Problem Set 2 Solutions

- 1. Provide some basic descriptive information about the students in this dataset. How many observations are there? What proportion attended a Catholic high school? What proportion graduated high school on time? What proportion entered post-secondary education after high school? What are the overall means and standard deviations for 12th grade math and reading scores, respectively? (5 points)
 - See attached log. There are 5,671 student observations. 10.4% attended a Catholic high school. 91.7% graduated from high school. 70.9% enrolled in post-secondary education. The means (standard deviations) for math and reading are 51.1 (9.5) and 51.0 (9.5).
- 2. Now provide descriptive statistics (means, standard deviations) separately for Catholic and all other students. How do these populations of students differ, if at all? Include all outcomes and background variables that you think are relevant for a comparison of academic achievement between Catholic and all other students. (5 points)
 - See attached log. Some notable differences between Catholic high school students and non-Catholic high school students include the following. Mean academic outcomes were higher for Catholic school students (test scores in 12th grade, graduation rates, and post-secondary enrollment rates). Their baseline 8th grade scores were also higher. Catholic school students were more likely to be female and white, and their family income was about \$8,000 higher on average. Parents of Catholic school students were more likely to be college-educated, and had higher educational aspirations for their child.
- 3. Estimate simple bivariate regressions relating each of the following outcomes to Catholic high school attendance: math z-score, reading z-score, on-time high school graduation, enrolled in post-secondary education. Provide a brief explanation of what you find. Does there appear to be a Catholic high school "advantage"? If so, is the difference statistically significant? Practically significant? Is the slope coefficient here a plausible estimate of the average treatment effect on the treated (ATT)? Why or why not? (5 points)

See attached log. For all four outcomes, Catholic high school graduates fared better than other high school graduates. All estimated slope coefficients are statistically significant at the 1% level or better. Catholic high school students performed 0.41 and 0.37 standard deviations higher in math and reading, on average, than other students, were 6.4 percentage points

more likely to graduate from high school, and were 17 percentage points more likely to be enrolled in post-secondary education. These are all large and practically significant effects. For perspective on practical significance, compare these effects to the sample standard deviations in the outcomes.

The ATT assumes we have a comparison group that is on average equivalent to the treatment group. (Its potential outcomes in the untreated state should be the same in expectation as those of the treated group). It seems unlikely that students who did not attend Catholic school have the same potential outcomes, on average, as those who attended Catholic school. Our comparison of observable characteristics in part (2) suggests these populations are actually quite different. Thus, any direct comparison of their mean outcomes is likely to be affected by selection bias.

4. Provide a brief rationale for why one might prefer matching or a re-weighting approach to a regression model. What conditions must be satisfied for a treatment effect estimate based on matching or re-weighting to be convincing as a causal estimate? (5 points)

Matching and re-weighting allows us to compare treated cases to "observationally similar" non-treated cases with similar covariates and/or a similar probability of treatment. These analyses strive for unconfoundedness or conditional independence. For unbiasedness, it must be the case that potential outcomes are independent of treatment assignment conditional on the covariates or propensity score. (For example, two individuals with the same propensity score should have the same potential outcomes, in expectation, in each treatment state). These approaches also require common support, a sufficient number of treated and untreated observations with the same covariates and/or propensity scores.

As noted in class, these methods are not necessarily "better" at solving the omitted variables bias problem than regression. Both assume that conditioning on observable controls is sufficient. Matching and re-weighting have some advantages over regression, however, since they do not rely on a functional form assumption.

5. Use teffects nnmatch to estimate the ATE and ATT (ATET in Stata) of Catholic school attendance by exact matching on family income, using the 3-category version you created above. Use the same four outcomes from part (3). Summarize what you find. (This analysis is comparable to the subclassification example from Table 12.1 in Murnane & Willett—see the lecture notes. If you want to compare your estimates to theirs, do this step again with *math12*, the math score on its original scale). Note you can use Stata's factor variable notation in this command. (5 points)

See attached log and tabular summary of results below. The estimated effects are somewhat smaller than those estimated using OLS in part (3). math12 is the math test score in its original units. The estimates for math12

can be compared to Murnane & Willett's ATE and ATT of 3.01 and 2.74, respectively.

6. Re-estimate the exact matching ATT from part (5) using the math z-score as the outcome. Following that, use the tebalance summarize command to check for balance on the following variables: your 3-category family income measure, your family income measure in dollars, and 8th grade math and reading z-scores. (You can conduct balance checks on variables that were not included in your original exact matching algorithm). How do the Catholic and public school students in the matched sample compare with respect to their distribution of these variables? In light of your findings here, how comfortable are you interpreting the estimates in part (5) as causal? (5 points)

Results shown in the attached log. Recall that the "standardized difference" is the difference in means between the treated and matched untreated cases, divided by the square root of a pooled variance estimator. (So it can be interpreted as standard deviation units). Family income appears well-balanced across the treated and matched untreated samples—even the continuous measure faminc8b. This is not too surprising, as exact matching was conducted on the 3-category income measure. Note the imbalance in the 8th grade math score, however, with Catholic school students still having noticeably higher prior math achievement than matched non-Catholic high school students. Since prior achievement is likely to be associated with 12th grade and post-secondary potential outcomes, we should be concerned about this lack of balance when interpreting these estimates.

7. In this part, you will develop a propensity score model to later estimate the ATT of Catholic school attendance. The first step of a propensity score analysis is to determine which confounding variables should be included in the estimation of the propensity score. You should be able to defend your choices based on theory and your understanding of the likely factors predicting selection into Catholic school and subsequent academic outcomes. Ultimately, however, the aim of a propensity score analysis is to create balance in the treated and untreated groups. (15 points)

Your first task will be to settle on a propensity score model, iterating on included covariates and model specifications and checking balance until you are reasonably satisfied with the balance you have attained.

My final model is shown in the attached log (yours will likely differ). Of the confounders, I prioritized balance on baseline test scores, family income, and parental education, given the results of prior studies which demonstrate the importance of these in explaining later outcomes. The probit model results confirm these are important predictors of Catholic school enrollment as well. I experimented with several different specifications and variable definitions; these included models linear in prior test scores, models that used the highest education of either parent, models without interactions between gender and race, and models with a continuous measure of risk factors. The covariate balance was quite good for most specifications, but this one seemed especially good, as the pstest results show. (The log file also shows the tebalance summarize for the same propensity score match).

The figure shows the distribution of estimated propensity scores for both the untreated and treated groups. These graphs are typically produced for the full dataset (i.e., not just the matched sample). I include some code that shows these distributions for the matched sample only.

- 8. Using your final propensity score model in part (5), answer the following:
 - (a) Interpret your propensity score model. What types of students are more or less likely to be "treated" (i.e., attend Catholic high school)? You do not need to interpret the specific probit or logit coefficients since these do not have a natural interpretation. Hint: if you used teffects psmatch, you will not see the coefficient estimates from the propensity score model. In this case I recommend estimating your model in psmatch2 or using logit or probit directly. (5 points) The probit model results are included in the log. Reading achievement is a strong predictor of Catholic high school enrollment, although math is not (when simultaneously controlling for reading achievement). Hispanic and white students are more likely to attend Catholic school than Black, Asian, and other race/ethnic groups, and in general girls are more likely to have attended Catholic school. Catholic school enrollment is positively related to family income, parents' education (especially that of the mother), and expectations of college enrollment. Students deemed at risk of drop out and those with a history of not completing homework are less likely to have attended Catholic school.
 - (b) Calculate a ATT estimate for each of the four outcome variables (math and reading test scores, high school graduation, and post-secondary enrollment). Provide a written interpretation of your treatment effect estimates. How do these differ from those you estimated using earlier methods (regression and exact matching)? (5 points)

The results are shown in the attached log and summarized in the table below. Attending a Catholic high school is estimated to have a 0.120 standard deviation effect on math scores and a -0.003 standard deviation effect on reading scores. Only the math effect is statistically significant. The ATT estimate is positive and statistically insignificant for high school graduation (1.2 percentage points) and positive and significant for post-secondary enrollment (5.7 points). These are much smaller effects than those estimated in part (3), although the treatment effects for post-secondary enrollment remains sizable.

(c) Keeping in mind that an untreated observation may be matched multiple times, how many *unique* students not enrolled in Catholic school were used as matches in your analysis? (2 points)

Results shown in the log using the _treat and _weight variables created by psmatch2. There are 521 unique students who did not attend Catholic school who are used as matches.

9. Using the same propensity score model from part (7), estimate the ATE and ATET using inverse probability weighting (teffects ipw). How do your results differ from those in part (8)? How many students not enrolled in Catholic school were used in this analysis? (5 points)

Results are shown in the attached log and in the summary table below. The point estimates are in the same ballpark as those obtained from propensity score matching.

Below is a table containing the point estimates and standard errors (in parentheses) from different estimation strategies, for comparison:

	math12z	read12z	hsgrad	inpse	math	read
OLS	0.410*** (0.0431)	0.367*** (0.0432)	0.0644*** (0.0120)	0.170*** (0.0196)		
Exact matching (ATE)	0.314*** (0.0436)	0.305*** (0.0459)	0.0515*** (0.0111)	0.143*** (0.0189)	2.979*** (0.414)	2.888*** (0.435)
Exact matching (ATT)	0.283*** (0.0389)	0.259*** (0.0406)	0.0453*** (0.00737)	0.119*** (0.0156)	2.692*** (0.369)	2.459*** (0.384)
Propensity score match (ATT)	0.120** (0.0417)	-0.00310 (0.0437)	0.0118 (0.0109)	0.0574** (0.0211)		
IPW (ATT)	0.116*** (0.0234)	0.0380 (0.0279)	0.0210** (0.00713)	0.0637*** (0.0149)		

```
. // LPO-8852 Problem set 2 solutions
. // Last updated: September 15, 2021
  use https://stats.idre.ucla.edu/stat/stata/examples/methods_matter/chapter12/catholic, c
> lear
  // *******
  // Question 1
  // *******
 summ, sep(0)
   Variable |
                    Obs
                                       Std. Dev.
                               Mean
                                                      Min
                                                                Max
                                        2700654
         id |
                   5,671
                            4626664
                                                   124902
                                                             7979086
     read12 |
                   5,671
                           51.00126
                                       9.476733
                                                    29.15
                                                               68.09
     math12 |
                   5,671
                           51.05124
                                       9.502415
                                                    29.88
                                                               71.37
     hsgrad |
                   5,671
                            .9169459
                                       .2759884
                                                        0
                                                                  1
      inpse |
                   5,671
                           .7092224
                                                        0
                                                                  1
                                       .4541612
   catholic |
                   5,671
                           .1043908
                                       .3057938
                                                        0
                                                                  1
      read8 |
                   5,671
                           51.54138
                                       9.695829
                                                    32.05
                                                               70.55
                                                                77.2
      math8 |
                   5,671
                           51.48952
                                       9.683425
                                                    34.48
                   5,671
                                                                  1
     female |
                            .5200141
                                       .4996433
                                                        0
                   5,671
                                                                   5
       race |
                           3.532887
                                       .9537466
                                                        1
                   5,671
                                                        0
                                                                   1
      white |
                            .6892964
                                       .4628225
      black |
                   5,671
                           .0975137
                                       .2966821
                                                        0
                                                                  1
       hisp |
                   5,671
                            .1162053
                                       .3204992
                                                        0
                                                                  1
        api |
                   5,671
                           .0585435
                                       .2347889
                                                        0
                                                                   1
    nativam |
                   5,671
                                                        0
                            .0384412
                                       .1922758
                                                                   1
                   5,671
                                                        1
                                                                  6
    parmar8 |
                           5.344384
                                       1.576191
                                       2.217688
                   5,671
                                                                  12
    faminc8 |
                           9.526186
                                                        1
    fathed8 |
                   5,671
                           3.606948
                                       2.267043
                                                        1
                                                                  8
    mothed8 |
                   5,671
                           3.380356
                                                        1
                                                                  8
                                       2.141246
    fhowfar |
                   5,671
                           4.818198
                                       1.105028
                                                        1
                                                                  6
                                                                  6
    mhowfar |
                   5,671
                                                        1
                           4.858226
                                       1.074148
                                                        0
                                                                  2
     fight8 |
                   5,671
                           .2191853
                                       .5005381
      nohw8 |
                   5,671
                            .143361
                                       .3504715
                                                        0
                                                                  1
   disrupt8 |
                   5,671
                            .1795098
                                       .3838125
                                                        0
                                                                   1
  riskdrop8 |
                   5,671
                            .6236995
                                       .9031568
                                                        0
                                                                   5
 // *******
  // Question 2
  // *******
. // some re-coded variables for analysis \,
```

. tabulate faminc8
total annual |
family income |
in 8th grade |

in 8th grade	Freq.	Percent	Cum.
none	18	0.32	0.32
<\$1000	1 42	0.74	1.06
\$1000-\$2999	l 84	1.48	2.54
\$3000-\$4999	l 85	1.50	4.04
\$5000-\$7499	144	2.54	6.58
7500-\$9999	175	3.09	9.66
\$10000-\$14999	447	7.88	17.55
\$15000-\$19999	l 441	7.78	25.32
\$20000-\$24999	l 655	11.55	36.87
\$25000-\$34999	1,267	22.34	59.21
35000-\$49999	1,419	25.02	84.24
50000-\$74999	894	15.76	100.00
	+		

Total | 5,671 100.00

. tabulate faminc8, nolabel

total |

annual |

family |

income in |

8th grade	İ	Freq.	Percent	Cum.
1	-+- 	 18	0.32	0.32
2	1	42	0.74	1.06
3	1	84	1.48	2.54
4	1	85	1.50	4.04
5	1	144	2.54	6.58
6	1	175	3.09	9.66
7	1	447	7.88	17.55
8	1	441	7.78	25.32
9	1	655	11.55	36.87
10	1	1,267	22.34	59.21
11	1	1,419	25.02	84.24
12	1	894	15.76	100.00
Total	-+- 	5,671	100.00	

. gen faminc8b=0 if faminc8==1

(5,653 missing values generated)

. replace faminc8b = (0+1000)/2 if faminc8==2

(42 real changes made)

. replace faminc8b = (1000+2999)/2 if faminc8==3 (84 real changes made)

. replace faminc8b = (3000+4999)/2 if faminc8==4 (85 real changes made)

. replace faminc8b = (5000+7499)/2 if faminc8==5

(144 real changes made)

. replace faminc8b = (7500+9999)/2 if faminc8==6

(175 real changes made)

. replace faminc8b = (10000+14999)/2 if faminc8==7

(447 real changes made)

. replace faminc8b = (15000+19999)/2 if faminc8==8 (441 real changes made)

```
. replace faminc8b = (20000+24999)/2 if faminc8==9
(655 real changes made)
. replace faminc8b = (25000+34999)/2 if faminc8==10
(1,267 real changes made)
. replace faminc8b = (35000+49999)/2 if faminc8==11
(1,419 real changes made)
. replace faminc8b = (50000+74999)/2 if faminc8==12
(894 real changes made)
. label var faminc8b "family income in 8th grade (dollars)"
. gen faminc8c = 1 if faminc8<=7
(4,676 missing values generated)
. replace faminc8c = 2 if faminc8>=8 & faminc8<=10</pre>
(2,363 real changes made)
. replace faminc8c = 3 if faminc8>10 & faminc8~=.
(2,313 real changes made)
. label var faminc8c "family income in 8th grade (three categories)"
. codebook fathed8
fathed8
                                         father's highest level of education
______
                type: numeric (byte)
                label: farcat
                range: [1,8]
                                                  units: 1
        unique values: 8
                                               missing .: 0/5,671
           tabulation: Freq. Numeric Label
                        873 1 not finish hs
                       1,778
                                   2 hs grad
                         660
                                   3 junior coll
                         443 4 coll <4
743 5 coll gra
346 6 masters
                                   5 coll grad
                               7 doctorate
8 dont know
                         141
                         687
. gen fathed1 = fathed8==1 /* hs dropout */
. gen fathed2 = fathed8==2 /* hs grad */
. gen fathed3 = (fathed8>=3 & fathed8<=4) /* some college */
. gen fathed4 = (fathed8>=5 & fathed8<=8) /* 4yr college or more */
. label var fathed1 "father's highest ed: hs dropout"
. label var fathed2 "father's highest ed: hs grad"
. label var fathed3 "father's highest ed: some college"
. label var fathed4 "father's highest ed: 4yr college or more"
```

```
mothed8
                                         mother's highest level of education
______
               type: numeric (byte)
               label: farcat
               range: [1,8]
                                                 units: 1
        unique values: 8
                                              missing .: 0/5,671
           tabulation: Freq. Numeric Label
                       815 1 not finish hs
                       2,091
                                  2 hs grad
                         686
                                  3 junior coll
                                  4 coll <4
                         468
                                  5 coll grad
                         655
                         299
                                  6 masters
                                  7 doctorate
                          82
                         575
                                  8 dont know
. gen mothed1 = mothed8==1 /* hs dropout */
. gen mothed2 = mothed8==2 /* hs grad */
. gen mothed3 = (mothed8>=3 & mothed8<=4) /* some college */
. gen mothed4 = (mothed8>=5 & mothed8<=8) /* 4yr college or more */
. label var mothed1 "mother's highest ed: hs dropout"
. label var mothed2 "mother's highest ed: hs grad"
. label var mothed3 "mother's highest ed: some college"
. label var mothed4 "mother's highest ed: 4yr college or more"
. forvalues j=1/4 {
       replace fathed'j'=. if fathed8==.
 2.
       replace mothed'j'=. if mothed8==.
 4.
       }
(0 real changes made)
. egen pared8=rowmax(fathed8 mothed8)
. gen pared1 = pared8==1 /* hs dropout */
. gen pared2 = pared8==2 /* hs grad */
. gen pared3 = (pared8>=3 & pared8<=4) /* some college */
. gen pared4 = (pared8>=5 & pared8<=8) /* 4yr college or more */
. label var pared1 "parent's highest ed: hs dropout"
. label var pared2 "parent's highest ed: hs grad"
. label var pared3 "parent's highest ed: some college"
. label var pared4 "parent's highest ed: 4yr college or more"
```

```
parmar8
                                 parents marital status in 8th grade
.
------
             type: numeric (byte)
             label: parmarcat
             range: [1,6]
                                          units: 1
       unique values: 6
                                       missing .: 0/5,671
         tabulation: Freq. Numeric Label
                     543 1 divorced
                      77
                             2 widowed
                     140
                             3 separated
                      91
                             4 never married
                    93 5 not married but cohabit 4,727 6 married
. gen parmar1 = parmar8==1 | parmar8==3 /* divorced or separated */
. gen parmar2 = parmar8==2 | parmar8==4 | parmar8==5 /* widow never married or cohabit */
. gen parmar3 = parmar8==6
. label var parmar1 "parents divorced or separated"
. label var parmar2 "parents widowed never married or cohabit"
. label var parmar3 "parents married"
. codebook fhowfar mhowfar
                              how far in schl r's father wants r to go
fhowfar
______
              type: numeric (byte)
             label: farcat2
             range: [1,6]
                                          units: 1
       unique values: 6
                                       missing .: 0/5,671
         tabulation: Freq. Numeric Label
                      47
                           1 not finish hs
                     309
                              2 hs grad
                     384
                             3 junior coll
                             4 coll <4
                     595
                    2,889
                              5 coll grad
                    1,447 6 postsec ed
-----
                              how far in schl r^s mother wants r to go
mhowfar
             type: numeric (byte)
             label: farcat2
             range: [1,6]
                                          units: 1
                                       missing .: 0/5,671
       unique values: 6
         tabulation: Freq. Numeric Label
                     42
                           1 not finish hs
                     275
                              2 hs grad
                     361
                             3 junior coll
                             4 coll <4
                     577
                    2,928
                              5 coll grad
1,488 6 postsec ed . gen collegexp=(fhowfar>=5 & fhowfar<.) | (mhowfar>=5 & mhowfar<.)
. label var collegexp "=1 if mother or father expects college+"
```

```
. codebook riskdrop8
```

riskdrop8 # of risk factors for later dropout

```
type: numeric (byte)
               range: [0,5]
                                                 units: 1
        unique values: 6
                                            missing .: 0/5,671
           tabulation: Freq. Value
                       3,369 0
                       1,406 1
                        623 2
                         214 3
                         52 4
                          7 5
. gen riskdrop1=riskdrop8==1
. gen riskdrop2=(riskdrop8>=2 & riskdrop8<=5)</pre>
. for
each j in read12 math12 read8 math8 {
       egen 'j'z = std('j')
       label var 'j'z "standardized 'j'"
 3.
 4.
     }
```

. summ read12-female white-nativam fight8-collegexp if catholic==0, sep(0) Variable | Obs Mean Std. Dev. Min Max

Variable	l Obs	Mean	Std. Dev.	Min	Max
read12	5,079	50.63865	9.495149	29.15	68.09
math12	5,079	50.64465	9.534295	29.88	71.37
hsgrad	5,079	.9102185	.2858965	0	1
inpse	5,079	.6914747	.4619301	0	1
catholic	5,079	0	0	0	0
read8	5,079	51.11096	9.653501	32.05	70.55
math8	5,079	51.23648	9.747724	34.48	77.2
female	5,079	.5174247	.4997455	0	1
white	5,079	.6759205	.468076	0	1
black	5,079	.1019886	.3026631	0	1
hisp	5,079	.1193148	.3241905	0	1
api	5,079	.0626108	.2422854	0	1
${\tt nativam}$	5,079	.0401654	.1963663	0	1
fight8	5,079	.2234692	.5074014	0	2
nohw8	5,079	.1523922	.359436	0	1
disrupt8	5,079	.1815318	.3854961	0	1
riskdrop8	5,079	.6601693	.927189	0	5
faminc8b	5,079	31854.35	17282.99	0	62499.5
faminc8c	5,079	2.199646	.7320868	1	3
fathed1	5,079	.1636149	.3699622	0	1
fathed2	5,079	.3124631	.4635431	0	1
fathed3	5,079	.1925576	.3943473	0	1
fathed4	5,079	.3313644	.4707501	0	1
mothed1	5,079	.1543611	.3613301	0	1
mothed2	5,079	.3673952	.4821428	0	1
mothed3	5,079	.2012207	.4009521	0	1
mothed4	5,079	. 277023	.447572	0	1
pared8	5,079	4.036031	2.293045	1	8
pared1	5,079	.0874188	.2824756	0	1
pared2	5,079	. 2862768	.4520649	0	1
pared3	5,079	.2205158	.4146353	0	1
pared4	5,079	.4057885	.4910923	0	1
parmar1	5,079	.1265997	.3325568	0	1
parmar2	5,079	.0490254	.2159423	0	1
parmar3	5,079	.8243749	.3805384	0	1
collegexp	5,079	.7914944	.4062801	0	1

Variable	Obs	Mean	Std. Dev.	Min	Max
read12	592	54.11226	8.725462	29.86	68.09
math12	l 592	54.53951	8.463153	32.92	71.08
hsgrad	l 592	.9746622	.157282	0	1
inpse	592	.8614865	.3457303	0	1
catholic	592	1	0	1	1
read8	592	55.23412	9.271124	33.07	70.55
math8	592	53.66039	8.82731	35.53	77.2
female	592	.5422297	.4986348	0	1
white	592	.8040541	.3972628	0	1
black	l 592	.0591216	.2360516	0	1
hisp	l 592	.089527	.2857444	0	1
api	l 592	.0236486	.1520804	0	1
nativam	l 592	.0236486	.1520804	0	1
fight8	l 592	.1824324	.4359041	0	2
nohw8	l 592	.0658784	.2482792	0	1
disrupt8	l 592	.1621622	.3689112	0	1
riskdrop8	l 592	.3108108	.5741334	0	3
faminc8b	l 592	39534.13	16291.77	0	62499.5
faminc8c	l 592	2.513514	.6209701	1	3
fathed1	592	.0709459	.2569516	0	1
fathed2	592	.3226351	.4678798	0	1
fathed3	592	.2111486	.4084688	0	1
fathed4	592	.3952703	.4893221	0	1
mothed1	592	.0523649	.2229501	0	1
mothed2	592	.3800676	.4858136	0	1
mothed3	592	.222973	.4165923	0	1
mothed4	592	.3445946	.4756378	0	1
pared8	l 592	4.407095	2.0612	1	8
pared1	592	.0202703	.1410425	0	1
pared2	l 592	.2516892	.4343506	0	1
pared3	592	.2280405	.4199237	0	1
pared4	592	.5	.5004228	0	1
		0.075.07.0	0540440	^	

. // *********
. // Question 3
. // ********

parmar1 |

parmar2 |

parmar3 |

collegexp |

592

592

592

592

.0675676

.0202703

.9121622

.8851351

.2512146

.1410425

.2832983

.3191284

0

0

0

1

1

1

1

```
. foreach j in math12z read12z hsgrad inpse {
2. _eststo ols'j': reg 'j' catholic
   }
                    MS
                       Number of obs = 5,671
  Source |
90.48
   = 0.0000
 Residual | 5580.925 5,669 .984463751 R-squared = 0.0157
------ Adj R-squared = 0.0155
   Total | 5670.00001 5,670
                     1 Root MSE
______
  math12z |
        Coef. Std. Err. t P>|t| [95% Conf. Interval]
______
 catholic | .409881 .0430903 9.51 0.000 .3254075 .4943545

_cons | -.0427878 .0139223 -3.07 0.002 -.0700808 -.0154948
  Source | SS
               df
                   MS
                       Number of obs =
                                  5,671
----- F(1, 5669) = 72.13
 ------ Adj R-squared = 0.0124
  Total | 5670 5,670 .999999999 Root MSE = .99379
______
  read12z | Coef. Std. Err. t P>|t| [95% Conf. Interval]
______
                  8.49 0.000
 catholic | .3665415 .0431591
                           .2819331
                                 .4511499
  _cons | -.0382635 .0139445 -2.74 0.006 -.0656002 -.0109269
______
                       Number of obs =
  Source | SS
                   MS
               df
                                  5,671
= 0.0000
  Residual | 429.67959 5,669 .0757946 R-squared
                               = 0.0051
------ Adj R-squared = 0.0049
   Total | 431.881502 5,670 .076169577 Root MSE =
                                  .27531
______
  hsgrad |
        Coef. Std. Err. t P>|t| [95% Conf. Interval]
______
 catholic | .0644436 .0119564 5.39 0.000 .0410046

_cons | .9102185 .003863 235.62 0.000 .9026455
                                 .9177916
______
  Source |
                   MS
         SS
                       Number of obs =
            df
                                  5,671
75.27
 ------ Adj R-squared = 0.0129
   Total | 1169.50767 5,670 .206262376 Root MSE
______
        Coef. Std. Err. t P>|t| [95% Conf. Interval]
______
 catholic | .1700118 .0195958 8.68 0.000 .1315964 .2084271 
_cons | .6914747 .0063313 109.21 0.000 .6790629 .7038865
. // *******
. // Question 5
```

. // *******

```
. foreach j in math12 read12 math12z read12z hsgrad inpse {
 2. display in red "Outcome: 'j'"
 3. _eststo em'j'ate: teffects nnmatch ('j' i.faminc8c) (catholic), ematch(i.faminc8c)
> ate
 4. _eststo em'j'att: teffects nnmatch ('j' i.faminc8c) (catholic), ematch(i.faminc8c)
> atet
 5. }
Outcome: math12
                                     Number of obs = 5,671
Treatment-effects estimation
Estimator : nearest-neighbor matching Matches: requested =
Outcome model : matching
                                                           40
                                              min =
                                                 max = 2155
Distance metric: Mahalanobis
                    AI Robust
   math12 | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATE |
   catholic |
(yes vs no) | 2.979077 .4139935 7.20 0.000 2.167664
______
Treatment-effects estimation
                                     Number of obs =
                                                        5,671
Estimator : nearest-neighbor matching Matches: requested =
Outcome model : matching
                                                 min =
                                                           40
Distance metric: Mahalanobis
                                                 max =
                                                          2155
                    AI Robust
  math12 | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATET |
  catholic |
(yes vs no) | 2.692406 .3694927 7.29 0.000 1.968214 3.416599
Outcome: read12
                                    Number of obs = 5,671
Treatment-effects estimation
Estimator : nearest-neighbor matching Matches: requested =
                                                           40
Outcome model : matching
                                             min =
                                                 max = 2155
Distance metric: Mahalanobis
               AI Robust
        read12 | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATE |
   catholic |
(yes vs no) | 2.887636 .4354489 6.63 0.000 2.034172
Treatment-effects estimation
                                     Number of obs =
Estimator : nearest-neighbor matching Matches: requested =
                                                           40
Outcome model : matching
                                               min =
Distance metric: Mahalanobis
                                                         2155
                                                 max =
               AI Robust
    read12 | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATET
  catholic |
(yes vs no) | 2.459126 .3844986 6.40 0.000 1.705523 3.21273
Outcome: math12z
```

= 5,671

Number of obs

Treatment-effects estimation

math12z	Coef.	AI Robust Std. Err.	z	P> z	[95% Conf.	Interval]
ATE catholic	+ 					
(yes vs no)	•	.0435672	7.20	0.000	.2281172	.3988974
Treatment-effe					obs =	-
Estimator		eighbor match	ning	Matches:		
Outcome model	_					40
Distance metri	ıc: Mahalanobi	.s 			max =	2155
		AI Robust				
	Coef.	Std. Err.			[95% Conf.	Interval]
ATET	 					
catholic						
(yes vs no)	.2833391	.0388841	7.29	0.000	.2071277	.3595505
Outcome: readi						5 054
Treatment-effe			ing		obs =	
Estimator Outcome model		signoor matti	ıııß	Macches.		40
Distance metri	_	.S				2155
		AI Robust				
	Coef.					Interval]
ATE						
catholic		0450402	6 63	0 000	0146401	2047660
(yes vs no)	.304708	.0459495	0.03	0.000	.2146491	.3947009
Treatment-effe					obs =	
Estimator						
Outcome model	_				min =	
Distance metri	ic: Mahalanobi	.S			max =	2155
	 	AI Robust				
,	Coef.		z	P> z	[95% Conf.	Interval]
	·					
ATET						
catholic		0405700	6 40	0.000	170000	2200404
(yes vs no)	.2594909					.3390124
Outcome: hsgra	ad					
Treatment-effe	ects estimatio	on		Number of	obs =	5,671
Estimator		eighbor match	ning	Matches:	requested =	1
Outcome model	•					40
Distance metri	ic: Mahalanobi 					2155
		AI Robust				
hsgrad	Coef.	Std. Err.				
ATE						
catholic (yes vs no)	•	.0110895	4.65	0.000	.0298054	.0732754
T				N		
Treatment-effe Estimator			ning		obs = requested =	

hsgrad	Coef.	Std. Err.	z	P> z	[95% Conf.	<pre>Interval]</pre>
ATET						
catholic						
(yes vs no)	.0452857	.0073688	6.15	0.000	.0308431	.0597283
Outcome: inpse						
Treatment-effe		on		Number of	f obs =	5.671
Estimator						-
Outcome model		6	6		min =	
Distance metri	•	is			max =	
		AI Robust			F0=0/ # 0	
inpse		Std. Err.	Z	P> z	[95% Conf.	Interval
ATE						
catholic						
(yes vs no)		.0189111	7.56	0.000	.105941	.1800712
Treatment-effe	cts estimation	on		Number of	f obs =	5,671
Estimator	: nearest-ne	eighbor match	ing	Matches:	requested =	1
Outcome model	: matching				min =	40
Distance metri	c: Mahalanob	is			max =	2155
		AT Dahara				
innse l		AI Robust	7	DSIzl	[95% Conf.	Intervall
+						
ATET						
catholic						
(yes vs no)	.1193482	.0155742	7.66	0.000	.0888234	.1498731
•						
. // *******	**					
. // Question	6					
. // ******						
. teffects nnm			(cathol			
Treatment-effe					f obs =	•
Estimator		eighbor match	ing	Matches:		
Outcome model	•					40
Distance metri		is 			max =	2155
		AI Robust				
math12			z	P> z	[95% Conf.	Interval]
ATE						
catholic						
(yes vs no)	2.979077	.4139935	7.20	0.000	2.167664	3.790489

. tebalance summarize faminc8b i.faminc8c math8z note: refitting the model using the generate() option Covariate balance summary

			Raw 	Matched
	Numbe	r of obs =	5,671	11,342
	Treat	ed obs =	592	5,671
	Contr	ol obs =	5,079	5,671
	Standardized	differences	Vari	ance ratio
	Raw	Matched	Raw	Matched
faminc8b	'	.0489535		
faminc8c	I			
2	1500581	-1.35e-15	.9343974	1
3	.3942688	3.28e-15	1.026989	1
	1			
math8z	.2606656	.181422	.8200688	.8230263

```
. // *********
. // Question 7
. // *******
```

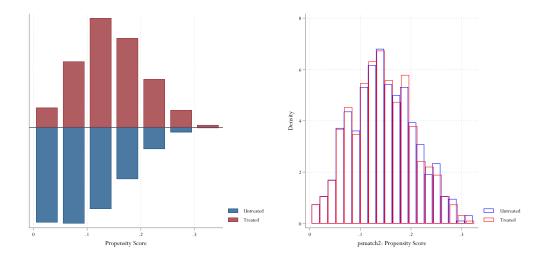
. // uses psmatch2 (default is probit model, 1 nn match). I use the "ties" option . // to line up with teffects psmatch (used later). If there are multiple nearest . // neighbors with the same pscore, all of them will be matched.

. psmatch2 catholic c.math8z##c.math8z c.read8z##c.read8z female black hisp api /// nati > vam riskdrop1 riskdrop2 disrupt nohw8 faminc8b fathed2-fathed4 /// mothed2-mothed4 par

> mar1 parmar2 Probit regress		ties		Number	of obs =	5,671
Flobic legiess	51011			LR chi2		270.89
				Prob >		0.0000
Log likelihood	l = -1762.210	2		Pseudo		0.0714
		- 				
catholic	Coef.	Std. Err.	z	P> z	[95% Conf.	Interval]
math8z	0334847	.0375223	-0.89	0.372	107027	.0400576
c.math8z#						
c.math8z	0852282	.0251197	-3.39	0.001	134462	0359944
 read8z 	.1766125	.0345038	5.12	0.000	.1089862	. 2442387
c.read8z#						
c.read8z 		.025047	0.51	0.607	0362045	.0619777
female		.050002	0.41	0.678	0772715	.1187328
black	124419	.0969725	-1.28	0.199	3144816	.0656435
hisp	0186785	.0820854	-0.23	0.820	179563	.1422059
api	554642	.129953	-4.27	0.000	8093452	2999388
nativam	2416373	.1417836	-1.70	0.088	519528	.0362535
riskdrop1	0060125	.0633419	-0.09	0.924	1301603	.1181352
riskdrop2	2225268	.1159476	-1.92	0.055	4497798	.0047262
disrupt8	.0966356	.0675578	1.43	0.153	0357753	.2290466
nohw8	3036905	.0848934	-3.58	0.000	4700786	1373025
faminc8b	7.40e-06	1.66e-06	4.47	0.000	4.15e-06	.0000106
fathed2	.1835301	.0932529	1.97	0.049	.0007578	.3663024
fathed3	.1129928	.101119	1.12	0.264	0851969	.3111824
fathed4	.1260127	.0986007	1.28	0.201	0672412	.3192665
mothed2	.2766157	.1005447	2.75	0.006	.0795518	.4736796
mothed3	.2626728	.1087212	2.42	0.016	.0495831	.4757624
mothed4	.3072885	.107914	2.85	0.004	.095781	.5187959
parmar1	1165539	.0997515	-1.17	0.243	3120631	.0789554
parmar2	11798	.1497204	-0.79	0.431	4114266	.1754665
collegexp	.1585568	.0706054	2.25	0.025	.0201728	.2969408
aana l	_1 000060	1/01610	_12 EE	0 000	_O 17/EO	_1 605150

_cons | -1.899869 .1401612 -13.55 0.000 -2.17458 -1.625159

Variable	İ	Mea					est	V(T)/	
		Treated C	Control						
	+				-+-			+	
math8z			.18838						
c.math8z#c.math8z			.88282				0.963	1.05	
read8z	- 1	.38086					0.723	1.02	
c.read8z#c.read8z			1.0254				0.616	1.00	
female	- 1	.54223				-1.05	0.293	Ι.	
black	- 1	.05912	.0625				0.808	Ι.	
hisp	1	.08953	.11149				0.209		
api	- 1	.02365	.01689	3.3		0.82	0.410	Ι.	
nativam	- 1	.02365	.03041			-0.72	0.474	Ι.	
riskdrop1	- 1	.21115	.23311	-5.2		-0.91	0.364	Ι.	
riskdrop2	1	.0473	.04561	0.6		0.14	0.890	١.	
disrupt8	- 1	.16216	.15203	2.7		0.48	0.632	١.	
nohw8	1	.06588	.06926	-1.1		-0.23	0.817	Ι.	
faminc8b	1	39534	39587		1	-0.06	0.955	0.97	
fathed2		.32264	.31588				0.803	Ι.	
fathed3			.21622				0.832		
fathed4		.39527	.40878				0.636		
mothed2		.38007	.36486				0.589		
mothed3	i	.22297	.22973				0.781		
mothed4	i	.34459							
parmar1		.06757							
parmar2		.02027							
collegexp		.88514							
* if variance ratio Ps R2 LR chi2 p									
0.006 9.67							1	0	
* if B>25%, R outsi. tabulate _treated psmatch2: Treatment psm assignment	de [0. _weig atch2:	5; 2] ht weight of 2	matched	l contro 3	ls	4	Total		
Untreated							521		
Treated									
Total 1,	048	60				+			
<pre>. count if _treated 521 psgraph, name(psg</pre>									
<pre>. psgraph, name(psg, replace) twoway (histogram _pscore if catholic==0 [fweight = _weight], /// bin(20) fcolor(none) > lcolor(blue)) (histogram _pscore if catholic==1, /// bin(20) fcolor(none) lcolor(red) > , legend(label(1 Untreated) label(2 Treated)) /// name(psg2, replace) graph combine psg psg2, row(1) xsize(8) ysize(4) name(q7, replace) . graph export q7.pdf, name(q7) as(pdf) replace (file q7.pdf written in PDF format)</pre>									



```
. // now show tebalance and overlap after teffects psmatch
. capture drop _*
. capture drop nn*
. quietly teffects psmatch (math12z) (catholic c.math8z##c.math8z c.read8z##c.read8z ///
> female black hisp api nativam riskdrop1 riskdrop2 disrupt nohw8 faminc8b /// fathed2-f
> athed4 mothed2-mothed4 parmar1 parmar2 collegexp, probit), atet gen(nn)
```

. tebalance summarize math8z read8z female black hisp api nativam riskdrop1 /// riskdrop
> 2 disrupt nohw8 faminc8b fathed2-fathed4 mothed2-mothed4 parmar1 /// parmar2 collegexp
Covariate balance summary

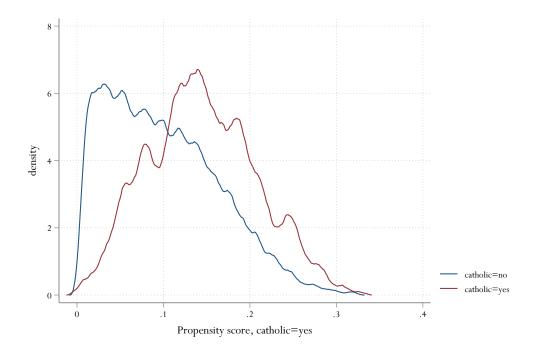
Matched

		Numbe:	r of obs =	5,671	1,184
		Treat	ed obs =	592	592
		Contr	ol obs =	5,079	592
	IS	tandardized	differences	Varia	nce ratio
		Raw	Matched	Raw	Matched
math8z		.2606656	.0390663	.8200688	.9790608
read8z		.4356571	.0206151	.9223486	1.019996
female		.0496905	0611929	.9955598	1.014271
black		1579425	0141249	.6082682	.9493548
hisp		0974819	073021	.7768816	.8228771
api		1926191	.0479196	.3939964	1.390378
nativam		0940456	0416405	.5998088	.7831978
riskdrop1		0974069	052802	.8844729	.931734
riskdrop2		4043775	.0080193	.3184886	1.035202
disrupt8		0513382	.0278314	.9158064	1.053918
nohw8		2800718	0134485	.4771307	.9546722
faminc8b		.4572743	0032443	.8885848	.9738385
fathed2		.0218417	.0144817	1.018799	1.011303
fathed3		.0463075	0123526	1.072902	.9828765
fathed4		.1331022	0275406	1.08046	.9890437
mothed2		.0261837	.0314228	1.015285	1.016733
mothed3		.0532039	0161333	1.079537	.9791022
mothed4		.1463164	021237	1.129346	.9866866
parmar1		2003089	0514933	.5706348	.8455882
1		157667	0546141	.4266036	.7120205
collegexp		.2563305	0661787	.6169925	1.1871

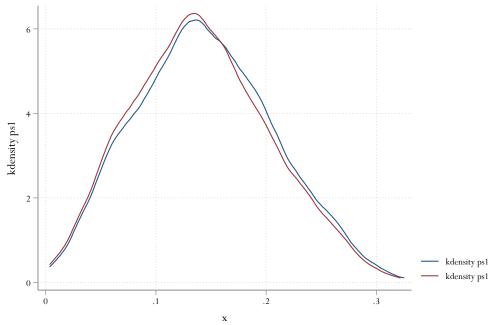
[.] teffects overlap, ptlevel(1)

[.] graph export q7b.pdf, as(pdf) replace

⁽file q7b.pdf written in PDF format)



```
. predict ps0 ps1, ps
. gen ob=_n
. preserve
. tempfile treated
. keep if catholic==1
(5,079 observations deleted)
. keep nn1
. bysort nn1: gen weight=_N
. by nn1: keep if _n==1
(71 observations deleted)
. rename nn1 ob
. save 'treated'
file \ C:\Users\corcorsp\AppData\Local\Temp\ST\_42dcc\_000002.tmp \ saved
. restore
. merge m:1 ob using 'treated'
(note: variable ob was float, now double to accommodate using data's values)
    Result
                                      # of obs.
    not matched
                                         5,150
        from master
                                         5,150 (_merge==1)
        from using
                                           0 (_merge==2)
                                           521
                                               (_merge==3)
. replace weight=1 if catholic==1
(592 real changes made)
```



```
.
.
. // *********
. // Question 8
. // ********
.
. foreach j in math12z read12z hsgrad inpse {
    2.     display in red "Outcome: 'j'"
    3.     _eststo psm'j': teffects psmatch ('j') (catholic c.math8z##c.math8z c.read8z##c.re
> ad8z ///     female black hisp api nativam riskdrop1 riskdrop2 disrupt nohw8 faminc8b /
> // fathed2-fathed4 mothed2-mothed4 parmar1 parmar2 collegexp, probit), atet
    4.
```

```
psmatch2 catholic c.math8z##c.math8z c.read8z##c.read8z female black hisp ///
> i nativam riskdrop1 riskdrop2 disrupt nohw8 faminc8b fathed2-fathed4 ///
                                                                mothed
> 2-mothed4 parmar1 parmar2 collegexp, outcome('j')
Outcome: math12z
Treatment-effects estimation
                                   Number of obs =
                                                      5,671
Estimator : propensity-score matching
                                   Matches: requested =
Outcome model : matching
                                               min =
Treatment model: probit
                                               max =
______
        AI Robust
   math12z | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATET |
  catholic |
(yes vs no) | .1204992 .0417063 2.89 0.004 .0387563 .2022421
______
                                    Number of obs =
                                                     5,671
Probit regression
                                                     270.89
                                    LR chi2(23)
                                                 =
                                    Prob > chi2
                                                = 0.0000
Log likelihood = -1762.2102
                                   Pseudo R2 = 0.0714
              Coef. Std. Err. z
                                    P>|z| [95% Conf. Interval]
   catholic |
______
    math8z | -.0334847 .0375223 -0.89 0.372
                                            -.107027
                                                     .0400576
  c.math8z#|
   c.math8z | -.0852282 .0251197 -3.39 0.001 -.134462 -.0359944
    read8z | .1766125 .0345038 5.12 0.000 .1089862 .2442387
   c.read8z#|
   c.read8z |
            .0128866 .025047
                            0.51
                                   0.607
                                          -.0362045
                                                     .0619777
                    .050002
                                           -.0772715 .1187328
    female | .0207307
                              0.41
                                   0.678
     black | -.124419 .0969725
                              -1.28 0.199 -.3144816 .0656435
     hisp | -.0186785 .0820854 -0.23 0.820 -.179563 .1422059
                    .129953
                              -4.27 0.000 -.8093452 -.2999388
      api | -.554642
   nativam | -.2416373
                              -1.70 0.088 -.519528 .0362535
                     . 1417836
                                           -.1301603 .1181352
  riskdrop1 | -.0060125
                     .0633419
                              -0.09
                                   0.924
  riskdrop2 | -.2225268
                     .1159476 -1.92 0.055
                                          -.4497798 .0047262
                              1.43
                                           -.0357753 .2290466
   disrupt8 |
                     .0675578
           .0966356
                                    0.153
                     .0848934
                              -3.58 0.000
     nohw8 | -.3036905
                                           -.4700786 -.1373025
   faminc8b | 7.40e-06
                     1.66e-06
                             4.47
                                    0.000
                                           4.15e-06 .0000106
                                           .0007578
                                                   .3663024
   fathed2 |
           .1835301
                     .0932529
                              1.97
                                   0.049
   fathed3 |
           .1129928
                    .101119
                              1.12 0.264
                                           -.0851969 .3111824
                            1.28
                                           -.0672412 .3192665
   fathed4 | .1260127
                                   0.201
                     .0986007
                     .1005447 2.75 0.006
   mothed2 | .2766157
                                          .0795518 .4736796
   mothed3 | .2626728
                    .1087212
                             2.42 0.016
                                           .0495831 .4757624
                    .107914
   mothed4 |
           .3072885
                              2.85
                                   0.004
                                            .095781 .5187959
   parmar1 | -.1165539 .0997515 -1.17 0.243 -.3120631 .0789554
                     .1497204 -0.79 0.431
                                           -.4114266 .1754665
   parmar2 |
            -.11798
  collegexp | .1585568
                     .0706054
                             2.25 0.025 .0201728
                                                   . 2969408
     _cons | -1.899869
                     .1401612 -13.55
                                    0.000
                                            -2.17458 -1.625159
                                 Controls Difference
               Sample |
                       Treated
      Variable
                                                         S.E.
______
      math12z Unmatched | .367093217 -.04278779 .409881007 .043090321
```

ap

9.51

psmatch2:						
Treatment	On suppor	Total				
	+					
Untreated	5,079	5,079				
Treated	592	592				
	+					
	5,671	5,671				
Outcome: rea		:		Numban	ef aba	F 671
	fects estimat:				of obs = : requested =	-
	l : matching	-	curing	nacches	min =	
Treatment mo	_				max =	_
		AI Robust				
read12z					[95% Conf.	Interval]
	-+					
ATET catholic	1					
	•	0437156	-0 07	0 943	0887824	0825794
(yes vs no)						
Probit regre	ssion			Number	of obs =	5,671
<u> </u>				LR chi2	2(23) =	270.89
					chi2 =	
Log likeliho	od = -1762.210	02		Pseudo	R2 =	0.0714
					[95% Conf.	Interval]
	-+				107027	0400576
matrioz	0334647	.0375223	-0.09	0.372	107027	.0400376
c.math8z	#					
c.math8z	0852282	.0251197	-3.39	0.001	134462	0359944
	1					
read8z	1 .1766125	.0345038	5.12	0.000	.1089862	. 2442387
c.read8z		005047	0 54	0 607	0260045	0040777
c.read8z	.0128866	.025047	0.51	0.607	0362045	.0619777
female	1 .0207307	.050002	0.41	0.678	0772715	.1187328
black		.0969725	-1.28	0.199	3144816	.0656435
hisp		.0820854	-0.23	0.820	179563	.1422059
api		.129953	-4.27	0.000	8093452	2999388
nativam		.1417836	-1.70	0.088	519528	.0362535
riskdrop1	0060125	.0633419	-0.09	0.924	1301603	.1181352
riskdrop2		.1159476	-1.92	0.055	4497798	.0047262
disrupt8		.0675578	1.43	0.153	0357753	.2290466
nohw8	3036905	.0848934	-3.58	0.000	4700786	1373025
faminc8b	7.40e-06	1.66e-06	4.47	0.000	4.15e-06	.0000106
fathed2	1 .1835301	.0932529	1.97	0.049	.0007578	.3663024
fathed3	1 .1129928	.101119	1.12	0.264	0851969	.3111824
fathed4		.0986007	1.28	0.201	0672412	.3192665
mothed2	.2766157	.1005447	2.75	0.006	.0795518	.4736796
mothed3		.1087212	2.42	0.016	.0495831	.4757624
mothed4		.107914	2.85	0.004	.095781	.5187959
parmar1			-1.17	0.243	3120631	.0789554
parmar2			-0.79	0.431	4114266	. 1754665
collegexp			2.25	0.025	.0201728	.2969408
_cons	-1.899869	.1401612	-13.55	0.000	-2.17458	-1.625159

ATT | .328277963 .331379442 -.003101478 .055027185 -0.06

 Note: S.E. doe 1	es not take in	nto account	that the	propensi	ty score is e	stimated.		
psmatch2:	Common							
Treatment	support							
assignment (
Untreated								
	592 +							
	5,671							
Outcome: hsgra	ad							
reatment-eff					f obs =			
	stimator : propensity-score matching			Matches:	1			
utcome model : matching				min =				
reatment mode	el: probit 				max =	1		
		AI Robust						
	Coef.				[95% Conf.	Interval]		
TET	•							
catholic		a.a			0000			
(yes vs no)	.0118243 			0.280	0096059 	.0332545		
Probit regress	sion				of obs =			
					(23) =			
					chi2 =			
•	d = -1762.210				R2 =			
catholic	Coef.	Std. Err.	z	P> z	[95% Conf.	Interval]		
	•				107027			
c.math8z#								
c.math8z	0852282	.0251197	-3.39	0.001	134462	0359944		
read8z	. 1766125	.0345038	5.12	0.000	.1089862	. 2442387		
c.read8z#								
c.read8z		.025047	0.51	0.607	0362045	.0619777		
female	.0207307	.050002	0.41	0.678	0772715	.1187328		
black	124419	.0969725	-1.28	0.199	3144816	.0656435		
hisp	0186785	.0820854	-0.23	0.820	179563	.1422059		
api	554642	.129953	-4.27	0.000	8093452	2999388		
nativam	2416373	.1417836	-1.70	0.088	519528	.0362535		
riskdrop1	0060125	.0633419	-0.09	0.924	1301603	.1181352		
riskdrop2	2225268	.1159476	-1.92	0.055	4497798	.0047262		
disrupt8	.0966356	.0675578	1.43	0.153	0357753	.2290466		
nohw8	3036905	.0848934	-3.58	0.000	4700786	1373025		
faminc8b	7.40e-06	1.66e-06	4.47	0.000	4.15e-06	.0000106		
fathed2	.1835301	.0932529	1.97	0.049	.0007578	.3663024		
fathed3	.1129928	.101119	1.12	0.264	0851969	.3111824		
fathed4	.1260127	.0986007	1.28	0.201	0672412	.3192665		
mothed2	.2766157	.1005447	2.75	0.006	.0795518	.4736796		
mothed3	.2626728	.1087212	2.42	0.016	.0495831	.4757624		
mothed4	.3072885	.107914	2.85	0.004	.095781	.5187959		
parmar1	- 1165539	.0997515	-1.17	0.243	3120631	.0789554		

parmar1 | -.1165539 .0997515 -1.17 0.243 -.3120631 .0789554

					Difference		T-stat
					.064443615		5.39
		ГТ .9746621 +			.011824324	.010813283	1.09
					ity score is e	estimated.	
psmatch2:	_						
Treatment	support						
-	On suppor						
•							
	5,079						
	592 						
Total	5,671	5,671					
Outcome: inps							
	ects estimat				of obs =		
		ty-score matc	hing	Matches	: requested =		
	: matching				min =		
Treatment mod	el: probit				max =	1	
	1	AI Robust					
inpse	•		z	P> z	[95% Conf	. Interval]	
	+						
ATET	1						
catholic							
(yes vs no)	.0574324	.0210931	2.72	0.006	.0160907	.0987742	
Probit regres	sion			Number	of obs =	5.671	
					2(23) =		
				Prob >	chi2 =		
Log likelihoo	d = -1762.210	02		Pseudo	R2 =	0.0714	
catholic	Coef.	Std. Err.	z	P> z	[95% Conf	. Interval]	
math8z	0334847	.0375223	-0.89	0.372	107027	.0400576	
c.math8z#	 						
c.math8z		.0251197	-3.39	0.001	134462	0359944	
0 1 0		7020110.	0.00	0.002			
read8z	1 .1766125	.0345038	5.12	0.000	.1089862	.2442387	
	1						
c.read8z#	•	205245	0.54	0 007	0000015	0.04.0000	
c.read8z	.0128866	.025047	0.51	0.607	0362045	.0619777	
female	.0207307	.050002	0.41	0.678	0772715	.1187328	
black		.0969725	-1.28	0.199	3144816	.0656435	
hisp		.0820854	-0.23	0.820	179563	.1422059	
api	554642	.129953	-4.27	0.000	8093452	2999388	
nativam	2416373	.1417836	-1.70	0.088	519528	.0362535	
riskdrop1	0060125	.0633419	-0.09	0.924	1301603	.1181352	
riskdrop2	2225268	.1159476	-1.92	0.055	4497798	.0047262	
disrupt8		.0675578	1.43	0.153	0357753	.2290466	
nohw8	3036905	.0848934	-3.58	0.000	4700786	1373025	
faminc8b	7.40e-06	1.66e-06	4.47	0.000	4.15e-06	.0000106	
fathed2	1 .1835301	.0932529	1.97	0.049	.0007578	.3663024	
fathed3	1 .1129928	.101119	1.12	0.264	0851969	.3111824	
	1 1060107	0006007	1 00			2100665	

.1260127

fathed4 |

.0986007

1.28

0.201 -.0672412

.3192665

```
parmar2 | -.11798 .1497204 -0.79 0.431 -.4114266 .1754665
  collegexp | .1585568 .0706054 2.25 0.025 .0201728 .2969408
    _cons | -1.899869 .1401612 -13.55 0.000 -2.17458 -1.625159
 _____
Variable Sample | Treated Controls Difference S.E. T-stat
      ATT | .861486486 .804054054 .057432432 .023261301
______
Note: S.E. does not take into account that the propensity score is estimated.
      | psmatch2:
psmatch2: | Common
Treatment | support
assignment | On suppor | Total
-----
Untreated | 5,079 | 5,079
Treated | 592 | 592
-----
  Total | 5,671 | 5,671
. table _treated _weight, row col
psmatch2: |
Treatment | psmatch2: weight of matched
assignmen | controls t | 1 2 3 4 Total
Untreated | 456 60 4 1 521
 Treated | 592
                           592
   1
  Total | 1,048 60 4 1 1,113
. count if _treated==0 & _weight~=.
 521
. // *******
. // Question 9
. // *******
```

```
. foreach j in math12z read12z hsgrad inpse {
 2. display in red "Outcome: 'j'"
3. _eststo ipw'j': teffects ipw ('j') (catholic c.math8z##c.math8z c.read8z##c.read8z
> /// female black hisp api nativam riskdrop1 riskdrop2 disrupt nohw8 faminc8b ///
> fathed2-fathed4 mothed2-mothed4 parmar1 parmar2 collegexp, probit), atet
Outcome: math12z
Iteration 0: EE criterion = 3.162e-24
Iteration 1: EE criterion = 4.938e-33
Treatment-effects estimation
                                       Number of obs = 5.671
Estimator : inverse-probability weights
Outcome model : weighted mean
Treatment model: probit
______
                       Robust
   math12z | Coef. Std. Err. z P>|z| [95% Conf. Interval]
catholic |
(yes vs no) | .1156927 .0233943 4.95 0.000 .0698407 .1615447
  catholic |
      no | .2514005 .0316755 7.94 0.000 .1893178 .3134833
Outcome: read12z
Iteration 0: EE criterion = 3.162e-24
Iteration 1: EE criterion = 8.060e-34
                                       Number of obs = 5,671
Treatment-effects estimation
Estimator : inverse-probability weights
Outcome model : weighted mean
Treatment model: probit
         Robust
    read12z | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATET
   catholic |
(yes vs no) | .0379625 .02795 1.36 0.174 -.0168185 .0927435
POmean |
   catholic |
    no | .2903155 .0292765 9.92 0.000 .2329346 .3476964
Outcome: hsgrad
Iteration 0: EE criterion = 3.162e-24
Iteration 1: EE criterion = 4.743e-34
                                       Number of obs = 5,671
Treatment-effects estimation
Estimator : inverse-probability weights
Outcome model : weighted mean
Treatment model: probit
                       Robust
   hsgrad | Coef. Std. Err. z P>|z| [95% Conf. Interval]
ATET
   catholic |
```

(yes vs no) | .0210462 .0071286 2.95 0.003 .0070743 .0350181

Outcome: inpse

Iteration 0: EE criterion = 3.162e-24
Iteration 1: EE criterion = 2.194e-33

Treatment-effects estimation Number of obs = 5,671

Estimator : inverse-probability weights

Outcome model : weighted mean

Treatment model: probit

inpse			z	P> z		Interval]
ATET catholic (yes vs no)	.0637249		4.26	0.000	.0344404	. 0930095
POmean catholic no	.7977616	.0089675	88.96	0.000	.7801857	.8153374

[.] esttab _all using PS2estimates.csv, se paren csv replace

(note: file PS2estimates.csv not found)
(output written to PS2estimates.csv)

[.] capture log close