

13. Propensity Score Matching Technique

The basic idea behind propensity score matching (PSM) is to match each participant with an identical nonparticipant and then measure the average difference in the outcome variable between the participants and the nonparticipants. This exercise illustrates how to implement PSM in the Stata program.

The estimation command in Stata is “pscore.ado,” developed by Becker and Ichino (2002). The “pscore” command estimates the propensity score, which is the probability of getting a treatment for each household, and tests the balancing property—that is, observations with the same propensity score must have the same distribution of observable characteristics independent of treatment status. After balancing is done, different commands can be used to carry out different types of matching and then derive the average treatment effect.

Propensity Score Equation: Satisfying the Balancing Property

The first step in PSM is to determine the propensity score and satisfy the balancing property. It is done using the “pscore” command in Stata. Use the 1998 data, hh_98.dta. Start with the male program participation variable “dmmfd” as the treatment variable. The following command shows the application of the “pscore” command:

```
pscore dmmfd sexhead agehead educhead lnland vaccess pcirr rice  
wheat milk oil egg [pw=weight], pscore(ps98) blockid(blockf1)  
comsup level(0.001);
```

The results include probit regression output, the estimation and description of the propensity scores, the number of blocks and stratification using propensity scores, and the balancing property test. The area of common support is those propensity scores within the range of the lowest and highest estimated values for households in the treatment group.

The following output shows that the identified region of common support is [.00180123, .50022341], the final number of blocks is 4, and the balancing property is not satisfied. The most important element to look for in the output is the list of variables that cause the balancing property not to be satisfied. The output shows the “egg” variable is not balanced in block 2. The solution to this problem is to use a different set of covariates and rerun the “pscore” command.

 Algorithm to estimate the propensity score

The treatment is dmmfd

HH has male microcredit participant : 1=Y, 0=N	Freq.	Percent	Cum.
0	909	80.51	80.51
1	220	19.49	100.00
Total	1,129	100.00	

Estimation of the propensity score

(sum of wgt is 1.1260e+03)
 Iteration 0: log pseudolikelihood = -424.61883
 Iteration 1: log pseudolikelihood = -390.85321
 Iteration 2: log pseudolikelihood = -389.10243
 Iteration 3: log pseudolikelihood = -389.05511
 Iteration 4: log pseudolikelihood = -389.05501

Probit estimates	Number of obs	=	1129
	Wald chi2(11)	=	64.36
	Prob > chi2	=	0.0000
Log pseudolikelihood = -389.05501	Pseudo R2	=	0.0838

dmmfd	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]
sexhead	.915108	.2432905	3.76	0.000	.4382675 1.391949
agehead	-.0036952	.0046186	-0.80	0.424	-.0127475 .005357
educhead	.0161662	.0170125	0.95	0.342	-.0171777 .04951
lnland	-.3341691	.1113146	-3.00	0.003	-.5523417 -.1159965
vaccess	-.0752904	.1770457	-0.43	0.671	-.4222935 .2717128
pcirr	.2088394	.1753383	1.19	0.234	-.1348174 .5524961
rice	.145771	.0384417	3.79	0.000	.0704268 .2211153
wheat	.0465751	.0648087	0.72	0.472	-.0804475 .1735977
milk	-.0017358	.023861	-0.07	0.942	-.0485026 .045031
oil	-.0249797	.0135856	-1.84	0.066	-.051607 .0016476
egg	-.7687454	.2311995	-3.33	0.001	-1.221888 -.3156028
_cons	-1.188481	.8358266	-1.42	0.155	-2.826671 .4497088

Note: the common support option has been selected
 The region of common support is [.00180123, .50022341]

Description of the estimated propensity score
 in region of common support

Estimated propensity score

Percentiles		Smallest		
1%	.0055359	.0018012	Obs	1127
5%	.0170022	.0020871	Sum of Wgt.	1127
10%	.0346036	.0026732	Mean	.1339801
25%	.069733	.0028227	Std. Dev.	.0850809
50%	.1206795		Variance	.0072388
		Largest	Skewness	.8931864
75%	.1811405	.4698302	Kurtosis	3.942122
90%	.2527064	.472444		
95%	.2965199	.4735467		
99%	.3903884	.5002234		

 Step 1: Identification of the optimal number of blocks
 Use option detail if you want more detailed output

The final number of blocks is 4

This number of blocks ensures that the mean propensity score is not different for treated and controls in each blocks

 Step 2: Test of balancing property of the propensity score
 Use option detail if you want more detailed output

Variable egg is not balanced in block 2

The balancing property is not satisfied

Try a different specification of the propensity score

Inferior of block of pscore	HH has male microcredit participant: 1=Y, 0=N		Total
	0	1	
0	380	49	429
.1	382	97	479
.2	140	70	210
.4	5	4	9
Total	907	220	1,127

Note: the common support option has been selected

 End of the algorithm to estimate the pscore

After a few iterations, you will find that dropping “egg” and “Inland” allows the “pscore” command to be rerun with the balancing property satisfied. So “pscore” on

“dfmfd” is run again, this time excluding the “egg” and “Inland” variables. Before rerunning the “pscore” command, it is important to drop the “ps98” and “blockf1” variables that were created as a result of the earlier run. Because female program participation is of more interest, the “pscore” command is shown here with female participation only.

```
pscore dfmfd sexhead agehead educhead lnland vaccess pcirr rice
wheat milk oil egg [pw=weight], pscore(ps98) blockid(blockf1)
comsup level(0.001);
```

This time the balancing property is satisfied, as shown here:

```
*****
Algorithm to estimate the propensity score
*****
```

The treatment is dfmfd

HH has female microcredit participant : 1=Y, 0=N	Freq.	Percent	Cum.
0	534	47.30	47.30
1	595	52.70	100.00
Total	1,129	100.00	

Estimation of the propensity score

```
(sum of wgt is 1.1260e+03)
Iteration 0: log pseudolikelihood = -750.38718
Iteration 1: log pseudolikelihood = -682.82636
Iteration 2: log pseudolikelihood = -680.63459
Iteration 3: log pseudolikelihood = -680.62452
Iteration 4: log pseudolikelihood = -680.62452
```

```
Probit estimates                               Number of obs   =   1129
                                                Wald chi2(11)   =   85.21
                                                Prob > chi2     =   0.0000
Log pseudolikelihood = -680.62452           Pseudo R2      =   0.0930
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
dmmfd						
sexhead	-.037986	.1662857	-0.23	0.819	-.3639	.287928
agehead	.0013931	.0037305	0.37	0.709	-.0059185	.0087047
educhead	-.0465567	.0151559	-3.07	0.002	-.0762618	-.0168516
lnland	-.6662184	.101586	-6.56	0.000	-.8653232	-.4671136
vaccess	-.1173796	.13358	-0.88	0.380	-.3791916	.1444323
pcirr	.4304416	.154365	2.79	0.005	.1278917	.7329915

rice		.0571981	.0307982	1.86	0.063	-.0031652	.1175615
wheat		-.0055393	.056959	-0.10	0.923	-.1171769	.1060982
milk		.015395	.0184184	0.84	0.403	-.0207044	.0514944
oil		.0235048	.01239	1.90	0.058	-.000779	.0477887
egg		-.1114687	.1647319	-0.68	0.499	-.4343373	.2113999
_cons		-1.483823	.7367316	-2.01	0.044	-2.927791	-.0398558

Note: the common support option has been selected
 The region of common support is [.02576077, .71555996]

-----OUTPUT OMITTED-----

 Step 2: Test of balancing property of the propensity score
 Use option detail if you want more detailed output

The balancing property is satisfied

-----OUTPUT OMITTED-----

 End of the algorithm to estimate the pscore

With the propensity scores generated, the outcomes of interest (such as total per capita expenditure) between the treatment group and the matched control group are now compared to see whether the microcredit programs affect the outcome of interest. The following sections estimate the treatment effect of microcredit program participation, using different matching techniques that are available.

Average Treatment Effect Using Nearest-Neighbor Matching

The command to estimate the average treatment effect on the treated group using nearest-neighbor matching is “attnd.” Following is the application of the “attnd” command to estimate the average treatment effect of female participation in microcredit programs on per capita total expenditure using nearest-neighbor matching:

```
attnd lexptot dfmfd [pweight=weight], pscore(ps98) comsup;
```

Estimating “attnd” with or without weights does not affect the results. Just for the purpose of this exercise, “attnd” was shown with weights estimation.

As the following output shows, female microcredit participation does have a significant impact on household per capita expenditure with the nearest-neighborhood matching method ($t = 3.256$). The average treatment of the treated (ATT) on per capita expenditure for female program participation is 13.6 percent.

ATT estimation with Nearest Neighbor Matching method
(random draw version)
Analytical standard errors

n. treat.	n. contr.	ATT	Std. Err.	t
595	293	0.136	0.042	3.256

Note: the numbers of treated and controls refer to actual nearest-neighbor matches

Average Treatment Effect Using Stratification Matching

The “atts” command calculates the average treatment effect on the treated using stratification matching. To estimate the average treatment effect of female participation on the treated for per capita total expenditure, use the following:

```
atts lexptot dfmfd, pscore(ps98) blockid(blockf1) comsup
```

The result that follows shows a 9.9 percent increase in per capita expenditure because of women’s participation in the microcredit programs. The impact is significant at the 5 percent level ($t = 3.320$).

ATT estimation with the Stratification method
Analytical standard errors

n. treat.	n. contr.	ATT	Std. Err.	t
595	529	0.099	0.030	3.320

Average Treatment Effect Using Radius Matching

The “attr” command calculates the average treatment effect on the treated using radius matching. Following is a demonstration:

```
. attr lexptot dfmfd, pscore(ps98) radius(0.001) comsup
```

The result shows an increased impact (14.6 percent) with high significance ($t = 3.793$) of women’s microcredit participation on per capita expenditure:

ATT estimation with the Radius Matching method
Analytical standard errors

n. treat.	n. contr.	ATT	Std. Err.	t
478	386	0.146	0.039	3.793

Note: the numbers of treated and controls refer to actual matches within radius

Further Exercises

Do the same exercise using male participation (“dmmfd”). Discuss your results.

Reference

Becker, Sascha, and Andrea Ichino. 2002. “Estimation of Average Treatment Effects Based on Propensity Scores.” *Stata Journal* 2 (4): 358–77.