	1.53	93.85
20.5	1	0.02	93.87
21	53	1.24	95.12
21.5	6	0.14	95.26
22	1	0.02	95.28
22.5	32	0.75	96.03
23	3	0.07	96.10
24	166	3.90	100.00
Total	4,261	100.00	

93 . pwcorr rr16 rr16round imm24 imm24round

	rr16	rr16ro~d	imm24	imm24r~d
rr16	1.0000			
rr16round	1.0000	1.0000		
imm24	0.5639	0.5638	1.0000	
imm24round	0.5647	0.5647	0.9993	1.0000

```
94 . tab1 rr16 rr16round imm24 imm24round
```

```
-> tabulation of rr16
```

rr16	Freq.	Percent	Cum.
0	187	5.14	5.14
1	124	3.41	8.55
1.333333	1	0.03	8.57
2	110	3.02	11.60
2.666667	1	0.03	11.62
3	119	3.27	14.89
4	190	5.22	20.12
5	164	4.51	24.62
5.333333	2	0.05	24.68
6	196	5.39	30.06
7	227	6.24	36.30
8	422	11.60	47.90
9	257	7.06	54.96
9.333333	3	0.08	55.04
10	273	7.50	62.54
11	254	6.98	69.52
12	264	7.25	76.78
13	172	4.73	81.51
13.33333	1	0.03	81.53
14	212	5.83	87.36
14.66667	1	0.03	87.39
15	167	4.59	91.98
16	292	8.02	100.00
Total	3,639	100.00	

```
-> tabulation of rr16round
```

rr16round	Freq.	Percent	Cum.
0	187	5.14	5.14
1	125	3.44	8.57
2	110	3.02	11.60
3	120	3.30	14.89
4	190	5.22	20.12
5	166	4.56	24.68
6	196	5.39	30.06
7	227	6.24	36.30
8	422	11.60	47.90
9	260	7.14	55.04
10	273	7.50	62.54
11	254	6.98	69.52
12	264	7.25	76.78
13	173	4.75	81.53
14	212	5.83	87.36
15	168	4.62	91.98
16	292	8.02	100.00
Total	3,639	100.00	

```
-> tabulation of imm24
```

imm24	Freq.	Percent	Cum.
0	495	11.62	11.62
1	17	0.40	12.02
1.5	241	5.66	17.67
2	1	0.02	17.70
2.5	21	0.49	18.19
2.666667	1	0.02	18.21
3	190	4.46	22.67
3.5	7	0.16	22.84
4	81	1.90	24.74
4.5	226	5.30	30.04
5	8	0.19	30.23
5.333333	1	0.02	30.25
5.5	53	1.24	31.49
6	202	4.74	36.24
6.5	7	0.16	36.40
6.5	13	0.31	36.70
7	41	0.96	37.67
7	1	0.02	37.69
7.333333	1	0.02	37.71
7.5	166	3.90	41.61
8	15	0.35	41.96
8	22	0.52	42.48
8.5	49	1.15	43.63
9	213	5.00	48.63
9.5	34	0.80	49.43
10	60	1.41	50.83
10.5	200	4.69	55.53
10.66667	1	0.02	55.55
11	50	1.17	56.72
11.5	38	0.89	57.62
12	440	10.33	67.94
12.5	53	1.24	69.19
12.66667	1	0.02	69.21
13	25	0.59	69.80
13.33333	1	0.02	69.82
13.5	181	4.25	74.07
14	72	1.69	75.76
14.5	13	0.31	76.06
14.66667	1	0.02	76.09
15	193	4.53	80.61
15.5	60	1.41	82.02
16	1	0.02	82.05
16	26	0.61	82.66
16.5	145	3.40	86.06
17	42	0.99	87.05
17.5	4	0.09	87.14
18	102	2.39	89.53
18.5	34	0.80	90.33
18.66667	1	0.02	90.35
19	3	0.07	90.42
19.5	81	1.90	92.33
20	65	1.53	93.85
20.5	1	0.02	93.87
21	53	1.24	95.12
21.5	6	0.14	95.26
22	1	0.02	95.28
22.5	32	0.75	96.03
23	3	0.07	96.10
24	166	3.90	100.00
Total	4,261	100.00	

-> tabulation of imm24round

imm24round	Freq.	Percent	Cum.
0	495	11.62	11.62
1	17	0.40	12.02
2	242	5.68	17.70
3	212	4.98	22.67
4	88	2.07	24.74
5	235	5.52	30.25
6	262	6.15	36.40
7	56	1.31	37.71
8	203	4.76	42.48
9	262	6.15	48.63
10	94	2.21	50.83
11	251	5.89	56.72
12	478	11.22	67.94
13	80	1.88	69.82
14	253	5.94	75.76
15	207	4.86	80.61
16	87	2.04	82.66
17	187	4.39	87.05
18	106	2.49	89.53
19	38	0.89	90.42
20	146	3.43	93.85
21	54	1.27	95.12
22	7	0.16	95.28
23	35	0.82	96.10
24	166	3.90	100.00
Total	4,261	100.00	

```

95 .
96 . logit votechoice i.rrl6round i.imm24round if white==1 [pw=wgt] // Omitted low rr values

```

```

note: 1.rrl6round != 0 predicts failure perfectly
      1.rrl6round dropped and 64 obs not used

note: 2.rrl6round != 0 predicts failure perfectly
      2.rrl6round dropped and 55 obs not used

note: 21.imm24round != 0 predicts success perfectly
      21.imm24round dropped and 34 obs not used

```

```

Iteration 0:  log pseudolikelihood = -1117.2071
Iteration 1:  log pseudolikelihood = -587.81732
Iteration 2:  log pseudolikelihood = -571.55216
Iteration 3:  log pseudolikelihood = -570.42417
Iteration 4:  log pseudolikelihood = -570.41188
Iteration 5:  log pseudolikelihood = -570.41187

```

```

Logistic regression                                Number of obs      =      1,790
                                                    Wald chi2(37)      =      538.43
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -570.41187                Pseudo R2          =      0.4894

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rrl6round						
1	0	(empty)				
2	0	(empty)				
3	1.148106	.8463152	1.36	0.175	-.5106409	2.806854
4	.7859543	.9036483	0.87	0.384	-.9851639	2.557073
5	1.706728	.7983593	2.14	0.033	.1419724	3.271483
6	1.127425	.8234377	1.37	0.171	-.4864833	2.741333
7	2.05236	.8007948	2.56	0.010	.4828313	3.621889
8	1.888056	.7905502	2.39	0.017	.3386059	3.437506

9	1.669032	.8167431	2.04	0.041	.0682453	3.269819
10	2.377132	.8023585	2.96	0.003	.8045384	3.949726
11	2.570551	.7930683	3.24	0.001	1.016166	4.124936
12	3.137101	.8257223	3.80	0.000	1.518715	4.755487
13	3.580326	.8458196	4.23	0.000	1.922549	5.238102
14	3.219937	.8057853	4.00	0.000	1.640627	4.799247
15	3.947142	.8605893	4.59	0.000	2.260418	5.633866
16	3.251135	.9008838	3.61	0.000	1.485435	5.016835
imm24round						
1	2.093493	.9892854	2.12	0.034	.1545291	4.032457
2	-1.535747	.8644035	-1.78	0.076	-3.229946	.1584528
3	-.7601596	.8004143	-0.95	0.342	-2.328943	.8086235
4	1.73367	.8242706	2.10	0.035	.1181292	3.34921
5	.8505526	.6885616	1.24	0.217	-.4990033	2.200109
6	1.452888	.6465774	2.25	0.025	.1856196	2.720157
7	2.378847	.7218854	3.30	0.001	.963977	3.793716
8	2.719737	.668056	4.07	0.000	1.410371	4.029103
9	2.317062	.662935	3.50	0.000	1.017734	3.616391
10	3.858543	.7640667	5.05	0.000	2.361	5.356086
11	3.93805	.6826028	5.77	0.000	2.600173	5.275927
12	3.656291	.6528768	5.60	0.000	2.376676	4.935907
13	3.189651	.7621122	4.19	0.000	1.695939	4.683364
14	4.279261	.6869152	6.23	0.000	2.932932	5.62559
15	3.411211	.6890118	4.95	0.000	2.060773	4.761649
16	5.477321	.9405011	5.82	0.000	3.633972	7.320669
17	4.325964	.7878484	5.49	0.000	2.78181	5.870119
18	4.158137	.7968636	5.22	0.000	2.596313	5.719961
19	4.477188	1.344479	3.33	0.001	1.842058	7.112317
20	5.332588	.9899514	5.39	0.000	3.392319	7.272857
21	0	(empty)				
22	2.908573	1.103557	2.64	0.008	.7456401	5.071506
23	5.070016	1.192806	4.25	0.000	2.732159	7.407872
24	4.674652	1.076562	4.34	0.000	2.56463	6.784674
_cons	-4.559434	.644869	-7.07	0.000	-5.823354	-3.295514

```

97 .
98 . recode rr16round (0/1=1), gen(rr16round01)
    (187 differences between rr16round and rr16round01)

99 . logit votechoice i.rr16round01 i.imm24round if white==1 [pw=wgt] // Still omitted low rr values

note: 2.rr16round01 != 0 predicts failure perfectly
      2.rr16round01 dropped and 55 obs not used

note: 21.imm24round != 0 predicts success perfectly
      21.imm24round dropped and 34 obs not used

Iteration 0:  log pseudolikelihood = -1171.2308
Iteration 1:  log pseudolikelihood = -592.27397
Iteration 2:  log pseudolikelihood = -573.00781
Iteration 3:  log pseudolikelihood = -571.29948
Iteration 4:  log pseudolikelihood = -571.26167
Iteration 5:  log pseudolikelihood = -571.26153
Iteration 6:  log pseudolikelihood = -571.26153

Logistic regression                                Number of obs      =       1,854
                                                    Wald chi2(37)      =       550.48
                                                    Prob > chi2        =       0.0000
Log pseudolikelihood = -571.26153                Pseudo R2          =       0.5123

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rr16round01						
2	0	(empty)				
3	1.712411	.861061	1.99	0.047	.0247624	3.40006
4	1.368338	.8804087	1.55	0.120	-.3572316	3.093907
5	2.274473	.806527	2.82	0.005	.6937093	3.855237
6	1.697467	.8264396	2.05	0.040	.0776754	3.317259
7	2.621206	.8077373	3.25	0.001	1.03807	4.204342
8	2.456935	.7957285	3.09	0.002	.8973356	4.016534
9	2.237183	.8234425	2.72	0.007	.6232654	3.851101
10	2.946155	.8076286	3.65	0.000	1.363232	4.529078
11	3.145708	.7953887	3.95	0.000	1.586775	4.704641
12	3.706486	.8281898	4.48	0.000	2.083264	5.329708
13	4.14901	.8520835	4.87	0.000	2.478957	5.819063
14	3.788975	.8134846	4.66	0.000	2.194574	5.383375
15	4.516336	.8661205	5.21	0.000	2.818771	6.213901
16	3.818444	.9068652	4.21	0.000	2.041021	5.595867
imm24round						
1	1.692542	.9225115	1.83	0.067	-.1155473	3.500631
2	-1.593588	.8486793	-1.88	0.060	-3.256969	.0697928
3	-.810133	.7864117	-1.03	0.303	-2.351472	.7312056
4	1.681642	.8064299	2.09	0.037	.1010681	3.262215
5	.7953459	.6658293	1.19	0.232	-.5096556	2.100347
6	1.397243	.6226135	2.24	0.025	.176943	2.617543
7	2.318103	.6966136	3.33	0.001	.9527653	3.68344
8	2.665687	.6448602	4.13	0.000	1.401785	3.92959
9	2.263754	.6402835	3.54	0.000	1.008821	3.518686
10	3.805762	.7447123	5.11	0.000	2.346153	5.265371
11	3.885005	.6605473	5.88	0.000	2.590356	5.179654
12	3.607121	.632061	5.71	0.000	2.368304	4.845938
13	3.136539	.742605	4.22	0.000	1.681059	4.592018
14	4.226142	.6649474	6.36	0.000	2.922869	5.529415
15	3.358344	.6675068	5.03	0.000	2.050055	4.666633
16	5.421473	.9224609	5.88	0.000	3.613483	7.229464
17	4.272294	.7679182	5.56	0.000	2.767202	5.777386
18	4.12647	.7975695	5.17	0.000	2.563263	5.689678
19	4.42158	1.330961	3.32	0.001	1.812944	7.030216
20	5.279992	.9751006	5.41	0.000	3.36883	7.191154
21	0	(empty)				
22	2.856046	1.090288	2.62	0.009	.7191217	4.992971
23	5.017335	1.18051	4.25	0.000	2.703578	7.331092
24	4.622574	1.06286	4.35	0.000	2.539406	6.705741
_cons	-5.075915	.7265587	-6.99	0.000	-6.499943	-3.651886

100 .

101 . recode rr16round (0/2=2), gen(rr16round012)
(312 differences between rr16round and rr16round012)

102 . logit votechoice i.rr16round012 i.imm24round if white==1 [pw=wgt] // Now work on omitted imm val

note: 21.imm24round != 0 predicts success perfectly
21.imm24round dropped and 34 obs not used

Iteration 0: log pseudolikelihood = -1214.7279
Iteration 1: log pseudolikelihood = -600.43606
Iteration 2: log pseudolikelihood = -575.45523
Iteration 3: log pseudolikelihood = -572.70479
Iteration 4: log pseudolikelihood = -572.61813
Iteration 5: log pseudolikelihood = -572.61802
Iteration 6: log pseudolikelihood = -572.61802

Logistic regression	Number of obs	=	1,909
	Wald chi2(37)	=	529.08
	Prob > chi2	=	0.0000
Log pseudolikelihood = -572.61802	Pseudo R2	=	0.5286

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rr16round012						
3	2.604827	.9046201	2.88	0.004	.8318039	4.37785
4	2.257383	.8958639	2.52	0.012	.5015219	4.013244
5	3.175406	.8389099	3.79	0.000	1.531172	4.819639
6	2.60414	.8459811	3.08	0.002	.9460473	4.262232
7	3.524296	.8347702	4.22	0.000	1.888176	5.160415
8	3.362895	.8181348	4.11	0.000	1.75938	4.966409
9	3.142946	.8486136	3.70	0.000	1.479694	4.806198
10	3.852461	.8290114	4.65	0.000	2.227629	5.477294
11	4.04677	.8204185	4.93	0.000	2.438779	5.65476
12	4.612304	.8537886	5.40	0.000	2.938909	6.285699
13	5.051906	.8790931	5.75	0.000	3.328915	6.774897
14	4.690373	.841416	5.57	0.000	3.041228	6.339518
15	5.42096	.8879005	6.11	0.000	3.680707	7.161213
16	4.723709	.9346336	5.05	0.000	2.891861	6.555557
imm24round						
1	1.845332	.9455569	1.95	0.051	-.0079259	3.698589
2	-1.652681	.8434373	-1.96	0.050	-3.305788	.0004255
3	-.8705176	.7787814	-1.12	0.264	-2.396901	.655866
4	1.602847	.7901433	2.03	0.043	.054195	3.1515
5	.7208154	.6496884	1.11	0.267	-.5525504	1.994181
6	1.323846	.606105	2.18	0.029	.1359019	2.51179
7	2.242081	.680847	3.29	0.001	.9076452	3.576516
8	2.59161	.6271373	4.13	0.000	1.362443	3.820776
9	2.185853	.6216365	3.52	0.000	.967468	3.404238
10	3.727573	.7286915	5.12	0.000	2.299364	5.155782
11	3.798631	.6388648	5.95	0.000	2.546479	5.050783
12	3.534388	.6159669	5.74	0.000	2.327115	4.741661
13	3.059443	.7268248	4.21	0.000	1.634892	4.483993
14	4.106065	.6316022	6.50	0.000	2.868148	5.343983
15	3.277854	.648719	5.05	0.000	2.006388	4.54932
16	5.346699	.9106661	5.87	0.000	3.561826	7.131572
17	4.195929	.7530524	5.57	0.000	2.719973	5.671884
18	4.074266	.80727	5.05	0.000	2.492046	5.656487
19	4.347526	1.32387	3.28	0.001	1.752787	6.942264
20	5.202479	.9633545	5.40	0.000	3.314339	7.090619
21	0	(empty)				
22	2.778792	1.080074	2.57	0.010	.6618858	4.895699
23	4.94037	1.170915	4.22	0.000	2.64542	7.23532
24	4.545382	1.051741	4.32	0.000	2.484008	6.606755
_cons	-5.90292	.8515452	-6.93	0.000	-7.571918	-4.233922

```

104 . recode imm24round (21/22=21) (23=22) (24=23), gen(imm24round23)
    (208 differences between imm24round and imm24round23)

105 . logit votechoice i.rr16round012 i.imm24round23 if white==1 [pw=wt]

```

```

Iteration 0:  log pseudolikelihood = -1233.4658
Iteration 1:  log pseudolikelihood = -603.54559
Iteration 2:  log pseudolikelihood = -577.58061
Iteration 3:  log pseudolikelihood = -574.83456
Iteration 4:  log pseudolikelihood = -574.74263
Iteration 5:  log pseudolikelihood = -574.74253
Iteration 6:  log pseudolikelihood = -574.74253

```

```

Logistic regression                                Number of obs      =      1,943
                                                    Wald chi2(37)      =      541.30
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -574.74253                Pseudo R2         =      0.5340

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rr16round012						
3	2.604378	.9046193	2.88	0.004	.8313573	4.3774
4	2.256984	.8959042	2.52	0.012	.5010441	4.012924
5	3.175412	.838916	3.79	0.000	1.531167	4.819657
6	2.604004	.8459606	3.08	0.002	.9459522	4.262057
7	3.524073	.8347689	4.22	0.000	1.887956	5.16019
8	3.35027	.8186718	4.09	0.000	1.745703	4.954837
9	3.160911	.8477246	3.73	0.000	1.499401	4.82242
10	3.853772	.8289403	4.65	0.000	2.229079	5.478466
11	4.052829	.8202775	4.94	0.000	2.445115	5.660544
12	4.606263	.8536467	5.40	0.000	2.933146	6.27938
13	5.041472	.8789195	5.74	0.000	3.318822	6.764123
14	4.677134	.8411167	5.56	0.000	3.028575	6.325692
15	5.427273	.8875433	6.11	0.000	3.68772	7.166826
16	4.743002	.9337388	5.08	0.000	2.912908	6.573096
imm24round23						
1	1.843269	.9443612	1.95	0.051	-.0076449	3.694183
2	-1.654169	.8433099	-1.96	0.050	-3.307026	-.0013118
3	-.8702204	.778135	-1.12	0.263	-2.395337	.6548963
4	1.601924	.789719	2.03	0.043	.0541035	3.149745
5	.7185709	.648778	1.11	0.268	-.5530107	1.990153
6	1.323237	.605472	2.19	0.029	.1365335	2.50994
7	2.243282	.6802215	3.30	0.001	.9100724	3.576492
8	2.591746	.6263509	4.14	0.000	1.36412	3.819371
9	2.185729	.6209344	3.52	0.000	.9687194	3.402738
10	3.723341	.7276617	5.12	0.000	2.297151	5.149532
11	3.796183	.6382533	5.95	0.000	2.54523	5.047137
12	3.533768	.6152108	5.74	0.000	2.327977	4.739559
13	3.056291	.7257183	4.21	0.000	1.63391	4.478673
14	4.104843	.6309896	6.51	0.000	2.868126	5.34156
15	3.276842	.6481712	5.06	0.000	2.00645	4.547234
16	5.345937	.9100587	5.87	0.000	3.562255	7.129619
17	4.195866	.7530209	5.57	0.000	2.719972	5.67176
18	4.070492	.8063177	5.05	0.000	2.490139	5.650846
19	4.347483	1.32397	3.28	0.001	1.75255	6.942416
20	5.197986	.9634702	5.40	0.000	3.30962	7.086353
21	5.306572	1.174345	4.52	0.000	3.004899	7.608245
22	4.93291	1.168934	4.22	0.000	2.641841	7.223979
23	4.535563	1.050782	4.32	0.000	2.476067	6.595058
_cons	-5.901813	.8511248	-6.93	0.000	-7.569987	-4.233639

```

106 .
107 . tab1 rr16round012 imm24round23

```

-> tabulation of rr16round012

RECODE of rr16round	Freq.	Percent	Cum.
2	422	11.60	11.60
3	120	3.30	14.89
4	190	5.22	20.12
5	166	4.56	24.68
6	196	5.39	30.06
7	227	6.24	36.30
8	422	11.60	47.90
9	260	7.14	55.04
10	273	7.50	62.54
11	254	6.98	69.52
12	264	7.25	76.78
13	173	4.75	81.53
14	212	5.83	87.36
15	168	4.62	91.98
16	292	8.02	100.00
Total	3,639	100.00	

-> tabulation of imm24round23

RECODE of imm24round	Freq.	Percent	Cum.
0	495	11.62	11.62
1	17	0.40	12.02
2	242	5.68	17.70
3	212	4.98	22.67
4	88	2.07	24.74
5	235	5.52	30.25
6	262	6.15	36.40
7	56	1.31	37.71
8	203	4.76	42.48
9	262	6.15	48.63
10	94	2.21	50.83
11	251	5.89	56.72
12	478	11.22	67.94
13	80	1.88	69.82
14	253	5.94	75.76
15	207	4.86	80.61
16	87	2.04	82.66
17	187	4.39	87.05
18	106	2.49	89.53
19	38	0.89	90.42
20	146	3.43	93.85
21	61	1.43	95.28
22	35	0.82	96.10
23	166	3.90	100.00
Total	4,261	100.00	

108 . logit votechoice i.rr16round012 i.imm24round23 if white==1 [pw=wgt]

Iteration 0: log pseudolikelihood = -1233.4658
 Iteration 1: log pseudolikelihood = -603.54559
 Iteration 2: log pseudolikelihood = -577.58061
 Iteration 3: log pseudolikelihood = -574.83456
 Iteration 4: log pseudolikelihood = -574.74263
 Iteration 5: log pseudolikelihood = -574.74253
 Iteration 6: log pseudolikelihood = -574.74253

Logistic regression	Number of obs	=	1,943
	Wald chi2(37)	=	541.30
	Prob > chi2	=	0.0000
Log pseudolikelihood = -574.74253	Pseudo R2	=	0.5340

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rr16round012						
3	2.604378	.9046193	2.88	0.004	.8313573	4.3774
4	2.256984	.8959042	2.52	0.012	.5010441	4.012924
5	3.175412	.838916	3.79	0.000	1.531167	4.819657
6	2.604004	.8459606	3.08	0.002	.9459522	4.262057
7	3.524073	.8347689	4.22	0.000	1.887956	5.16019
8	3.35027	.8186718	4.09	0.000	1.745703	4.954837
9	3.160911	.8477246	3.73	0.000	1.499401	4.82242
10	3.853772	.8289403	4.65	0.000	2.229079	5.478466
11	4.052829	.8202775	4.94	0.000	2.445115	5.660544
12	4.606263	.8536467	5.40	0.000	2.933146	6.27938
13	5.041472	.8789195	5.74	0.000	3.318822	6.764123
14	4.677134	.8411167	5.56	0.000	3.028575	6.325692
15	5.427273	.8875433	6.11	0.000	3.68772	7.166826
16	4.743002	.9337388	5.08	0.000	2.912908	6.573096
imm24round23						
1	1.843269	.9443612	1.95	0.051	-.0076449	3.694183
2	-1.654169	.8433099	-1.96	0.050	-3.307026	-.0013118
3	-.8702204	.778135	-1.12	0.263	-2.395337	.6548963
4	1.601924	.789719	2.03	0.043	.0541035	3.149745
5	.7185709	.648778	1.11	0.268	-.5530107	1.990153
6	1.323237	.605472	2.19	0.029	.1365335	2.50994
7	2.243282	.6802215	3.30	0.001	.9100724	3.576492
8	2.591746	.6263509	4.14	0.000	1.36412	3.819371
9	2.185729	.6209344	3.52	0.000	.9687194	3.402738
10	3.723341	.7276617	5.12	0.000	2.297151	5.149532
11	3.796183	.6382533	5.95	0.000	2.54523	5.047137
12	3.533768	.6152108	5.74	0.000	2.327977	4.739559
13	3.056291	.7257183	4.21	0.000	1.63391	4.478673
14	4.104843	.6309896	6.51	0.000	2.868126	5.34156
15	3.276842	.6481712	5.06	0.000	2.00645	4.547234
16	5.345937	.9100587	5.87	0.000	3.562255	7.129619
17	4.195866	.7530209	5.57	0.000	2.719972	5.67176
18	4.070492	.8063177	5.05	0.000	2.490139	5.650846
19	4.347483	1.32397	3.28	0.001	1.75255	6.942416
20	5.197986	.9634702	5.40	0.000	3.30962	7.086353
21	5.306572	1.174345	4.52	0.000	3.004899	7.608245
22	4.93291	1.168934	4.22	0.000	2.641841	7.223979
23	4.535563	1.050782	4.32	0.000	2.476067	6.595058
_cons	-5.901813	.8511248	-6.93	0.000	-7.569987	-4.233639

```
109 . margins, atmeans at(rr16round012=(2 16))
```

```
Adjusted predictions                                Number of obs    =      1,943  
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at           : rr16round012    =           2  
                  0.imm24ro~23    =    .0883014 (mean)  
                  1.imm24ro~23    =    .0064519 (mean)  
                  2.imm24ro~23    =    .0687469 (mean)  
                  3.imm24ro~23    =    .0464074 (mean)  
                  4.imm24ro~23    =    .0067487 (mean)  
                  5.imm24ro~23    =    .0450074 (mean)  
                  6.imm24ro~23    =    .0646337 (mean)  
                  7.imm24ro~23    =    .016623 (mean)  
                  8.imm24ro~23    =    .0459719 (mean)  
                  9.imm24ro~23    =    .0614723 (mean)  
                 10.imm24r~23     =    .0186531 (mean)  
                 11.imm24r~23     =    .0637626 (mean)  
                 12.imm24r~23     =    .0701655 (mean)  
                 13.imm24r~23     =    .0259049 (mean)  
                 14.imm24r~23     =    .0844433 (mean)  
                 15.imm24r~23     =    .0744172 (mean)  
                 16.imm24r~23     =    .0194284 (mean)  
                 17.imm24r~23     =    .0593815 (mean)  
                 18.imm24r~23     =    .0335658 (mean)  
                 19.imm24r~23     =    .0084834 (mean)  
                 20.imm24r~23     =    .0421918 (mean)  
                 21.imm24r~23     =    .0207639 (mean)  
                 22.imm24r~23     =    .0143712 (mean)  
                 23.imm24r~23     =    .0141028 (mean)
```

```
2._at           : rr16round012    =          16  
                  0.imm24ro~23    =    .0883014 (mean)  
                  1.imm24ro~23    =    .0064519 (mean)  
                  2.imm24ro~23    =    .0687469 (mean)  
                  3.imm24ro~23    =    .0464074 (mean)  
                  4.imm24ro~23    =    .0067487 (mean)  
                  5.imm24ro~23    =    .0450074 (mean)  
                  6.imm24ro~23    =    .0646337 (mean)  
                  7.imm24ro~23    =    .016623 (mean)  
                  8.imm24ro~23    =    .0459719 (mean)  
                  9.imm24ro~23    =    .0614723 (mean)  
                 10.imm24r~23     =    .0186531 (mean)  
                 11.imm24r~23     =    .0637626 (mean)  
                 12.imm24r~23     =    .0701655 (mean)  
                 13.imm24r~23     =    .0259049 (mean)  
                 14.imm24r~23     =    .0844433 (mean)  
                 15.imm24r~23     =    .0744172 (mean)  
                 16.imm24r~23     =    .0194284 (mean)  
                 17.imm24r~23     =    .0593815 (mean)  
                 18.imm24r~23     =    .0335658 (mean)  
                 19.imm24r~23     =    .0084834 (mean)  
                 20.imm24r~23     =    .0421918 (mean)  
                 21.imm24r~23     =    .0207639 (mean)  
                 22.imm24r~23     =    .0143712 (mean)  
                 23.imm24r~23     =    .0141028 (mean)
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.0321534	.0242476	1.33	0.185	-.0153711	.0796779
2	.7922344	.0809679	9.78	0.000	.6335402	.9509286

```
110 . di 79.2-3.2 // 76.0
```

```
111 . margins, atmeans at(imm24round23=(0 23))
```

Adjusted predictions	Number of obs	=	1,943
Model VCE : Robust			

Model VCE : Robust

Expression : `Pr(votechoice), predict()`

```
1._at      : 2.rr16ro~012      = .1182603 (mean)
            3.rr16ro~012      = .0316779 (mean)
            4.rr16ro~012      = .0426164 (mean)
            5.rr16ro~012      = .0453131 (mean)
            6.rr16ro~012      = .0485361 (mean)
            7.rr16ro~012      = .0501375 (mean)
            8.rr16ro~012      = .0851097 (mean)
            9.rr16ro~012      = .052753 (mean)
            10.rr16r~012      = .0871472 (mean)
            11.rr16r~012      = .0688509 (mean)
            12.rr16r~012      = .0811801 (mean)
            13.rr16r~012      = .0559266 (mean)
            14.rr16r~012      = .0795804 (mean)
            15.rr16r~012      = .0608571 (mean)
            16.rr16r~012      = .0920535 (mean)
            imm24round23      = 0
```

```
2._at      : 2.rr16ro~012      =      .1182603 (mean)
              3.rr16ro~012      =      .0316779 (mean)
              4.rr16ro~012      =      .0426164 (mean)
              5.rr16ro~012      =      .0453131 (mean)
              6.rr16ro~012      =      .0485361 (mean)
              7.rr16ro~012      =      .0501375 (mean)
              8.rr16ro~012      =      .0851097 (mean)
              9.rr16ro~012      =      .052753 (mean)
             10.rr16r~012       =      .0871472 (mean)
             11.rr16r~012       =      .0688509 (mean)
             12.rr16r~012       =      .0811801 (mean)
             13.rr16r~012       =      .0559266 (mean)
             14.rr16r~012       =      .0795804 (mean)
             15.rr16r~012       =      .0608571 (mean)
             16.rr16r~012       =      .0920535 (mean)
             imm24round23       =      23
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at_1	.0820818	.0416943	1.97	0.049	.0003625	.163801
2	.8929438	.0839591	10.64	0.000	.728387	1.057501

```
112 . di 89.3-8.2 // 81.1
      81.1
```

```
113 .
114 . logit votechoice jardina3 i.rrl6round i.imm24round sexismf econ1 age3044 age4564 age65p fem
    > uth ba incl incdk pid1 ideoll if white==1 [pw=wgt]
```

```
note: 1.rrl6round != 0 predicts failure perfectly
      1.rrl6round dropped and 61 obs not used
```

```
note: 2.rrl6round != 0 predicts failure perfectly
      2.rrl6round dropped and 54 obs not used
```

```
note: 21.imm24round != 0 predicts success perfectly
      21.imm24round dropped and 33 obs not used
```

```
Iteration 0: log pseudolikelihood = -1082.8924
Iteration 1: log pseudolikelihood = -319.36372
Iteration 2: log pseudolikelihood = -293.72847
Iteration 3: log pseudolikelihood = -284.83038
Iteration 4: log pseudolikelihood = -284.70328
Iteration 5: log pseudolikelihood = -284.70321
Iteration 6: log pseudolikelihood = -284.70321
```

```
Logistic regression                                Number of obs      =      1,731
                                                    Wald chi2(51)      =      455.00
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -284.70321                Pseudo R2         =      0.7371
```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-1.360235	.6061244	-2.24	0.025	-2.548217	-.1722527
rrl6round						
1	0	(empty)				
2	0	(empty)				
3	1.238058	.9667889	1.28	0.200	-.6568136	3.132929
4	1.124682	1.186531	0.95	0.343	-1.200876	3.45024
5	1.362138	.8878194	1.53	0.125	-.3779565	3.102232
6	.9805327	.9076777	1.08	0.280	-.798483	2.759548
7	.5325362	.8604678	0.62	0.536	-1.15395	2.219022
8	.7614394	.8379483	0.91	0.364	-.8809092	2.403788
9	.1191418	.8591763	0.14	0.890	-1.564813	1.803096
10	1.507404	.9816143	1.54	0.125	-.4165252	3.431332
11	1.695403	.9056046	1.87	0.061	-.0795489	3.470356
12	2.424287	.9124382	2.66	0.008	.635941	4.212633
13	2.273862	.9351661	2.43	0.015	.4409703	4.106754
14	2.466966	1.013438	2.43	0.015	.4806645	4.453267
15	2.682943	1.027577	2.61	0.009	.6689288	4.696957
16	1.376619	.9316125	1.48	0.139	-.4493078	3.202546
imm24round						
1	.0031107	1.258666	0.00	0.998	-2.46383	2.470051
2	-2.894812	1.116009	-2.59	0.009	-5.08215	-.7074745
3	-2.669313	1.083711	-2.46	0.014	-4.793349	-.5452779
4	.2890952	1.443697	0.20	0.841	-2.540499	3.11869
5	-.7535717	.9708624	-0.78	0.438	-2.656427	1.149284
6	-.2039783	.9229202	-0.22	0.825	-2.012869	1.604912
7	.129697	1.166984	0.11	0.912	-2.15755	2.416944
8	1.023332	.9013909	1.14	0.256	-.7433615	2.790026
9	.9875586	.939922	1.05	0.293	-.8546547	2.829772
10	2.287368	1.015988	2.25	0.024	.2960692	4.278667
11	2.238614	.9592215	2.33	0.020	.3585748	4.118654
12	1.80754	.9409219	1.92	0.055	-.0366326	3.651713
13	1.822785	.9957536	1.83	0.067	-.1288567	3.774426

14	1.811171	1.015689	1.78	0.075	-.1795427	3.801885
15	1.365314	1.021048	1.34	0.181	-.635903	3.366531
16	3.085653	1.096335	2.81	0.005	.9368754	5.234431
17	2.160122	1.049036	2.06	0.039	.1040504	4.216194
18	3.061605	1.220568	2.51	0.012	.6693357	5.453874
19	1.267609	1.538855	0.82	0.410	-1.748492	4.28371
20	4.792068	1.089169	4.40	0.000	2.657335	6.9268
21	0	(empty)				
22	1.737327	1.20117	1.45	0.148	-.6169228	4.091577
23	3.062737	1.699716	1.80	0.072	-.2686444	6.394119
24	4.25284	1.549512	2.74	0.006	1.215851	7.289828
sexismf	4.176798	.8962458	4.66	0.000	2.420189	5.933408
econ1	-1.453844	.6481585	-2.24	0.025	-2.724212	-.1834771
age3044	.0353163	.6007675	0.06	0.953	-1.142166	1.212799
age4564	-.1219501	.6454971	-0.19	0.850	-1.387101	1.143201
age65p	-.0099405	.6191039	-0.02	0.987	-1.223362	1.203481
female	.4560316	.2443606	1.87	0.062	-.0229064	.9349697
prot	-.1462985	.316117	-0.46	0.644	-.7658765	.4732796
south	.7413509	.2934715	2.53	0.012	.1661574	1.316544
ba	.0387831	.2710296	0.14	0.886	-.4924251	.5699912
incl	-.453162	.5745646	-0.79	0.430	-1.579288	.672964
incdk	-.3889345	.9073175	-0.43	0.668	-2.167244	1.389375
pid1	5.032201	.4902605	10.26	0.000	4.071308	5.993094
ideol1	3.505632	.6607525	5.31	0.000	2.21058	4.800683
_cons	-6.878139	1.082619	-6.35	0.000	-9.000034	-4.756244

```
115 . logit votechoice jardina3 i.rrl6round012 i.imm24round23 sexismf econ1 age3044 age4564 age65p fem
> uth ba incl incdk pid1 ideol1 if white==1 [pw=wgt]
```

```
Iteration 0: log pseudolikelihood = -1196.262
Iteration 1: log pseudolikelihood = -323.81793
Iteration 2: log pseudolikelihood = -296.10628
Iteration 3: log pseudolikelihood = -288.72287
Iteration 4: log pseudolikelihood = -288.62445
Iteration 5: log pseudolikelihood = -288.62436
Iteration 6: log pseudolikelihood = -288.62436
```

```
Logistic regression                                Number of obs      =      1,879
                                                    Wald chi2(51)      =      447.70
                                                    Prob > chi2        =      0.0000
Log pseudolikelihood = -288.62436                Pseudo R2         =      0.7587
```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
jardina3	-1.460508	.6010373	-2.43	0.015	-2.638519	-.2824961
rrl6round012						
3	4.172859	1.28474	3.25	0.001	1.654815	6.690903
4	4.037584	1.385623	2.91	0.004	1.321813	6.753356
5	4.264569	1.235029	3.45	0.001	1.843957	6.685182
6	3.909617	1.262409	3.10	0.002	1.435341	6.383893
7	3.440582	1.193263	2.88	0.004	1.10183	5.779335
8	3.656739	1.150018	3.18	0.001	1.402744	5.910734
9	3.060675	1.193358	2.56	0.010	.7217373	5.399613
10	4.445417	1.285035	3.46	0.001	1.926795	6.964039
11	4.658896	1.188712	3.92	0.000	2.329063	6.988729
12	5.361303	1.214471	4.41	0.000	2.980984	7.741621
13	5.218073	1.206862	4.32	0.000	2.852667	7.583479
14	5.371923	1.256815	4.27	0.000	2.908611	7.835235
15	5.617647	1.308772	4.29	0.000	3.052501	8.182794
16	4.348249	1.185607	3.67	0.000	2.024503	6.671995
imm24round23						

	1	-.1989153	1.215561	-0.16	0.870	-2.581372	2.183541
	2	-2.983056	1.091964	-2.73	0.006	-5.123267	-.8428459
	3	-2.783561	1.060296	-2.63	0.009	-4.861703	-.7054188
	4	.1996232	1.426739	0.14	0.889	-2.596734	2.99598
	5	-.8583467	.9523305	-0.90	0.367	-2.72488	1.008187
	6	-.3495073	.8941347	-0.39	0.696	-2.101979	1.402965
	7	.0385868	1.133548	0.03	0.973	-2.183126	2.2603
	8	.9816799	.9001843	1.09	0.275	-.7826489	2.746009
	9	.9035449	.9173969	0.98	0.325	-.89452	2.70161
	10	2.205729	1.002076	2.20	0.028	.2416968	4.169761
	11	2.14349	.9395974	2.28	0.023	.3019126	3.985067
	12	1.72459	.92162	1.87	0.061	-.0817518	3.530932
	13	1.712184	.9748539	1.76	0.079	-.1984946	3.622862
	14	1.612346	.9789099	1.65	0.100	-.306282	3.530974
	15	1.299144	1.002458	1.30	0.195	-.6656364	3.263925
	16	2.980065	1.078309	2.76	0.006	.8666181	5.093511
	17	2.086941	1.035096	2.02	0.044	.0581895	4.115692
	18	3.212203	1.206565	2.66	0.008	.8473784	5.577028
	19	1.17873	1.52854	0.77	0.441	-1.817154	4.174613
	20	4.686969	1.073047	4.37	0.000	2.583837	6.790102
	21	4.222169	1.167087	3.62	0.000	1.93472	6.509618
	22	3.012481	1.704432	1.77	0.077	-.328145	6.353106
	23	4.179501	1.541984	2.71	0.007	1.157268	7.201734
sexismf		4.005603	.8934777	4.48	0.000	2.254418	5.756787
econl		-1.519063	.6490017	-2.34	0.019	-2.791083	-.2470426
age3044		.1432959	.5874567	0.24	0.807	-1.008098	1.29469
age4564		-.0627157	.6189685	-0.10	0.919	-1.275872	1.15044
age65p		.0999623	.5975137	0.17	0.867	-1.071143	1.271068
female		.3891505	.2440323	1.59	0.111	-.089144	.867445
prot		-.1220793	.316237	-0.39	0.699	-.7418924	.4977337
south		.8161393	.2948909	2.77	0.006	.2381637	1.394115
ba		.0678251	.2712383	0.25	0.803	-.4637922	.5994423
incl		-.506915	.5565349	-0.91	0.362	-1.597703	.5838735
incdk		-.4115485	.8799854	-0.47	0.640	-2.136288	1.313191
pidl		5.047592	.4940029	10.22	0.000	4.079364	6.01582
ideoll		3.457998	.6588623	5.25	0.000	2.166651	4.749344
_cons		-9.598498	1.354384	-7.09	0.000	-12.25304	-6.943954

```
116 . margins, atmeans at(rr16round012=(2 16))
```

Adjusted predictions	Number of obs	=	1,879
Model VCE : Robust			

Expression : `Pr(votechoice), predict()`

```
1._at      : jardina3      = .4270238 (mean)
            rr16round012   = 2
            0.imm24ro~23   = .0855609 (mean)
            1.imm24ro~23   = .0066538 (mean)
            2.imm24ro~23   = .0708983 (mean)
            3.imm24ro~23   = .0455311 (mean)
            4.imm24ro~23   = .0065541 (mean)
            5.imm24ro~23   = .0451681 (mean)
            6.imm24ro~23   = .0656377 (mean)
            7.imm24ro~23   = .016595 (mean)
            8.imm24ro~23   = .0454328 (mean)
            9.imm24ro~23   = .0629932 (mean)
            10.imm24r~23   = .0186011 (mean)
            11.imm24r~23   = .0626665 (mean)
            12.imm24r~23   = .0696351 (mean)
            13.imm24r~23   = .026348 (mean)
            14.imm24r~23   = .0849504 (mean)
            15.imm24r~23   = .0749089 (mean)
            16.imm24r~23   = .0197335 (mean)
            17.imm24r~23   = .0585239 (mean)
```

18.imm24r~23	=	.0338458	(mean)
19.imm24r~23	=	.0086352	(mean)
20.imm24r~23	=	.0435122	(mean)
21.imm24r~23	=	.0212146	(mean)
22.imm24r~23	=	.0138076	(mean)
23.imm24r~23	=	.0125921	(mean)
sexismf	=	.3719823	(mean)
econ1	=	.4743608	(mean)
age3044	=	.1959655	(mean)
age4564	=	.4108636	(mean)
age65p	=	.2584021	(mean)
female	=	.5317754	(mean)
prot	=	.3606696	(mean)
south	=	.2612745	(mean)
ba	=	.4006297	(mean)
incl	=	.6104942	(mean)
incdk	=	.0277694	(mean)
pid1	=	.5391016	(mean)
ideoll	=	.5575229	(mean)
2._at	:	jardina3	= .4270238 (mean)
		rrl6round012	= 16
		0.imm24ro~23	= .0855609 (mean)
		1.imm24ro~23	= .0066538 (mean)
		2.imm24ro~23	= .0708983 (mean)
		3.imm24ro~23	= .0455311 (mean)
		4.imm24ro~23	= .0065541 (mean)
		5.imm24ro~23	= .0451681 (mean)
		6.imm24ro~23	= .0656377 (mean)
		7.imm24ro~23	= .016595 (mean)
		8.imm24ro~23	= .0454328 (mean)
		9.imm24ro~23	= .0629932 (mean)
		10.imm24r~23	= .0186011 (mean)
		11.imm24r~23	= .0626665 (mean)
		12.imm24r~23	= .0696351 (mean)
		13.imm24r~23	= .026348 (mean)
		14.imm24r~23	= .0849504 (mean)
		15.imm24r~23	= .0749089 (mean)
		16.imm24r~23	= .0197335 (mean)
		17.imm24r~23	= .0585239 (mean)
		18.imm24r~23	= .0338458 (mean)
		19.imm24r~23	= .0086352 (mean)
		20.imm24r~23	= .0435122 (mean)
		21.imm24r~23	= .0212146 (mean)
		22.imm24r~23	= .0138076 (mean)
		23.imm24r~23	= .0125921 (mean)
		sexismf	= .3719823 (mean)
		econ1	= .4743608 (mean)
		age3044	= .1959655 (mean)
		age4564	= .4108636 (mean)
		age65p	= .2584021 (mean)
		female	= .5317754 (mean)
		prot	= .3606696 (mean)
		south	= .2612745 (mean)
		ba	= .4006297 (mean)
		incl	= .6104942 (mean)
		incdk	= .0277694 (mean)
		pid1	= .5391016 (mean)
		ideoll	= .5575229 (mean)

	Delta-method				[95% Conf. Interval]	
	Margin	Std. Err.	z	P> z		
_at						
1	.0234656	.0253047	0.93	0.354	-.0261306	.0730618
2	.6501676	.0977818	6.65	0.000	.4585187	.8418164

```
117 . di 65.0-2.3 // 62.7
62.7
```

```
118 . margins, atmeans at(imm24round23=(0 23))
```

```
Adjusted predictions          Number of obs      =      1,879
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at      :  jardim3          =      .4270238 (mean)
              2.rr16ro~012      =      .1184233 (mean)
              3.rr16ro~012      =      .0316444 (mean)
              4.rr16ro~012      =      .0424169 (mean)
              5.rr16ro~012      =      .0443095 (mean)
              6.rr16ro~012      =      .0473743 (mean)
              7.rr16ro~012      =      .0508838 (mean)
              8.rr16ro~012      =      .0859489 (mean)
              9.rr16ro~012      =      .0541805 (mean)
              10.rr16r~012       =      .0888366 (mean)
              11.rr16r~012       =      .0696695 (mean)
              12.rr16r~012       =      .0800114 (mean)
              13.rr16r~012       =      .0559795 (mean)
              14.rr16r~012       =      .0792267 (mean)
              15.rr16r~012       =      .0611771 (mean)
              16.rr16r~012       =      .0899185 (mean)
              imm24round23       =      0
              sexismf            =      .3719823 (mean)
              econ1              =      .4743608 (mean)
              age3044            =      .1959655 (mean)
              age4564            =      .4108636 (mean)
              age65p             =      .2584021 (mean)
              female              =      .5317754 (mean)
              prot                =      .3606696 (mean)
              south               =      .2612745 (mean)
              ba                  =      .4006297 (mean)
              incl                =      .6104942 (mean)
              incdk               =      .0277694 (mean)
              pid1                =      .5391016 (mean)
              ideoll              =      .5575229 (mean)
```

```
2._at      :  jardim3          =      .4270238 (mean)
              2.rr16ro~012      =      .1184233 (mean)
              3.rr16ro~012      =      .0316444 (mean)
              4.rr16ro~012      =      .0424169 (mean)
              5.rr16ro~012      =      .0443095 (mean)
              6.rr16ro~012      =      .0473743 (mean)
              7.rr16ro~012      =      .0508838 (mean)
              8.rr16ro~012      =      .0859489 (mean)
              9.rr16ro~012      =      .0541805 (mean)
              10.rr16r~012       =      .0888366 (mean)
              11.rr16r~012       =      .0696695 (mean)
              12.rr16r~012       =      .0800114 (mean)
              13.rr16r~012       =      .0559795 (mean)
              14.rr16r~012       =      .0792267 (mean)
              15.rr16r~012       =      .0611771 (mean)
              16.rr16r~012       =      .0899185 (mean)
              imm24round23       =      23
```

```

sexismf      =    .3719823 (mean)
econl        =    .4743608 (mean)
age3044      =    .1959655 (mean)
age4564      =    .4108636 (mean)
age65p       =    .2584021 (mean)
female       =    .5317754 (mean)
prot         =    .3606696 (mean)
south        =    .2612745 (mean)
ba           =    .4006297 (mean)
incl         =    .6104942 (mean)
incdk        =    .0277694 (mean)
pidl         =    .5391016 (mean)
ideol1       =    .5575229 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.3187344	.1793949	1.78	0.076	-.0328732	.670342
2	.9683209	.0388598	24.92	0.000	.8921572	1.044485

```

119 . di 96.8-31.9 // 64.9
64.9

```

```

120 .
121 . logit votechoice i.rrl6round012 i.imm24round23 if white==1 [pw=wt]

```

```

Iteration 0:  log pseudolikelihood = -1233.4658
Iteration 1:  log pseudolikelihood = -603.54559
Iteration 2:  log pseudolikelihood = -577.58061
Iteration 3:  log pseudolikelihood = -574.83456
Iteration 4:  log pseudolikelihood = -574.74263
Iteration 5:  log pseudolikelihood = -574.74253
Iteration 6:  log pseudolikelihood = -574.74253

```

```

Logistic regression              Number of obs      =      1,943
                                Wald chi2(37)         =      541.30
                                Prob > chi2           =      0.0000
Log pseudolikelihood = -574.74253 Pseudo R2          =      0.5340

```

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rrl6round012						
3	2.604378	.9046193	2.88	0.004	.8313573	4.3774
4	2.256984	.8959042	2.52	0.012	.5010441	4.012924
5	3.175412	.838916	3.79	0.000	1.531167	4.819657
6	2.604004	.8459606	3.08	0.002	.9459522	4.262057
7	3.524073	.8347689	4.22	0.000	1.887956	5.16019
8	3.35027	.8186718	4.09	0.000	1.745703	4.954837
9	3.160911	.8477246	3.73	0.000	1.499401	4.82242
10	3.853772	.8289403	4.65	0.000	2.229079	5.478466
11	4.052829	.8202775	4.94	0.000	2.445115	5.660544
12	4.606263	.8536467	5.40	0.000	2.933146	6.27938
13	5.041472	.8789195	5.74	0.000	3.318822	6.764123
14	4.677134	.8411167	5.56	0.000	3.028575	6.325692
15	5.427273	.8875433	6.11	0.000	3.68772	7.166826
16	4.743002	.9337388	5.08	0.000	2.912908	6.573096
imm24round23						
1	1.843269	.9443612	1.95	0.051	-.0076449	3.694183
2	-1.654169	.8433099	-1.96	0.050	-3.307026	-.0013118
3	-.8702204	.778135	-1.12	0.263	-2.395337	.6548963
4	1.601924	.789719	2.03	0.043	.0541035	3.149745

5	.7185709	.648778	1.11	0.268	-.5530107	1.990153
6	1.323237	.605472	2.19	0.029	.1365335	2.50994
7	2.243282	.6802215	3.30	0.001	.9100724	3.576492
8	2.591746	.6263509	4.14	0.000	1.36412	3.819371
9	2.185729	.6209344	3.52	0.000	.9687194	3.402738
10	3.723341	.7276617	5.12	0.000	2.297151	5.149532
11	3.796183	.6382533	5.95	0.000	2.54523	5.047137
12	3.533768	.6152108	5.74	0.000	2.327977	4.739559
13	3.056291	.7257183	4.21	0.000	1.63391	4.478673
14	4.104843	.6309896	6.51	0.000	2.868126	5.34156
15	3.276842	.6481712	5.06	0.000	2.00645	4.547234
16	5.345937	.9100587	5.87	0.000	3.562255	7.129619
17	4.195866	.7530209	5.57	0.000	2.719972	5.67176
18	4.070492	.8063177	5.05	0.000	2.490139	5.650846
19	4.347483	1.32397	3.28	0.001	1.75255	6.942416
20	5.197986	.9634702	5.40	0.000	3.30962	7.086353
21	5.306572	1.174345	4.52	0.000	3.004899	7.608245
22	4.93291	1.168934	4.22	0.000	2.641841	7.223979
23	4.535563	1.050782	4.32	0.000	2.476067	6.595058
_cons	-5.901813	.8511248	-6.93	0.000	-7.569987	-4.233639

```
122 . margins, atmeans at(rr16round012=(2(1)16))
```

```
Adjusted predictions                                Number of obs      =      1,943
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at      : rr16round012      =      2
0.imm24ro~23      =      .0883014 (mean)
1.imm24ro~23      =      .0064519 (mean)
2.imm24ro~23      =      .0687469 (mean)
3.imm24ro~23      =      .0464074 (mean)
4.imm24ro~23      =      .0067487 (mean)
5.imm24ro~23      =      .0450074 (mean)
6.imm24ro~23      =      .0646337 (mean)
7.imm24ro~23      =      .016623 (mean)
8.imm24ro~23      =      .0459719 (mean)
9.imm24ro~23      =      .0614723 (mean)
10.imm24r~23      =      .0186531 (mean)
11.imm24r~23      =      .0637626 (mean)
12.imm24r~23      =      .0701655 (mean)
13.imm24r~23      =      .0259049 (mean)
14.imm24r~23      =      .0844433 (mean)
15.imm24r~23      =      .0744172 (mean)
16.imm24r~23      =      .0194284 (mean)
17.imm24r~23      =      .0593815 (mean)
18.imm24r~23      =      .0335658 (mean)
19.imm24r~23      =      .0084834 (mean)
20.imm24r~23      =      .0421918 (mean)
21.imm24r~23      =      .0207639 (mean)
22.imm24r~23      =      .0143712 (mean)
23.imm24r~23      =      .0141028 (mean)
```

```

2._at      : rr16round012      =          3
0.imm24ro~23 = .0883014 (mean)
1.imm24ro~23 = .0064519 (mean)
2.imm24ro~23 = .0687469 (mean)
3.imm24ro~23 = .0464074 (mean)
4.imm24ro~23 = .0067487 (mean)
5.imm24ro~23 = .0450074 (mean)
6.imm24ro~23 = .0646337 (mean)
7.imm24ro~23 = .016623 (mean)
8.imm24ro~23 = .0459719 (mean)
9.imm24ro~23 = .0614723 (mean)
10.imm24r~23 = .0186531 (mean)
11.imm24r~23 = .0637626 (mean)
12.imm24r~23 = .0701655 (mean)
13.imm24r~23 = .0259049 (mean)
14.imm24r~23 = .0844433 (mean)
15.imm24r~23 = .0744172 (mean)
16.imm24r~23 = .0194284 (mean)
17.imm24r~23 = .0593815 (mean)
18.imm24r~23 = .0335658 (mean)
19.imm24r~23 = .0084834 (mean)
20.imm24r~23 = .0421918 (mean)
21.imm24r~23 = .0207639 (mean)
22.imm24r~23 = .0143712 (mean)
23.imm24r~23 = .0141028 (mean)

```

```

3._at      : rr16round012      =          4
0.imm24ro~23 = .0883014 (mean)
1.imm24ro~23 = .0064519 (mean)
2.imm24ro~23 = .0687469 (mean)
3.imm24ro~23 = .0464074 (mean)
4.imm24ro~23 = .0067487 (mean)
5.imm24ro~23 = .0450074 (mean)
6.imm24ro~23 = .0646337 (mean)
7.imm24ro~23 = .016623 (mean)
8.imm24ro~23 = .0459719 (mean)
9.imm24ro~23 = .0614723 (mean)
10.imm24r~23 = .0186531 (mean)
11.imm24r~23 = .0637626 (mean)
12.imm24r~23 = .0701655 (mean)
13.imm24r~23 = .0259049 (mean)
14.imm24r~23 = .0844433 (mean)
15.imm24r~23 = .0744172 (mean)
16.imm24r~23 = .0194284 (mean)
17.imm24r~23 = .0593815 (mean)
18.imm24r~23 = .0335658 (mean)
19.imm24r~23 = .0084834 (mean)
20.imm24r~23 = .0421918 (mean)
21.imm24r~23 = .0207639 (mean)
22.imm24r~23 = .0143712 (mean)
23.imm24r~23 = .0141028 (mean)

```

```

4._at      : rr16round012      =          5
0.imm24ro~23 = .0883014 (mean)
1.imm24ro~23 = .0064519 (mean)
2.imm24ro~23 = .0687469 (mean)
3.imm24ro~23 = .0464074 (mean)
4.imm24ro~23 = .0067487 (mean)
5.imm24ro~23 = .0450074 (mean)
6.imm24ro~23 = .0646337 (mean)
7.imm24ro~23 = .016623 (mean)
8.imm24ro~23 = .0459719 (mean)
9.imm24ro~23 = .0614723 (mean)
10.imm24r~23 = .0186531 (mean)
11.imm24r~23 = .0637626 (mean)
12.imm24r~23 = .0701655 (mean)
13.imm24r~23 = .0259049 (mean)

```

```

14.imm24r~23      =      .0844433 (mean)
15.imm24r~23      =      .0744172 (mean)
16.imm24r~23      =      .0194284 (mean)
17.imm24r~23      =      .0593815 (mean)
18.imm24r~23      =      .0335658 (mean)
19.imm24r~23      =      .0084834 (mean)
20.imm24r~23      =      .0421918 (mean)
21.imm24r~23      =      .0207639 (mean)
22.imm24r~23      =      .0143712 (mean)
23.imm24r~23      =      .0141028 (mean)

5._at      : rrl6round012      =      6
0.imm24ro~23     =      .0883014 (mean)
1.imm24ro~23     =      .0064519 (mean)
2.imm24ro~23     =      .0687469 (mean)
3.imm24ro~23     =      .0464074 (mean)
4.imm24ro~23     =      .0067487 (mean)
5.imm24ro~23     =      .0450074 (mean)
6.imm24ro~23     =      .0646337 (mean)
7.imm24ro~23     =      .016623 (mean)
8.imm24ro~23     =      .0459719 (mean)
9.imm24ro~23     =      .0614723 (mean)
10.imm24r~23     =      .0186531 (mean)
11.imm24r~23     =      .0637626 (mean)
12.imm24r~23     =      .0701655 (mean)
13.imm24r~23     =      .0259049 (mean)
14.imm24r~23     =      .0844433 (mean)
15.imm24r~23     =      .0744172 (mean)
16.imm24r~23     =      .0194284 (mean)
17.imm24r~23     =      .0593815 (mean)
18.imm24r~23     =      .0335658 (mean)
19.imm24r~23     =      .0084834 (mean)
20.imm24r~23     =      .0421918 (mean)
21.imm24r~23     =      .0207639 (mean)
22.imm24r~23     =      .0143712 (mean)
23.imm24r~23     =      .0141028 (mean)

6._at      : rrl6round012      =      7
0.imm24ro~23     =      .0883014 (mean)
1.imm24ro~23     =      .0064519 (mean)
2.imm24ro~23     =      .0687469 (mean)
3.imm24ro~23     =      .0464074 (mean)
4.imm24ro~23     =      .0067487 (mean)
5.imm24ro~23     =      .0450074 (mean)
6.imm24ro~23     =      .0646337 (mean)
7.imm24ro~23     =      .016623 (mean)
8.imm24ro~23     =      .0459719 (mean)
9.imm24ro~23     =      .0614723 (mean)
10.imm24r~23     =      .0186531 (mean)
11.imm24r~23     =      .0637626 (mean)
12.imm24r~23     =      .0701655 (mean)
13.imm24r~23     =      .0259049 (mean)
14.imm24r~23     =      .0844433 (mean)
15.imm24r~23     =      .0744172 (mean)
16.imm24r~23     =      .0194284 (mean)
17.imm24r~23     =      .0593815 (mean)
18.imm24r~23     =      .0335658 (mean)
19.imm24r~23     =      .0084834 (mean)
20.imm24r~23     =      .0421918 (mean)
21.imm24r~23     =      .0207639 (mean)
22.imm24r~23     =      .0143712 (mean)
23.imm24r~23     =      .0141028 (mean)

```

```

7._at      : rr16round012      =          8
0.imm24ro~23 = .0883014 (mean)
1.imm24ro~23 = .0064519 (mean)
2.imm24ro~23 = .0687469 (mean)
3.imm24ro~23 = .0464074 (mean)
4.imm24ro~23 = .0067487 (mean)
5.imm24ro~23 = .0450074 (mean)
6.imm24ro~23 = .0646337 (mean)
7.imm24ro~23 = .016623 (mean)
8.imm24ro~23 = .0459719 (mean)
9.imm24ro~23 = .0614723 (mean)
10.imm24r~23 = .0186531 (mean)
11.imm24r~23 = .0637626 (mean)
12.imm24r~23 = .0701655 (mean)
13.imm24r~23 = .0259049 (mean)
14.imm24r~23 = .0844433 (mean)
15.imm24r~23 = .0744172 (mean)
16.imm24r~23 = .0194284 (mean)
17.imm24r~23 = .0593815 (mean)
18.imm24r~23 = .0335658 (mean)
19.imm24r~23 = .0084834 (mean)
20.imm24r~23 = .0421918 (mean)
21.imm24r~23 = .0207639 (mean)
22.imm24r~23 = .0143712 (mean)
23.imm24r~23 = .0141028 (mean)

```

```

8._at      : rr16round012      =          9
0.imm24ro~23 = .0883014 (mean)
1.imm24ro~23 = .0064519 (mean)
2.imm24ro~23 = .0687469 (mean)
3.imm24ro~23 = .0464074 (mean)
4.imm24ro~23 = .0067487 (mean)
5.imm24ro~23 = .0450074 (mean)
6.imm24ro~23 = .0646337 (mean)
7.imm24ro~23 = .016623 (mean)
8.imm24ro~23 = .0459719 (mean)
9.imm24ro~23 = .0614723 (mean)
10.imm24r~23 = .0186531 (mean)
11.imm24r~23 = .0637626 (mean)
12.imm24r~23 = .0701655 (mean)
13.imm24r~23 = .0259049 (mean)
14.imm24r~23 = .0844433 (mean)
15.imm24r~23 = .0744172 (mean)
16.imm24r~23 = .0194284 (mean)
17.imm24r~23 = .0593815 (mean)
18.imm24r~23 = .0335658 (mean)
19.imm24r~23 = .0084834 (mean)
20.imm24r~23 = .0421918 (mean)
21.imm24r~23 = .0207639 (mean)
22.imm24r~23 = .0143712 (mean)
23.imm24r~23 = .0141028 (mean)

```

```

9._at      : rr16round012      =         10
0.imm24ro~23 = .0883014 (mean)
1.imm24ro~23 = .0064519 (mean)
2.imm24ro~23 = .0687469 (mean)
3.imm24ro~23 = .0464074 (mean)
4.imm24ro~23 = .0067487 (mean)
5.imm24ro~23 = .0450074 (mean)
6.imm24ro~23 = .0646337 (mean)
7.imm24ro~23 = .016623 (mean)
8.imm24ro~23 = .0459719 (mean)
9.imm24ro~23 = .0614723 (mean)
10.imm24r~23 = .0186531 (mean)
11.imm24r~23 = .0637626 (mean)
12.imm24r~23 = .0701655 (mean)
13.imm24r~23 = .0259049 (mean)

```

	14.imm24r~23	=	.0844433	(mean)
	15.imm24r~23	=	.0744172	(mean)
	16.imm24r~23	=	.0194284	(mean)
	17.imm24r~23	=	.0593815	(mean)
	18.imm24r~23	=	.0335658	(mean)
	19.imm24r~23	=	.0084834	(mean)
	20.imm24r~23	=	.0421918	(mean)
	21.imm24r~23	=	.0207639	(mean)
	22.imm24r~23	=	.0143712	(mean)
	23.imm24r~23	=	.0141028	(mean)
10._at	: rrl6round012	=	11	
	0.imm24ro~23	=	.0883014	(mean)
	1.imm24ro~23	=	.0064519	(mean)
	2.imm24ro~23	=	.0687469	(mean)
	3.imm24ro~23	=	.0464074	(mean)
	4.imm24ro~23	=	.0067487	(mean)
	5.imm24ro~23	=	.0450074	(mean)
	6.imm24ro~23	=	.0646337	(mean)
	7.imm24ro~23	=	.016623	(mean)
	8.imm24ro~23	=	.0459719	(mean)
	9.imm24ro~23	=	.0614723	(mean)
	10.imm24r~23	=	.0186531	(mean)
	11.imm24r~23	=	.0637626	(mean)
	12.imm24r~23	=	.0701655	(mean)
	13.imm24r~23	=	.0259049	(mean)
	14.imm24r~23	=	.0844433	(mean)
	15.imm24r~23	=	.0744172	(mean)
	16.imm24r~23	=	.0194284	(mean)
	17.imm24r~23	=	.0593815	(mean)
	18.imm24r~23	=	.0335658	(mean)
	19.imm24r~23	=	.0084834	(mean)
	20.imm24r~23	=	.0421918	(mean)
	21.imm24r~23	=	.0207639	(mean)
	22.imm24r~23	=	.0143712	(mean)
	23.imm24r~23	=	.0141028	(mean)
11._at	: rrl6round012	=	12	
	0.imm24ro~23	=	.0883014	(mean)
	1.imm24ro~23	=	.0064519	(mean)
	2.imm24ro~23	=	.0687469	(mean)
	3.imm24ro~23	=	.0464074	(mean)
	4.imm24ro~23	=	.0067487	(mean)
	5.imm24ro~23	=	.0450074	(mean)
	6.imm24ro~23	=	.0646337	(mean)
	7.imm24ro~23	=	.016623	(mean)
	8.imm24ro~23	=	.0459719	(mean)
	9.imm24ro~23	=	.0614723	(mean)
	10.imm24r~23	=	.0186531	(mean)
	11.imm24r~23	=	.0637626	(mean)
	12.imm24r~23	=	.0701655	(mean)
	13.imm24r~23	=	.0259049	(mean)
	14.imm24r~23	=	.0844433	(mean)
	15.imm24r~23	=	.0744172	(mean)
	16.imm24r~23	=	.0194284	(mean)
	17.imm24r~23	=	.0593815	(mean)
	18.imm24r~23	=	.0335658	(mean)
	19.imm24r~23	=	.0084834	(mean)
	20.imm24r~23	=	.0421918	(mean)
	21.imm24r~23	=	.0207639	(mean)
	22.imm24r~23	=	.0143712	(mean)
	23.imm24r~23	=	.0141028	(mean)

```

12._at      : rr16round012      =          13
              0.imm24ro~23      =    .0883014 (mean)
              1.imm24ro~23      =    .0064519 (mean)
              2.imm24ro~23      =    .0687469 (mean)
              3.imm24ro~23      =    .0464074 (mean)
              4.imm24ro~23      =    .0067487 (mean)
              5.imm24ro~23      =    .0450074 (mean)
              6.imm24ro~23      =    .0646337 (mean)
              7.imm24ro~23      =    .016623  (mean)
              8.imm24ro~23      =    .0459719 (mean)
              9.imm24ro~23      =    .0614723 (mean)
              10.imm24r~23      =    .0186531 (mean)
              11.imm24r~23      =    .0637626 (mean)
              12.imm24r~23      =    .0701655 (mean)
              13.imm24r~23      =    .0259049 (mean)
              14.imm24r~23      =    .0844433 (mean)
              15.imm24r~23      =    .0744172 (mean)
              16.imm24r~23      =    .0194284 (mean)
              17.imm24r~23      =    .0593815 (mean)
              18.imm24r~23      =    .0335658 (mean)
              19.imm24r~23      =    .0084834 (mean)
              20.imm24r~23      =    .0421918 (mean)
              21.imm24r~23      =    .0207639 (mean)
              22.imm24r~23      =    .0143712 (mean)
              23.imm24r~23      =    .0141028 (mean)

```

```

13._at      : rr16round012      =          14
              0.imm24ro~23      =    .0883014 (mean)
              1.imm24ro~23      =    .0064519 (mean)
              2.imm24ro~23      =    .0687469 (mean)
              3.imm24ro~23      =    .0464074 (mean)
              4.imm24ro~23      =    .0067487 (mean)
              5.imm24ro~23      =    .0450074 (mean)
              6.imm24ro~23      =    .0646337 (mean)
              7.imm24ro~23      =    .016623  (mean)
              8.imm24ro~23      =    .0459719 (mean)
              9.imm24ro~23      =    .0614723 (mean)
              10.imm24r~23      =    .0186531 (mean)
              11.imm24r~23      =    .0637626 (mean)
              12.imm24r~23      =    .0701655 (mean)
              13.imm24r~23      =    .0259049 (mean)
              14.imm24r~23      =    .0844433 (mean)
              15.imm24r~23      =    .0744172 (mean)
              16.imm24r~23      =    .0194284 (mean)
              17.imm24r~23      =    .0593815 (mean)
              18.imm24r~23      =    .0335658 (mean)
              19.imm24r~23      =    .0084834 (mean)
              20.imm24r~23      =    .0421918 (mean)
              21.imm24r~23      =    .0207639 (mean)
              22.imm24r~23      =    .0143712 (mean)
              23.imm24r~23      =    .0141028 (mean)

```

```

14._at      : rr16round012      =          15
              0.imm24ro~23      =    .0883014 (mean)
              1.imm24ro~23      =    .0064519 (mean)
              2.imm24ro~23      =    .0687469 (mean)
              3.imm24ro~23      =    .0464074 (mean)
              4.imm24ro~23      =    .0067487 (mean)
              5.imm24ro~23      =    .0450074 (mean)
              6.imm24ro~23      =    .0646337 (mean)
              7.imm24ro~23      =    .016623  (mean)
              8.imm24ro~23      =    .0459719 (mean)
              9.imm24ro~23      =    .0614723 (mean)
              10.imm24r~23      =    .0186531 (mean)
              11.imm24r~23      =    .0637626 (mean)
              12.imm24r~23      =    .0701655 (mean)
              13.imm24r~23      =    .0259049 (mean)

```

```

14.imm24r~23      =      .0844433 (mean)
15.imm24r~23      =      .0744172 (mean)
16.imm24r~23      =      .0194284 (mean)
17.imm24r~23      =      .0593815 (mean)
18.imm24r~23      =      .0335658 (mean)
19.imm24r~23      =      .0084834 (mean)
20.imm24r~23      =      .0421918 (mean)
21.imm24r~23      =      .0207639 (mean)
22.imm24r~23      =      .0143712 (mean)
23.imm24r~23      =      .0141028 (mean)

15._at      : rrl6round012      =      16
0.imm24ro~23      =      .0883014 (mean)
1.imm24ro~23      =      .0064519 (mean)
2.imm24ro~23      =      .0687469 (mean)
3.imm24ro~23      =      .0464074 (mean)
4.imm24ro~23      =      .0067487 (mean)
5.imm24ro~23      =      .0450074 (mean)
6.imm24ro~23      =      .0646337 (mean)
7.imm24ro~23      =      .016623 (mean)
8.imm24ro~23      =      .0459719 (mean)
9.imm24ro~23      =      .0614723 (mean)
10.imm24r~23      =      .0186531 (mean)
11.imm24r~23      =      .0637626 (mean)
12.imm24r~23      =      .0701655 (mean)
13.imm24r~23      =      .0259049 (mean)
14.imm24r~23      =      .0844433 (mean)
15.imm24r~23      =      .0744172 (mean)
16.imm24r~23      =      .0194284 (mean)
17.imm24r~23      =      .0593815 (mean)
18.imm24r~23      =      .0335658 (mean)
19.imm24r~23      =      .0084834 (mean)
20.imm24r~23      =      .0421918 (mean)
21.imm24r~23      =      .0207639 (mean)
22.imm24r~23      =      .0143712 (mean)
23.imm24r~23      =      .0141028 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.0321534	.0242476	1.33	0.185	-.0153711	.0796779
2	.3099877	.1025143	3.02	0.002	.1090634	.5109121
3	.240933	.0797953	3.02	0.003	.0845372	.3973288
4	.4429636	.0709553	6.24	0.000	.3038937	.5820336
5	.3099077	.0669897	4.63	0.000	.1786104	.4412051
6	.5298439	.0724491	7.31	0.000	.3878462	.6718415
7	.4864321	.0582463	8.35	0.000	.3722713	.6005928
8	.4393884	.0762	5.77	0.000	.2900392	.5887376
9	.6104531	.0647097	9.43	0.000	.4836246	.7372817
10	.6566215	.0569714	11.53	0.000	.5449596	.7682834
11	.7688285	.056298	13.66	0.000	.6584864	.8791706
12	.837115	.0533136	15.70	0.000	.7326224	.9416076
13	.7811839	.0508953	15.35	0.000	.6814309	.8809368
14	.8831618	.0419484	21.05	0.000	.8009445	.965379
15	.7922344	.0809679	9.78	0.000	.6335402	.9509286

```
123 . marginsplot
```

```
Variables that uniquely identify margins: rr16round012
```

```
124 . margins, atmeans at(imm24round23=(0(1)23))
```

```
Adjusted predictions                                Number of obs    =      1,943  
Model VCE      : Robust
```

```
Expression      : Pr(votechoice), predict()
```

```
1._at           : 2.rr16ro~012      =      .1182603 (mean)  
                  3.rr16ro~012      =      .0316779 (mean)  
                  4.rr16ro~012      =      .0426164 (mean)  
                  5.rr16ro~012      =      .0453131 (mean)  
                  6.rr16ro~012      =      .0485361 (mean)  
                  7.rr16ro~012      =      .0501375 (mean)  
                  8.rr16ro~012      =      .0851097 (mean)  
                  9.rr16ro~012      =      .052753  (mean)  
                 10.rr16r~012       =      .0871472 (mean)  
                 11.rr16r~012       =      .0688509 (mean)  
                 12.rr16r~012       =      .0811801 (mean)  
                 13.rr16r~012       =      .0559266 (mean)  
                 14.rr16r~012       =      .0795804 (mean)  
                 15.rr16r~012       =      .0608571 (mean)  
                 16.rr16r~012       =      .0920535 (mean)  
                 imm24round23       =              0
```

```
2._at           : 2.rr16ro~012      =      .1182603 (mean)  
                  3.rr16ro~012      =      .0316779 (mean)  
                  4.rr16ro~012      =      .0426164 (mean)  
                  5.rr16ro~012      =      .0453131 (mean)  
                  6.rr16ro~012      =      .0485361 (mean)  
                  7.rr16ro~012      =      .0501375 (mean)  
                  8.rr16ro~012      =      .0851097 (mean)  
                  9.rr16ro~012      =      .052753  (mean)  
                 10.rr16r~012       =      .0871472 (mean)  
                 11.rr16r~012       =      .0688509 (mean)  
                 12.rr16r~012       =      .0811801 (mean)  
                 13.rr16r~012       =      .0559266 (mean)  
                 14.rr16r~012       =      .0795804 (mean)  
                 15.rr16r~012       =      .0608571 (mean)  
                 16.rr16r~012       =      .0920535 (mean)  
                 imm24round23       =              1
```

```
3._at           : 2.rr16ro~012      =      .1182603 (mean)  
                  3.rr16ro~012      =      .0316779 (mean)  
                  4.rr16ro~012      =      .0426164 (mean)  
                  5.rr16ro~012      =      .0453131 (mean)  
                  6.rr16ro~012      =      .0485361 (mean)  
                  7.rr16ro~012      =      .0501375 (mean)  
                  8.rr16ro~012      =      .0851097 (mean)  
                  9.rr16ro~012      =      .052753  (mean)  
                 10.rr16r~012       =      .0871472 (mean)  
                 11.rr16r~012       =      .0688509 (mean)  
                 12.rr16r~012       =      .0811801 (mean)  
                 13.rr16r~012       =      .0559266 (mean)  
                 14.rr16r~012       =      .0795804 (mean)  
                 15.rr16r~012       =      .0608571 (mean)  
                 16.rr16r~012       =      .0920535 (mean)  
                 imm24round23       =              2
```

```

4._at      : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          3

```

```

5._at      : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          4

```

```

6._at      : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          5

```

```

7._at      : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          6

```

```

8._at      : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          7

```

```

9._at      : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          8

```

```

10._at     : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)
             imm24round23     =          9

```

```

11._at     : 2.rr16ro~012    =    .1182603 (mean)
              3.rr16ro~012    =    .0316779 (mean)
              4.rr16ro~012    =    .0426164 (mean)
              5.rr16ro~012    =    .0453131 (mean)
              6.rr16ro~012    =    .0485361 (mean)
              7.rr16ro~012    =    .0501375 (mean)
              8.rr16ro~012    =    .0851097 (mean)
              9.rr16ro~012    =    .052753  (mean)
             10.rr16r~012     =    .0871472 (mean)
             11.rr16r~012     =    .0688509 (mean)
             12.rr16r~012     =    .0811801 (mean)
             13.rr16r~012     =    .0559266 (mean)
             14.rr16r~012     =    .0795804 (mean)
             15.rr16r~012     =    .0608571 (mean)
             16.rr16r~012     =    .0920535 (mean)

```

	imm24round23	=	10	
12._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	11	
13._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	12	
14._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	13	
15._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)

	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	14	
16._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	15	
17._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	16	
18._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	17	
19._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)

	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	18	
20._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	19	
21._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	20	
22._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)
	14.rr16r~012	=	.0795804	(mean)
	15.rr16r~012	=	.0608571	(mean)
	16.rr16r~012	=	.0920535	(mean)
	imm24round23	=	21	
23._at	: 2.rr16ro~012	=	.1182603	(mean)
	3.rr16ro~012	=	.0316779	(mean)
	4.rr16ro~012	=	.0426164	(mean)
	5.rr16ro~012	=	.0453131	(mean)
	6.rr16ro~012	=	.0485361	(mean)
	7.rr16ro~012	=	.0501375	(mean)
	8.rr16ro~012	=	.0851097	(mean)
	9.rr16ro~012	=	.052753	(mean)
	10.rr16r~012	=	.0871472	(mean)
	11.rr16r~012	=	.0688509	(mean)
	12.rr16r~012	=	.0811801	(mean)
	13.rr16r~012	=	.0559266	(mean)

```

14.rr16r~012      =      .0795804 (mean)
15.rr16r~012      =      .0608571 (mean)
16.rr16r~012      =      .0920535 (mean)
imm24round23      =           22

24._at      : 2.rr16ro~012      =      .1182603 (mean)
               3.rr16ro~012      =      .0316779 (mean)
               4.rr16ro~012      =      .0426164 (mean)
               5.rr16ro~012      =      .0453131 (mean)
               6.rr16ro~012      =      .0485361 (mean)
               7.rr16ro~012      =      .0501375 (mean)
               8.rr16ro~012      =      .0851097 (mean)
               9.rr16ro~012      =      .052753 (mean)
              10.rr16r~012      =      .0871472 (mean)
              11.rr16r~012      =      .0688509 (mean)
              12.rr16r~012      =      .0811801 (mean)
              13.rr16r~012      =      .0559266 (mean)
              14.rr16r~012      =      .0795804 (mean)
              15.rr16r~012      =      .0608571 (mean)
              16.rr16r~012      =      .0920535 (mean)
              imm24round23      =           23

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.0820818	.0416943	1.97	0.049	.0003625	.163801
2	.3609777	.1711666	2.11	0.035	.0254972	.6964581
3	.0168144	.0104095	1.62	0.106	-.0035879	.0372168
4	.0361028	.0192146	1.88	0.060	-.0015571	.0737627
5	.3073648	.1244753	2.47	0.014	.0633977	.5513319
6	.1550118	.0436486	3.55	0.000	.0694621	.2405614
7	.2514007	.0479311	5.25	0.000	.1574576	.3453439
8	.4573265	.0978675	4.67	0.000	.2655097	.6491433
9	.5442224	.0726129	7.49	0.000	.4019037	.6865411
10	.4430818	.0674056	6.57	0.000	.3109692	.5751944
11	.7873372	.0785142	10.03	0.000	.6334522	.9412223
12	.7992785	.0500245	15.98	0.000	.7012323	.8973247
13	.7538729	.0462422	16.30	0.000	.6632399	.8445059
14	.6551826	.1058797	6.19	0.000	.4476622	.8627029
15	.8442834	.0397079	21.26	0.000	.7664573	.9221095
16	.7031722	.0682062	10.31	0.000	.5694904	.8368539
17	.949384	.0335124	28.33	0.000	.8837009	1.015067
18	.8558788	.0617422	13.86	0.000	.7348663	.9768912
19	.8397137	.0793434	10.58	0.000	.6842035	.995224
20	.8735911	.1325655	6.59	0.000	.6137674	1.133415
21	.9417828	.0430033	21.90	0.000	.8574978	1.026068
22	.9474585	.0511478	18.52	0.000	.8472106	1.047706
23	.9254298	.070533	13.12	0.000	.7871878	1.063672
24	.8929438	.0839591	10.64	0.000	.728387	1.057501

```
125 . marginsplot
```

```
Variables that uniquely identify margins: imm24round23
```

```

126 .
127 . recode immf (0/0.1=0) (.1/.2=1) (.2/.3=2) (.3/.4=3) (.4/.5=4) (.5/.6=5) (.6/.7=6) (.7/.8=7) (.8/.
> =9), gen(imm_i)
(4234 differences between immf and imm_i)

```

```

128 . tab imm_i

```

RECODE of immf	Freq.	Percent	Cum.
0	249	5.84	5.84
1	467	10.95	16.78
2	586	13.74	30.52
3	641	15.03	45.55
4	660	15.47	61.02
5	618	14.49	75.50
6	498	11.67	87.18
7	322	7.55	94.73
8	148	3.47	98.20
9	77	1.80	100.00
Total	4,266	100.00	

```

129 . recode rrf (0/0.1=0) (.1/.2=1) (.2/.3=2) (.3/.4=3) (.4/.5=4) (.5/.6=5) (.6/.7=6) (.7/.8=7) (.8/.
> 9), gen(rr_i)
(3452 differences between rrf and rr_i)

```

```

130 . tab rr_i

```

RECODE of rrf	Freq.	Percent	Cum.
0	312	8.57	8.57
1	230	6.32	14.89
2	190	5.22	20.12
3	362	9.95	30.06
4	649	17.83	47.90
5	260	7.14	55.04
6	527	14.48	69.52
7	264	7.25	76.78
8	385	10.58	87.36
9	460	12.64	100.00
Total	3,639	100.00	

```

131 .
132 . logit votechoice rr_i imm_i if white==1 [pw=wgt]

```

```

Iteration 0:  log pseudolikelihood = -1233.4658
Iteration 1:  log pseudolikelihood = -645.83372
Iteration 2:  log pseudolikelihood = -642.27334
Iteration 3:  log pseudolikelihood = -642.25894
Iteration 4:  log pseudolikelihood = -642.25894

```

Logistic regression	Number of obs	=	1,943
	Wald chi2(2)	=	359.80
	Prob > chi2	=	0.0000
Log pseudolikelihood = -642.25894	Pseudo R2	=	0.4793


```
134 . di 87.0-19.0 // 68.0
68
```

```
135 . marginsplot
```

Variables that uniquely identify margins: rr_i

```
136 . margins, atmeans at(imm_i=(0(1)9))
```

Adjusted predictions Number of obs = **1,943**
Model VCE : **Robust**

Expression : **Pr(votechoice), predict()**

```
1._at      : rr_i      = 5.195676 (mean)
             imm_i      = 0
2._at      : rr_i      = 5.195676 (mean)
             imm_i      = 1
3._at      : rr_i      = 5.195676 (mean)
             imm_i      = 2
4._at      : rr_i      = 5.195676 (mean)
             imm_i      = 3
5._at      : rr_i      = 5.195676 (mean)
             imm_i      = 4
6._at      : rr_i      = 5.195676 (mean)
             imm_i      = 5
7._at      : rr_i      = 5.195676 (mean)
             imm_i      = 6
8._at      : rr_i      = 5.195676 (mean)
             imm_i      = 7
9._at      : rr_i      = 5.195676 (mean)
             imm_i      = 8
10._at     : rr_i      = 5.195676 (mean)
             imm_i      = 9
```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Intervall	
_at						
1	.0611209	.0131605	4.64	0.000	.0353268	.086915
2	.1234057	.0188499	6.55	0.000	.0864607	.1603508
3	.2333841	.0221106	10.56	0.000	.1900482	.27672
4	.3969873	.020484	19.38	0.000	.3568393	.4371353
5	.5874012	.0195585	30.03	0.000	.5490672	.6257352
6	.7548223	.0209736	35.99	0.000	.7137147	.7959298
7	.8694115	.0184179	47.20	0.000	.833313	.90551
8	.935053	.0131558	71.08	0.000	.9092682	.9608378
9	.9688803	.0082398	117.59	0.000	.9527305	.98503
10	.9853645	.0047738	206.41	0.000	.9760081	.994721

```
137 . di 98.5-6.1 // 92.4
92.4
```

```
138 . marginsplot
```

Variables that uniquely identify margins: imm_i

```
139 . di 92.4-68.0 // 24.4
24.4
```

```
140 .
141 . logit votechoice i.rr_i i.imm_i if white==1 [pw=wgt]
```

```
Iteration 0: log pseudolikelihood = -1233.4658
Iteration 1: log pseudolikelihood = -635.53544
Iteration 2: log pseudolikelihood = -618.56293
Iteration 3: log pseudolikelihood = -617.22844
Iteration 4: log pseudolikelihood = -617.20207
Iteration 5: log pseudolikelihood = -617.20201
Iteration 6: log pseudolikelihood = -617.20201
```

Logistic regression	Number of obs	=	1,943
	Wald chi2(18)	=	470.59
	Prob > chi2	=	0.0000
Log pseudolikelihood = -617.20201	Pseudo R2	=	0.4996

votechoice	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
rr_i						
1	.458318	1.020143	0.45	0.653	-1.541126	2.457762
2	1.394807	.9792606	1.42	0.154	-.5245089	3.314122
3	1.888208	.8586052	2.20	0.028	.205373	3.571043
4	2.243956	.8581964	2.61	0.009	.5619219	3.92599
5	2.01627	.8859927	2.28	0.023	.2797562	3.752784
6	2.785511	.8685295	3.21	0.001	1.083224	4.487797
7	3.502762	.9018225	3.88	0.000	1.735223	5.270302
8	3.614259	.8686398	4.16	0.000	1.911756	5.316761
9	3.744633	.8951441	4.18	0.000	1.990183	5.499083
imm_i						
1	-.6368315	1.275879	-0.50	0.618	-3.137509	1.863846
2	.8529341	1.22047	0.70	0.485	-1.539143	3.245011
3	2.030131	1.215377	1.67	0.095	-.3519638	4.412225
4	3.22718	1.214177	2.66	0.008	.8474377	5.606922
5	3.957259	1.215726	3.26	0.001	1.57448	6.340038
6	3.982279	1.226286	3.25	0.001	1.578804	6.385755
7	4.843445	1.242558	3.90	0.000	2.408077	7.278814
8	4.442466	1.307296	3.40	0.001	1.880214	7.004719
9	4.605309	1.403642	3.28	0.001	1.854221	7.356397
_cons	-5.016316	.9552258	-5.25	0.000	-6.888524	-3.144107

142 . margins, atmeans at(rr_i=(0(1)9))

Adjusted predictions Number of obs = 1,943
Model VCE : **Robust**

Expression : **Pr(votechoice), predict()**

1._at : rr_i = 0
 0.imm_i = .053635 (mean)
 1.imm_i = .1061179 (mean)
 2.imm_i = .1076429 (mean)
 3.imm_i = .1193712 (mean)
 4.imm_i = .1426666 (mean)
 5.imm_i = .1585204 (mean)
 6.imm_i = .1470988 (mean)
 7.imm_i = .0965598 (mean)
 8.imm_i = .0480068 (mean)
 9.imm_i = .0203807 (mean)

2._at : rr_i = 1
 0.imm_i = .053635 (mean)
 1.imm_i = .1061179 (mean)
 2.imm_i = .1076429 (mean)
 3.imm_i = .1193712 (mean)
 4.imm_i = .1426666 (mean)
 5.imm_i = .1585204 (mean)
 6.imm_i = .1470988 (mean)
 7.imm_i = .0965598 (mean)
 8.imm_i = .0480068 (mean)
 9.imm_i = .0203807 (mean)

3._at : rr_i = 2
 0.imm_i = .053635 (mean)
 1.imm_i = .1061179 (mean)
 2.imm_i = .1076429 (mean)
 3.imm_i = .1193712 (mean)
 4.imm_i = .1426666 (mean)
 5.imm_i = .1585204 (mean)
 6.imm_i = .1470988 (mean)
 7.imm_i = .0965598 (mean)
 8.imm_i = .0480068 (mean)
 9.imm_i = .0203807 (mean)

4._at : rr_i = 3
 0.imm_i = .053635 (mean)
 1.imm_i = .1061179 (mean)
 2.imm_i = .1076429 (mean)
 3.imm_i = .1193712 (mean)
 4.imm_i = .1426666 (mean)
 5.imm_i = .1585204 (mean)
 6.imm_i = .1470988 (mean)
 7.imm_i = .0965598 (mean)
 8.imm_i = .0480068 (mean)
 9.imm_i = .0203807 (mean)

5._at : rr_i = 4
 0.imm_i = .053635 (mean)
 1.imm_i = .1061179 (mean)
 2.imm_i = .1076429 (mean)
 3.imm_i = .1193712 (mean)
 4.imm_i = .1426666 (mean)
 5.imm_i = .1585204 (mean)
 6.imm_i = .1470988 (mean)
 7.imm_i = .0965598 (mean)
 8.imm_i = .0480068 (mean)
 9.imm_i = .0203807 (mean)

```

6._at      : rr_i      =          5
             0.imm_i    =    .053635 (mean)
             1.imm_i    =    .1061179 (mean)
             2.imm_i    =    .1076429 (mean)
             3.imm_i    =    .1193712 (mean)
             4.imm_i    =    .1426666 (mean)
             5.imm_i    =    .1585204 (mean)
             6.imm_i    =    .1470988 (mean)
             7.imm_i    =    .0965598 (mean)
             8.imm_i    =    .0480068 (mean)
             9.imm_i    =    .0203807 (mean)

```

```

7._at      : rr_i      =          6
             0.imm_i    =    .053635 (mean)
             1.imm_i    =    .1061179 (mean)
             2.imm_i    =    .1076429 (mean)
             3.imm_i    =    .1193712 (mean)
             4.imm_i    =    .1426666 (mean)
             5.imm_i    =    .1585204 (mean)
             6.imm_i    =    .1470988 (mean)
             7.imm_i    =    .0965598 (mean)
             8.imm_i    =    .0480068 (mean)
             9.imm_i    =    .0203807 (mean)

```

```

8._at      : rr_i      =          7
             0.imm_i    =    .053635 (mean)
             1.imm_i    =    .1061179 (mean)
             2.imm_i    =    .1076429 (mean)
             3.imm_i    =    .1193712 (mean)
             4.imm_i    =    .1426666 (mean)
             5.imm_i    =    .1585204 (mean)
             6.imm_i    =    .1470988 (mean)
             7.imm_i    =    .0965598 (mean)
             8.imm_i    =    .0480068 (mean)
             9.imm_i    =    .0203807 (mean)

```

```

9._at      : rr_i      =          8
             0.imm_i    =    .053635 (mean)
             1.imm_i    =    .1061179 (mean)
             2.imm_i    =    .1076429 (mean)
             3.imm_i    =    .1193712 (mean)
             4.imm_i    =    .1426666 (mean)
             5.imm_i    =    .1585204 (mean)
             6.imm_i    =    .1470988 (mean)
             7.imm_i    =    .0965598 (mean)
             8.imm_i    =    .0480068 (mean)
             9.imm_i    =    .0203807 (mean)

```

```

10._at     : rr_i      =          9
             0.imm_i    =    .053635 (mean)
             1.imm_i    =    .1061179 (mean)
             2.imm_i    =    .1076429 (mean)
             3.imm_i    =    .1193712 (mean)
             4.imm_i    =    .1426666 (mean)
             5.imm_i    =    .1585204 (mean)
             6.imm_i    =    .1470988 (mean)
             7.imm_i    =    .0965598 (mean)
             8.imm_i    =    .0480068 (mean)
             9.imm_i    =    .0203807 (mean)

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Interval]	
_at						
1	.0910048	.06587	1.38	0.167	-.0380982	.2201077
2	.1366839	.074524	1.83	0.067	-.0093806	.2827483
3	.2876917	.0849781	3.39	0.001	.1211376	.4542458
4	.3981403	.0572795	6.95	0.000	.2858745	.5104061
5	.4856359	.0464527	10.45	0.000	.3945902	.5766816
6	.4291897	.0682121	6.29	0.000	.2954965	.5628828
7	.6187115	.0501039	12.35	0.000	.5205097	.7169132
8	.7687621	.057633	13.34	0.000	.6558034	.8817207
9	.7879864	.0378254	20.83	0.000	.7138499	.8621229
10	.8089504	.0508568	15.91	0.000	.709273	.9086279

```
143 . di 80.9-9.1 // 71.8
    71.8
```

144 . marginsplot

Variables that uniquely identify margins: rr i

```
145 . margins, atmeans at(imm i=(0(1)9))
```

Adjusted predictions	Number of obs	=	1,943
Model VCE : Robust			

Expression : `Pr(votechoice), predict()`

```
1._at      : 0.rr_i      =      .0902979 (mean)
            1.rr_i      =      .0596403 (mean)
            2.rr_i      =      .0426164 (mean)
            3.rr_i      =      .0938492 (mean)
            4.rr_i      =      .1352472 (mean)
            5.rr_i      =      .052753  (mean)
            6.rr_i      =      .1559982 (mean)
            7.rr_i      =      .0811801 (mean)
            8.rr_i      =      .135507  (mean)
            9.rr_i      =      .1529106 (mean)
            imm i       =      0
```

```
2._at      : 0.rr_i      =      .0902979 (mean)
            1.rr_i      =      .0596403 (mean)
            2.rr_i      =      .0426164 (mean)
            3.rr_i      =      .0938492 (mean)
            4.rr_i      =      .1352472 (mean)
            5.rr_i      =      .052753  (mean)
            6.rr_i      =      .1559982 (mean)
            7.rr_i      =      .0811801 (mean)
            8.rr_i      =      .135507  (mean)
            9.rr_i      =      .1529106 (mean)
            imm i        =      1
```

```
3._at      : 0.rr_i      =      .0902979 (mean)
            1.rr_i      =      .0596403 (mean)
            2.rr_i      =      .0426164 (mean)
            3.rr_i      =      .0938492 (mean)
            4.rr_i      =      .1352472 (mean)
            5.rr_i      =      .052753  (mean)
            6.rr_i      =      .1559982 (mean)
            7.rr_i      =      .0811801 (mean)
            8.rr_i      =      .135507  (mean)
            9.rr_i      =      .1529106 (mean)
            imm i       =      2
```

```

4._at      : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i      = 3

5._at      : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i      = 4

6._at      : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i      = 5

7._at      : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i      = 6

8._at      : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i      = 7

```

```

9._at      : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i       = 8

10._at     : 0.rr_i      = .0902979 (mean)
              1.rr_i      = .0596403 (mean)
              2.rr_i      = .0426164 (mean)
              3.rr_i      = .0938492 (mean)
              4.rr_i      = .1352472 (mean)
              5.rr_i      = .052753 (mean)
              6.rr_i      = .1559982 (mean)
              7.rr_i      = .0811801 (mean)
              8.rr_i      = .135507 (mean)
              9.rr_i      = .1529106 (mean)
              imm_i       = 9

```

	Delta-method					
	Margin	Std. Err.	z	P> z	[95% Conf. Intervall	
_at						
1	.0716751	.0770223	0.93	0.352	-.079286	.2226361
2	.0392384	.0198494	1.98	0.048	.0003342	.0781426
3	.1533837	.0332362	4.61	0.000	.0882419	.2185255
4	.3702584	.0461767	8.02	0.000	.2797537	.460763
5	.6605937	.0424496	15.56	0.000	.5773941	.7437933
6	.8015516	.0358697	22.35	0.000	.7312482	.8718549
7	.8055014	.0389958	20.66	0.000	.729071	.8819318
8	.9073926	.0288178	31.49	0.000	.8509108	.9638744
9	.867752	.0590057	14.71	0.000	.752103	.983401
10	.885347	.0749602	11.81	0.000	.7384277	1.032266

```

146 . di 88.5-7.2 // 81.3
    81.3

```

```

147 . marginsplot

    Variables that uniquely identify margins: imm_i

```

```

148 . di 81.3 - 71.8 // 9.5
    9.5

```

```

149 .
    end of do-file

```

```

150 . log close
    name: <unnamed>
    log: C:\Users\ljzig\Desktop\buyaker et al 2020\LJZ buyaker et al 2020.smcl
    log type: smcl
    closed on: 31 Dec 2020, 12:19:31

```