

Propensity Scores II

Marcelo Coca Perrillon
HSR - 2012

Reminders

- Read Gelman & Hill's chapter 10
- Really, read Gelman & Hill...
- Homework will be based on Dehejia and Wahba (1999). You'll replicate some of the analyses in their paper

HELP data

```
. bysort homeless: sum age female il pcs drugrisk
```

```
-> homeless = 0
```

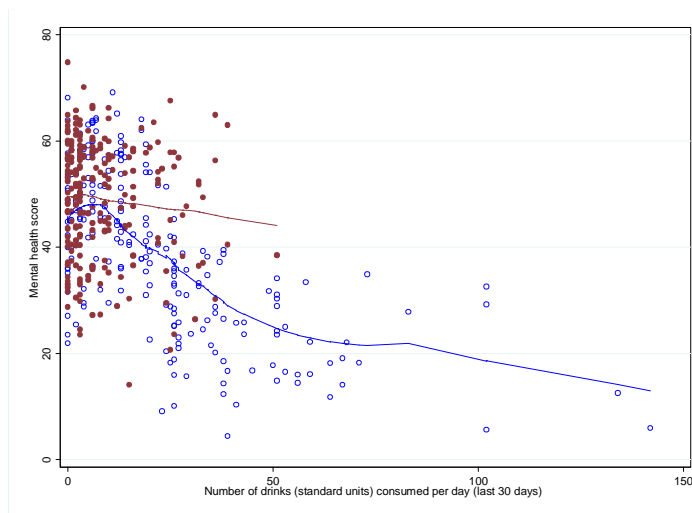
Variable	Obs	Mean	Std. Dev.	Min	Max
age	244	35.04098	7.165759	21	58
female	244	.2745902	.4472249	0	1
il	244	8.061475	9.743221	0	51
pcs	244	49.00083	10.82878	14.07429	74.80633
drugrisk	243	1.728395	3.975168	0	21

```
-> homeless = 1
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	209	36.36842	8.260958	19	60
female	209	.1913876	.3943379	0	1
il	209	23.03828	23.47315	0	142
pcs	209	38.42003	15.07664	4.435177	69.17161
drugrisk	209	2.07177	4.725098	0	21

Note: some people prefer to use the standardized mean difference instead. Always check distributions.

Lowess



Estimate the propensity score

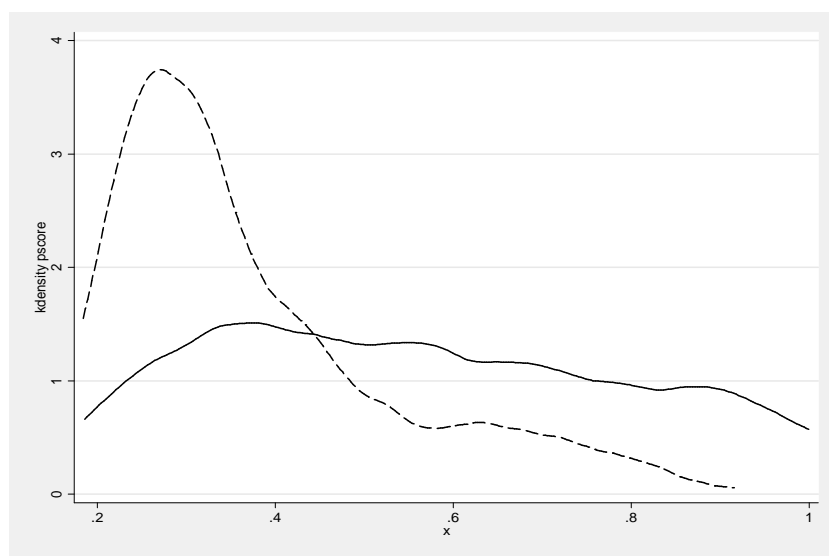
```
. logit homeless age female il drugrisk, nolog
```

```
Logistic regression      Number of obs   =      452
                        LR chi2(4)           =      95.70
                        Prob > chi2          =      0.0000
Log likelihood = -264.17304      Pseudo R2       =      0.1534
```

homeless	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
age	.0048899	.0138751	0.35	0.725	-.0223048 .0320846
female	-.3696801	.2572309	-1.44	0.151	-.8738435 .1344833
il	.0677485	.0089214	7.59	0.000	.0502629 .085234
drugrisk	.0483531	.0244174	1.98	0.048	.0004959 .0962103
_cons	-1.247658	.5101496	-2.45	0.014	-2.247532 -.2477826

```
predict pscore
```

Important: the outcome is NOT in the regression. We want to mimic randomization.



What to do with the propensity score?

Typical uses:

1. Stratification
2. Weighting
3. Matching
4. ~~Adding the propensity score as a covariate~~

Stratification

```
xtile quint5 = pscore, nq(5)
reg pcs homeless age female il drugrisk
est sto model_all
forvalues i= 1(1)5 {
  reg pcs homeless age female il drugrisk if quint5 == `i'
  est sto model_q`i'
}
est table model_all model_q1 model_q2 model_q3 model_q4 model_q5, star b($7.2f)
```

Variable	model_all	model_q1	model_q2	model_q3	model_q4	model_q5
homeless	-5.38***	-1.82	-1.46	-0.13	-5.58*	-18.37***
age	-0.19**	-0.26	-0.23	-0.09	-0.25	-0.29*
female	-5.62***	-3.77	-5.33	-1.99	-7.40	-4.88
il	-0.36***	0.96	0.20	-0.59	-0.85*	-0.12*
drugrisk	-0.33**	0.21	-0.50	-0.42	-1.01*	-0.03
_cons	60.83***	57.38***	60.25***	59.50***	74.42***	60.88***

legend: * p<0.05; ** p<0.01; *** p<0.001

Are the strata comparable?

```
. bysort homeless: sum age female i1 drugrisk if quint5 ==1
```

```
-> homeless = 0
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	70	34.6	7.645023	21	58
female	70	.6714286	.4730851	0	1
i1	70	1.257143	1.699836	0	6
drugrisk	70	.9285714	2.379389	0	9

```
-> homeless = 1
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	21	33.52381	7.737048	23	48
female	21	.5714286	.5070926	0	1
i1	21	1.285714	2.305273	0	7
drugrisk	21	.0952381	.3007926	0	1

What about stratum 4?

```
. bysort homeless: sum age female i1 drugrisk if quint5 ==4
```

```
-> homeless = 0
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	37	35.32432	4.904952	27	48
female	37	.0540541	.2292434	0	1
i1	37	18.62162	4.917945	9	26
drugrisk	37	1.513514	3.746269	0	14

```
-> homeless = 1
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	53	37.71698	9.810468	19	60
female	53	.1698113	.37906	0	1
i1	53	18.84906	6.794784	2	29
drugrisk	53	2.396226	5.248955	0	21

Stratum 5

```
. bysort homeless: sum age female i1 drugrisk if quint5 ==5
```

```
-----
```

```
-> homeless = 0
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	18	40.11111	7.583953	27	55
female	18	.1666667	.3834825	0	1
i1	18	33	6.087596	26	51
drugrisk	18	1.777778	5.341905	0	21

```
-----
```

```
-> homeless = 1
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	72	38.15278	7.328624	20	57
female	72	.1944444	.3985498	0	1
i1	72	46.77778	24.47604	15	142
drugrisk	72	2.277778	5.286769	0	20

Stats for i1 by quintile

```
. table quint5, c(mean i1 sd i1 min i1 max i1)
```

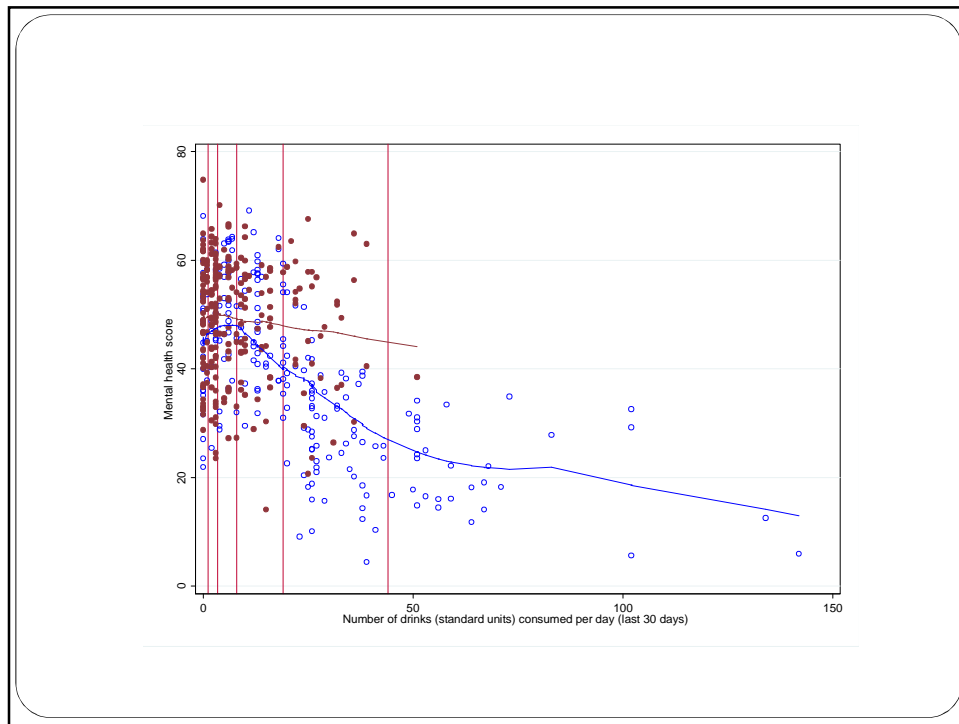
```
-----
```

5 quantiles of pscore	mean(i1)	sd(i1)	min(i1)	max(i1)
1	1.26374	1.842915	0	7
2	3.36264	2.382886	0	11
3	7.9	4.248066	0	19
4	18.7556	6.063912	2	29
5	44.0222	22.70919	15	142

```
-----
```

```
. tab quint5 homeless
```

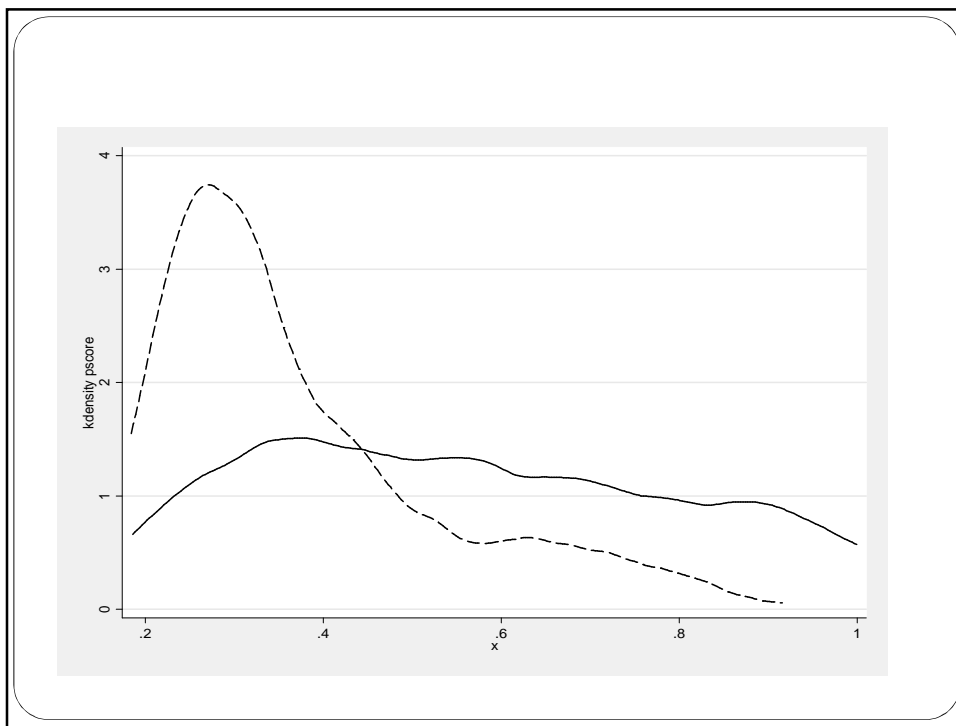
5 quantiles of pscore	1 if Homeless		Total
	0	1	
1	70	21	91
2	66	25	91
3	52	38	90
4	37	53	90
5	18	72	90
Total	243	209	452



Weighting

- Use the propensity score as a weight in a regression
- The main idea is that the weighted data will be balanced
- For treated units: $1/\text{pscore}$. For controls: $1/(1-\text{pscore})$.
- You may see this written in a different way: $w_i = Z_i/e_i + (1-Z_i)/(1-e_i)$

Warning: Unstable weights in some cases (propensity score close to zero or one)



Weighting example

```

gen      w = 1/pscore if homeless == 1
replace w = 1/(1-pscore) if homeless == 0
* now, compare the groups again weighting
. bysort homeless: sum age female il drugrisk [aweight=w]

```

-> homeless = 0

Variable	Obs	Weight	Mean	Std. Dev.	Min	Max
age	243	441.817611	35.58435	7.058303	21	58
female	243	441.817611	.2210831	.4158329	0	1
il	243	441.817611	12.82159	13.24991	0	51
drugrisk	243	441.817611	1.925442	4.516877	0	21

-> homeless = 1

Variable	Obs	Weight	Mean	Std. Dev.	Min	Max
age	209	450.540094	35.67488	8.281236	19	60
female	209	450.540094	.2286887	.4209967	0	1
il	209	450.540094	15.10925	19.02263	0	142
drugrisk	209	450.540094	1.810754	4.315663	0	21

Run a weighted regression

```
. reg pcs homeless age female il drugrisk [pweight = w]
(sum of wgt is 8.9236e+02)
```

Linear regression

Number of obs = 452
 F(5, 446) = 53.12
 Prob > F = 0.0000
 R-squared = 0.2931
 Root MSE = 11.111

	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
homeless	-5.198854	1.201281	-4.33	0.000	-7.559728	-2.837981
age	-.2360235	.0767722	-3.07	0.002	-.3869037	-.0851433
female	-5.687987	1.373758	-4.14	0.000	-8.38783	-2.988144
il	-.3369473	.0346501	-9.72	0.000	-.4050452	-.2688495
drugrisk	-.4088917	.1090059	-3.75	0.000	-.6231207	-.1946627
_cons	63.16712	2.798783	22.57	0.000	57.66668	68.66756

Matching

- Most common: 1 to 1, nearest-neighbor matching without replacement

Treatment		Controls		Treatment id = 1		
id	pscore	id	pscore		Abs Difference	
1	0.11	8	0.334	0.11	0.334	0.22
3	0.13	5	0.11	0.11	0.11	0.00
2	0.11	4	0.131	0.11	0.131	0.02
		7	0.107	0.11	0.107	0.00
		6	0.13	0.11	0.13	0.02

Matches: 1 to 5; 3 to 6; 2 to 7

In Stata

```

preserve
keep pscore id homeless
keep if homeless ==0
set seed 12345
gen ru = runiform()
sort ru
drop ru
gen matched = .
gen mdiff = .
save controls, replace
restore

gsort -homeless

forvalues i=1(1)209 {
    local ps = pscore['i']
    preserve
    use controls, clear
    gen diff = abs(`ps' - pscore)
    sort diff
    replace matched = 1 if _n==1
    replace mdiff = diff if _n==1
    replace pscore = . if _n==1
    drop diff
    save controls, replace
    restore
}

merge 1:1 id using controls
replace matched = 1 if homeless == 1
drop _merge
tab matched homeless

```

Prepare control dataset, save it

Do the matching

```

. tab matched homeless

```

matched	1 if Homeless		Total
	0	1	
1	209	209	418
Total	209	209	418

Is it a good match?

```

. sum age female il drugrisk if (matched == 1 & homeless == 0)

```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	209	34.89474	6.997144	21	58
female	209	.1866029	.3905278	0	1
il	209	9.177033	10.0877	0	51
drugrisk	209	1.933014	4.21254	0	21

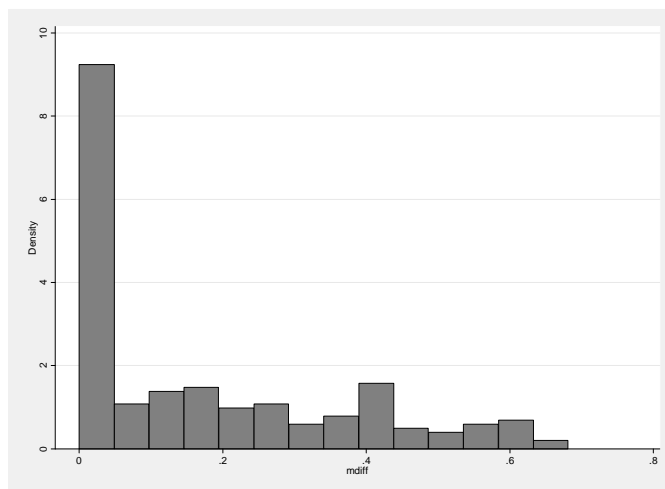
```

. sum age female il drugrisk if homeless == 1

```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	209	36.36842	8.260958	19	60
female	209	.1913876	.3943379	0	1
il	209	23.03828	23.47315	0	142
drugrisk	209	2.07177	4.725098	0	21

Check the difference in propensity scores



```
. sum mdiff, det
```

Percentiles		Smallest		
1%	0	0		
5%	.000043	0		
10%	.0001546	0	Obs	209
25%	.0015045	0	Sum of Wgt.	209
50%	.0574722		Mean	.1688572
		Largest	Std. Dev.	.2070842
75%	.2763925	.7031915		
90%	.5172513	.7037406	Variance	.0428839
95%	.5703443	.7057423	Skewness	1.013636
99%	.7037406	.7265043	Kurtosis	2.769244

It's a jungle out there...

Way too many flavors and combinations

- With replacement
- 1 to many
- Caliper
- Change metric of distance: Mahalanobis distance. You can then include other covariates in the distance metric.
- Global matching: you want to minimize the total distance, not just the local distance (can't do with large datasets)
- Full matching: a hybrid of stratification and matching. Minimizes the average distance
- Genetic algorithms (for large datasets)

Solutions?

- Could try caliper matching. Treatment units would be dropped
- Even harder to interpret results. It is actually a very common way to implement propensity scores
- In your homework you'll have thousands of potential controls. The matching will be very good

Any difference in the regression?

```
.reg pcs homeless age female il drugrisk if matched == 1
```

Source	SS	df	MS	Number of obs =	418
Model	34490.3348	5	6898.06697	F(5, 412) =	57.49
Residual	49430.4831	412	119.976901	Prob > F =	0.0000
				R-squared =	0.4110
				Adj R-squared =	0.4038
				Root MSE =	10.953
Total	83920.818	417	201.248964		

Note: No analysis for paired data here. See Gelman and Hill (2007)

pcs	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
homeless	-5.803444	1.150199	-5.05	0.000	-8.064434 -3.542453
age	-.1727124	.0715858	-2.41	0.016	-.3134314 -.0319935
female	-4.508204	1.368994	-3.29	0.001	-7.199288 -1.817121
il	-.363829	.0304983	-11.93	0.000	-.4237808 -.3038773
drugrisk	-.3475819	.1209357	-2.87	0.004	-.5853099 -.1098539
_cons	60.46967	2.602416	23.24	0.000	55.354 65.58534

- Coef for homeless was 5.38

Conclusions I

- Propensity scores are awesome to check the balance and overlap of covariates
- There are extensions for more than two treatments (multinomial models)
- The propensity score can make groups comparable but only on the variables used to estimate the propensity score. There is NO guarantee that you are balancing unobserved covariates
- If you know that there are important unobservable variables, you may need another tool
- Randomization ensures that both observable and unobservable variables are balanced

Conclusions II

- Think hard about this question: *Do you really need a propensity score analysis?*
- Why not adjusting for confounders?
- There is something about IPW
- If you do propensity scores, what is the external validity?
What are you estimating?