

Problem Set 7¹
Due 11/01

This exercise examines the following research question: What is the effect of maternal smoking during pregnancy on infant birth weight and death? The required reading for this problem set is Almond, Chay, and Lee (2005). The data extract, “lbw.dta”, is from the 1989 Linked National Natality-Mortality Details Files, which are an annual census of births in the U.S., derived from Certificates of Live Birth. Information on subsequent infant death within a year of birth is derived from Death Certificates. This extract consists of all births in Pennsylvania in 1989. The observational unit of the data is the mother-infant outcome match. There are 139,149 observations and 32 variables. For this problem set, observations with missing values for any of the variables below were dropped from the original sample (about 17%).

The key variables are:

dbirwt birth weight of the infant (in grams)
death indicator equal to one if the infant died within one year of birth and zero, otherwise
tobacco indicator equal to one if the mother smoked during pregnancy and zero, otherwise

The relevant control variables are:

- Mother’s attributes:

dage (mother’s age), dmeduc (mother’s educational attainment), mblack (indicator=1 if mother is Black), motherr (=1 if neither Black nor White), mhispan (=1 if Hispanic), dmar (=1 if mother is unmarried), foreignb (=1 if mother is foreign born)

- Father’s attributes:

dage (father’s age), dfeduc (father’s educational attainment), fblack (indicator=1 if father is Black), fotherr (=1 if neither Black nor White), fhispan (=1 if Hispanic)

- Other risky behavior:

alcohol (indicator=1 if mother drank alcohol during pregnancy), drink (number of drinks per week)

- Medical care:

tripre1, tripre2, tripre3 (indicators=1 if 1st prenatal care visit in 1st, 2nd, or 3rd trimester, respectively), tripre0 (=1 if no prenatal care visits), nprevist (total number of prenatal care visits)

- Pregnancy history and maternal health:

first (=1 if first-born), ddivord (birth order), deadkids (number previous births where newborn died), disllb (months since last birth), preterm (=1 if previous birth premature or small for gestational age), pre4000 (=1 if previously had > 4000 gram newborn), plural (=1 if twins or greater birth), phyper (=1 if mother had pregnancy associated hypertension), diabete (=1 if mother diabetic), anemia (=1 if mother anemic)

- a. Under what conditions can one identify the average treatment effect of maternal smoking by comparing the unadjusted difference in mean birth weight of infants of smoking and non-smoking mothers? Under the assumption that maternal smoking is randomly assigned, estimate its impact on birth weight. Provide some evidence for or against the hypothesis that maternal smoking is randomly assigned.

¹This problem set requires that you download a user-written command “pscore” and its associated files. First, type in “net search pscore” in *Stata*, and then click the link “st0026_2 from <http://www.stata-journal.com/software/sj5-3>” shown on the *Stata* result window.

b. Suppose that maternal smoking is randomly assigned conditional on the other observable determinants of infant birth weight.

(1) What does this imply about the relationship between maternal smoking and unobservable determinants of birth weight conditional on the observables?

(2) Use a basic linear regression model to estimate the impact of smoking and report your estimates.

- *Stata* code: `reg dbirwt tobacco dimage dmeduc mblack motherr mhispan dmar foreignb dfage dfeduc fblack fotherr fhispan alcohol drink trip1 trip2 trip3 trip0 nprevist first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia, vce(robust)`

(3) Under what conditions is the average treatment effect identified?

c. Under the assumption of random assignment conditional on the observables:

(1) What are the sources of misspecification bias in the estimates generated by the linear model estimated in part (b)?

(2) Use an approach in the spirit of multivariate matching, that is, estimate the smoking effects using a flexible functional form for the control variables (e.g., higher order terms and interactions).

- *Stata* code: `reg dbirwt tobacco dimage dmeduc mblack motherr mhispan dmar foreignb dfage dfeduc fblack fotherr fhispan alcohol drink trip1 trip2 trip3 trip0 nprevist first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia dimage2 dimage3 dmeduc2 dfage2 dfage3 dfeduc2 dimage_dmeduc dimage_mblack dimage_motherr dimage_mhispan dimage_dmar dfage_dfeduc dfage_fblack dfage_fotherr dfage_fhispan dimage_alcohol dimage_drink, vce(robust)`

(3) What are the benefits and drawbacks to this approach?

d. Describe the propensity score approach to the problem of estimating the average treatment effect of smoking when the treatment is randomly assigned conditional on the observables. How does it reduce the dimensionality problem of multivariate matching?

e. Implement the propensity score approach (hints below) to the evaluation problem using two methods:

(1) Method 1: control directly for the estimated propensity scores in a regression model.

(2) Method 2: use the estimated propensity score in a classification scheme to stratify the sample.

(3) Provide empirical evidence that your implementation is reasonable and evidence on the overlap of the observables of smokers and non-smokers. Present your findings and interpret the results. (This is an open-ended question, so show me what you know and be creative and thoughtful.)

- Below is my *Stata* code for estimating the propensity score without specifying the number of blocks at the beginning. *Stata* generated 33 blocks in the end. For balancing results, please see the attached log file.

```
pscore tobacco dimage dmeduc mblack motherr mhispan dmar foreignb dfage dfeduc fblack fotherr
fhispan alcohol drink trip1 trip2 trip3 trip0 nprevist first ddivord deadkids disllb preterm
pre4000 plural phyper diabete anemia dimage2 dmeduc2 dfage2 dfeduc2 dimage_dmeduc dimage_mblack
dimage_motherr dimage_mhispan dimage_dmar dfage_dfeduc dfage_fblack dfage_fotherr dfage_fhispan
dimage_alcohol dimage_drink, pscore(pscore) blockid(block) logit level(0.01)
```

- Below is my Stata code for estimating the propensity score with 100 blocks.

```
pscore tobacco dimage dmeduc mblack motherr mhispan dmar foreignb dfage dfeduc fblack fotherr
fhispan alcohol drink tripre1 tripre2 tripre3 tripre0 nprevist first ddivord deadkids disllb preterm
pre4000 plural phyper diabete anemia dimage2 dmeduc2 dfage2 dfeduc2 dimage_dmeduc dimage_mblack
dimage_motherr dimage_mhispan dimage_dmar dfage_dfeduc dfage_fblack dfage_fotherr dfage_fhispan
dimage_alcohol dimage_drink, pscore(pscore100) blockid(block100) logit level(0.001) numblo(101)
```

- Below is my Stata code for estimating the propensity score with 200 blocks.

```
pscore tobacco dimage dmeduc mblack motherr mhispan dmar foreignb dfage dfeduc fblack fotherr
fhispan alcohol drink tripre1 tripre2 tripre3 tripre0 nprevist first ddivord deadkids disllb preterm
pre4000 plural phyper diabete anemia dimage2 dmeduc2 dfage2 dfeduc2 dimage_dmeduc dimage_mblack
dimage_motherr dimage_mhispan dimage_dmar dfage_dfeduc dfage_fblack dfage_fotherr dfage_fhispan
dimage_alcohol dimage_drink, pscore(pscore200) blockid(block200) logit level(0.005) numblo(201)
```

- f. Use the estimated propensity scores to reweigh the outcomes of non-smokers and estimate the average treatment effect.
 - (1) Compare the estimates to those in part (e) and interpret your findings.
 - (2) What are the benefits and drawbacks of approaches that use the estimated propensity scores as weights?
- g. A more informative way to describe the birth weight effects of smoking is to estimate the nonparametric conditional mean of birth weight as a function of the estimated propensity score, separately for smokers and non-smokers.
 - (1) To do this, simply stratify the smokers into 100 equal-sized cells based on their propensity scores and calculate the mean birth weight and propensity score in each cell. Do the same for the non-smokers. Plot these two conditional mean functions, with the mean scores on the x -axis and mean birth weight on the y -axis.
 - (2) Interpret your findings and relate them to the results in part (e) and part (f).
 - (3) Redo the above using 200 equal-sized cells for smokers and non-smokers.
- h. Low birth weights (less than 2500 grams) are considered particularly undesirable since they comprise a large share of infant deaths).
 - (1) Redo part (g) using an indicator for low birth weight as the outcome of interest.
 - (2) Interpret your findings.
- i. Estimate the impact of maternal smoking on infant death using the methods in parts (a), (b), and (g), using the 100 equal-sized cells for smokers and non-smokers).
 - (1) Interpret your findings.
 - (2) From your results, what might you conclude about the relationship between birth weight and infant death?
- j. Concisely and coherently summarize your results above. Describe the estimated effects of maternal smoking on infant-birth weight and infant mortality, and whether you think your “best” estimate of the effects of smoking is credibly identified. State why or why not.

References

Almond, D., K. Y. Chay, and D. Lee (2005). “The Costs of Low Birth Weight.” *Quarterly Journal of Economics* 120(3): 1031-1083.

```

. /* ***** Trial versions *****
> pscore tobacco
>     dimage dmeduc mblack motherr mhispan dmar foreignb
>     dfage dfeduc fblack fotherr fhispan
>     alcohol drink triprel tripre2 tripre3 tripre0 nprevist
>     first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia
>     ,pscore(p_score) blockid(block_id) detail logit comsup;
>
> pscore tobacco
>     dimage dmeduc mblack motherr mhispan dmar foreignb
>     dfage dfeduc fblack fotherr fhispan
>     alcohol drink triprel tripre2 tripre3 tripre0 nprevist
>     first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia
>     ,pscore(p_score) blockid(block_id) logit;
>
> pscore tobacco
>     dimage dmeduc mblack motherr mhispan dmar foreignb
>     dfage dfeduc fblack fotherr fhispan
>     alcohol drink triprel tripre2 tripre3 tripre0 nprevist
>     first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia
>     ,pscore(p_score) blockid(block_id) logit comsup;
>
> pscore tobacco
>     dimage dmeduc mblack motherr mhispan dmar foreignb
>     dfage dfeduc fblack fotherr fhispan
>     alcohol drink triprel tripre2 tripre3 tripre0 nprevist
>     first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia
>     ,pscore(p_score) blockid(block_id) logit numblo(30);
>
> pscore tobacco
>     dimage dmeduc mblack motherr mhispan dmar foreignb
>     dfage dfeduc fblack fotherr fhispan
>     triprel tripre2 tripre3 tripre0 nprevist
>     first ddivord deadkids disllb preterm pre4000 plural phyper diabete anemia
>     ,pscore(p_score) blockid(block_id) logit level(0.001);
> ***** */
>
> gen dimage2=dimage^2;

. gen dmeduc2 = dmeduc^2;

. gen dfage2 = dfage^2;

. gen dfeduc2 = dfeduc^2;

. gen dimage_dmeduc = dimage*dmeduc;

. gen dimage_mblack = dimage*mblack;

. gen dimage_motherr = dimage*motherr;

. gen dimage_mhispan = dimage*mhispan;

```

```

. gen dimage_dmar =dimage*dmar;

. gen dfage_dfeduc = dfage*dfeduc;

. gen dfage_fblack = dfage*fblack;

. gen dfage_fotherr = dfage*fotherr;

. gen dfage_fhispan = dfage*fhispan;

. gen dimage_alcohol = dimage*alcohol;

. gen dimage_drink = dimage*drink;

. pscore tobacco
>      dimage dmeduc mblack motherr mhispan dmar foreignb
>      dfage dfeduc fblack fotherr fhispan
>      alcohol drink triprel tripre2 tripre3 tripre0 nprevist
>      first dlivord deadkids disl1b preterm pre4000 plural phyper diabete anemia
>      dimage2 dmeduc2 dfage2 dfeduc2
>      dimage_dmeduc dimage_mblack dimage_motherr dimage_mhispan dimage_dmar
>      dfage_dfeduc dfage_fblack dfage_fotherr dfage_fhispan
>      dimage_alcohol dimage_drink
>      ,pscore(pscore) blockid(block) logit level(0.01);

```

Algorithm to estimate the propensity score

The treatment is tobacco

indicator=1 if the mother smoked during pregnancy and zero, otherwise	Freq.	Percent	Cum.
0	112,782	81.29	81.29
1	25,957	18.71	100.00
Total	138,739	100.00	

Estimation of the propensity score

```

note: tripre3 dropped due to collinearity
Iteration 0:  log likelihood = -66869.313
Iteration 1:  log likelihood = -58462.913
Iteration 2:  log likelihood = -57422.224
Iteration 3:  log likelihood = -57356.981
Iteration 4:  log likelihood = -57356.052
Iteration 5:  log likelihood = -57356.051

```

Logit estimates

Log likelihood = -57356.051

```

Number of obs   =    138739
LR chi2(43)     =   19026.52
Prob > chi2     =    0.0000
Pseudo R2      =    0.1423

```

tobacco	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
dmage	-.1068329	.0168385	-6.34	0.000	-.1398358	-.07383
dmeduc	.3566266	.0372407	9.58	0.000	.2836362	.429617
mblack	-1.634605	.1438595	-11.36	0.000	-1.916565	-1.352646
motherr	-1.16326	.5555444	-2.09	0.036	-2.252107	-.0744127
mhispan	-1.872902	.2752192	-6.81	0.000	-2.412321	-1.333482
dmarr	-.450019	.0907707	-4.96	0.000	-.6279262	-.2721118
foreignb	-.6493657	.0593651	-10.94	0.000	-.7657192	-.5330122
dfage	.0493512	.0113155	4.36	0.000	.0271732	.0715293
dfeduc	.2903256	.0305091	9.52	0.000	.2305288	.3501224
fblack	-.6170809	.120269	-5.13	0.000	-.8528038	-.381358
fotherr	-.3703641	.5542399	-0.67	0.504	-1.456654	.7159261
fhispan	-.6086838	.2169956	-2.81	0.005	-1.033987	-.1833803
alcohol	3.189513	.3753923	8.50	0.000	2.453757	3.925268
drink	-.1376371	.0893637	-1.54	0.124	-.3127867	.0375125
triprel	-.1029287	.0424955	-2.42	0.015	-.1862183	-.0196391
tripre2	.0777792	.0421629	1.84	0.065	-.0048584	.1604169
tripre0	.3218592	.0694966	4.63	0.000	.1856484	.45807
nprevist	-.0174786	.0025198	-6.94	0.000	-.0224174	-.0125399
first	-.0132039	.0278816	-0.47	0.636	-.0678509	.0414431
dlivord	.048795	.0100513	4.85	0.000	.0290949	.0684952
deadkids	.1839926	.0098299	18.72	0.000	.1647264	.2032589
disllb	.0040391	.0003119	12.95	0.000	.0034278	.0046503
preterm	.5363455	.051284	10.46	0.000	.4358308	.6368602
pre4000	-.7662002	.0872055	-8.79	0.000	-.9371198	-.5952806
plural	-.1351962	.064253	-2.10	0.035	-.2611297	-.0092627
phyper	-.4634038	.0510504	-9.08	0.000	-.5634608	-.3633468
diabete	.0742374	.0604983	1.23	0.220	-.044337	.1928118
anemia	.3448649	.0611455	5.64	0.000	.225022	.4647079
dmage2	-.0026272	.0002931	-8.96	0.000	-.0032016	-.0020528
dmeduc2	-.0386059	.0017458	-22.11	0.000	-.0420276	-.0351843
dfage2	-.0007918	.000132	-6.00	0.000	-.0010504	-.0005331
dfeduc2	-.0188084	.0011074	-16.98	0.000	-.0209789	-.016638
dmage_dmeduc	.0155625	.0010838	14.36	0.000	.0134383	.0176868
dmage_mblack	.0408797	.0054963	7.44	0.000	.030107	.0516523
dmage_moth~r	.0335603	.0204456	1.64	0.101	-.0065124	.073633
dmage_mhis~n	.0472308	.0113564	4.16	0.000	.0249727	.0694888
dmage_dmar	.0574987	.003609	15.93	0.000	.0504252	.0645722
dfage_dfeduc	.0019303	.0006574	2.94	0.003	.0006418	.0032188
dfage_fblack	.0182842	.0038912	4.70	0.000	.0106576	.0259109
dfage_foth~r	-.0131404	.0184187	-0.71	0.476	-.0492405	.0229596
dfage_fhis~n	.010875	.0078056	1.39	0.164	-.0044237	.0261736
dmage_alco~l	-.0552314	.013497	-4.09	0.000	-.0816849	-.0287778
dmage_drink	.0074777	.0033175	2.25	0.024	.0009756	.0139798
_cons	-2.444141	.3329251	-7.34	0.000	-3.096662	-1.79162

Description of the estimated propensity score

Estimated propensity score			
Percentiles	Smallest		
1%	.0111771		
5%	.0219342		
10%	.0310177	Obs	138739
25%	.0725522	Sum of Wgt.	138739
50%	.1643964	Mean	.1870923
		Std. Dev.	.146644
75%	.2479226		
90%	.3873616	Variance	.0215045
95%	.4789602	Skewness	1.406159
99%	.6690482	Kurtosis	5.724969

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

The final number of blocks is 33

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 dmeduc is not balanced in block 1
Variable mblack is not balanced in block 1
Variable dmar is not balanced in block 1
Variable dfeduc is not balanced in block 1
Variable fblack is not balanced in block 1
Variable fotherr is not balanced in block 1
Variable tripre0 is not balanced in block 1
Variable dmeduc2 is not balanced in block 1
Variable dfeduc2 is not balanced in block 1
Variable dimage_dmeduc is not balanced in block 1
Variable dimage_mblack is not balanced in block 1
Variable dimage_dmar is not balanced in block 1
Variable dfage_dfeduc is not balanced in block 1
Variable dfage_fblack is not balanced in block 1
Variable dfage_fotherr is not balanced in block 1
Variable mblack is not balanced in block 2
Variable fblack is not balanced in block 2
Variable alcohol is not balanced in block 2
Variable drink is not balanced in block 2
Variable tripre0 is not balanced in block 2
Variable first is not balanced in block 2
Variable dimage2 is not balanced in block 2
Variable dfage2 is not balanced in block 2
Variable dimage_mblack is not balanced in block 2
```


Variable dimage_alcohol is not balanced in block 2

Variable dimage_drink is not balanced in block 2

Variable mblack is not balanced in block 3

Variable fblack is not balanced in block 3

Variable first is not balanced in block 3

Variable dlivord is not balanced in block 3

Variable disllb is not balanced in block 3

Variable dimage_mblack is not balanced in block 3

Variable dfage_fblack is not balanced in block 3

Variable dfeduc is not balanced in block 4

Variable fblack is not balanced in block 4

Variable dfeduc2 is not balanced in block 4

Variable dimage_mblack is not balanced in block 4

Variable dfage_fblack is not balanced in block 4

Variable dimage_mblack is not balanced in block 5

Variable dfage_fblack is not balanced in block 5

Variable dfeduc is not balanced in block 7

Variable dfeduc2 is not balanced in block 7

Variable dimage is not balanced in block 8

Variable deadkids is not balanced in block 8

Variable dimage2 is not balanced in block 8

Variable dimage_dmeduc is not balanced in block 8

Variable dimage is not balanced in block 9

Variable dfage is not balanced in block 9

Variable disllb is not balanced in block 9

Variable dimage2 is not balanced in block 9

Variable dfage2 is not balanced in block 9

Variable dimage_dmeduc is not balanced in block 9

Variable dimage is not balanced in block 10

Variable dfage is not balanced in block 10

Variable dimage2 is not balanced in block 10

Variable dfage2 is not balanced in block 10

Variable dimage_dmeduc is not balanced in block 10

Variable dfage_dfeduc is not balanced in block 10

Variable fotherr is not balanced in block 12

Variable fhispan is not balanced in block 12

Variable drink is not balanced in block 12

Variable dfage_fotherr is not balanced in block 12

Variable dfage_fhispan is not balanced in block 12

Variable dmage_drink is not balanced in block 12

Variable anemia is not balanced in block 13

Variable drink is not balanced in block 14

Variable first is not balanced in block 14

Variable dmage_drink is not balanced in block 14

Variable dfage_fblack is not balanced in block 16

Variable dfeduc is not balanced in block 17

Variable dmeduc is not balanced in block 18

Variable dfeduc is not balanced in block 18

Variable first is not balanced in block 18

Variable dmeduc2 is not balanced in block 18

Variable dfeduc2 is not balanced in block 18

Variable dfage_dfeduc is not balanced in block 18

Variable motherr is not balanced in block 19

Variable dfeduc is not balanced in block 19

Variable nprevist is not balanced in block 19

Variable dfeduc2 is not balanced in block 19

Variable dmage_motherr is not balanced in block 19

Variable dfage_dfeduc is not balanced in block 19

Variable dfeduc is not balanced in block 20

Variable tripre2 is not balanced in block 20

Variable dfeduc2 is not balanced in block 20

Variable dmeduc is not balanced in block 21

Variable deadkids is not balanced in block 21

Variable pre4000 is not balanced in block 21

Variable dmeduc2 is not balanced in block 21

Variable triprel is not balanced in block 23

Variable dfeduc is not balanced in block 24

Variable deadkids is not balanced in block 24

Variable fblack is not balanced in block 27

Variable dimage_mblack is not balanced in block 27

Variable dfage_fblack is not balanced in block 27

Variable dimage is not balanced in block 28

Variable dmeduc is not balanced in block 28

Variable mblack is not balanced in block 28

Variable dfage is not balanced in block 28

Variable disllb is not balanced in block 28

Variable dimage2 is not balanced in block 28

Variable dmeduc2 is not balanced in block 28

Variable dfage2 is not balanced in block 28

Variable dimage_dmeduc is not balanced in block 28

Variable dimage_mblack is not balanced in block 28

Variable dfage_dfeduc is not balanced in block 28

Variable dfage_fblack is not balanced in block 28

Variable dmeduc is not balanced in block 29

Variable dfage is not balanced in block 29

Variable triprel is not balanced in block 29

Variable disllb is not balanced in block 29

Variable dmeduc2 is not balanced in block 29

Variable dfage2 is not balanced in block 29

Variable dimage_dmeduc is not balanced in block 29

Variable dfage_dfeduc is not balanced in block 29

Variable dimage is not balanced in block 30

Variable dmeduc is not balanced in block 30

Variable deadkids is not balanced in block 30

Variable disllb is not balanced in block 30

Variable dimage2 is not balanced in block 30

Variable dmeduc2 is not balanced in block 30

Variable dimage_dmeduc is not balanced in block 30

Variable deadkids is not balanced in block 31

Variable anemia is not balanced in block 32

The balancing property is not satisfied

Try a different specification of the propensity score

Inferior of block of pscore	indicator=1 if the mother smoked during pregnancy and zero, otherwise		Total
	0	1	
0	8,916	192	9,108
.025	9,168	303	9,471
.0375	6,373	297	6,670
.05	5,119	303	5,422
.0625	4,611	354	4,965
.075	4,260	387	4,647
.0875	4,118	446	4,564
.1	8,406	1,052	9,458
.125	8,209	1,273	9,482
.15	4,132	692	4,824
.1625	4,545	855	5,400
.175	5,028	1,035	6,063
.1875	4,752	1,132	5,884
.2	8,512	2,164	10,676
.225	6,116	1,898	8,014
.25	2,260	751	3,011
.2625	1,774	744	2,518
.275	2,961	1,267	4,228
.3	4,468	2,154	6,622
.35	1,699	974	2,673
.375	1,413	941	2,354
.4	1,199	886	2,085
.425	536	397	933
.4375	144	96	240
.440625	119	123	242
.44375	211	221	432
.45	1,507	1,470	2,977
.5	997	1,138	2,135
.55	540	765	1,305
.6	442	741	1,183
.7	145	357	502
.8	84	381	465
.9	18	168	186
Total	112,782	25,957	138,739

End of the algorithm to estimate the pscore

```
. sort block pscore;

. drop dimage2 dmeduc2 dfage2 dfeduc2
>      dimage_dmeduc dimage_mblack dimage_motherr dimage_mhispan dimage_dmar
>      dfage_dfeduc dfage_fblack dfage_fotherr dfage_fhispan
>      dimage_alcohol dimage_drink;

. save "$dr\files\data_pscore.dta",replace;
```