# Using Propensity Scores Lucy D'Agostino McGowan Wake Forest University

# **Propensity scores**

Matching

Weighting

Stratification

Direct Adjustment

• • •



#### estimand

#### Ingredients

- 150g unsalted butter, plus extra for greasing
- 150g plain chocolate, broken into pieces
- 150g plain flour
- 1/2 tsp baking powder
- 1/2 tsp bicarbonate of soda 200g light muscovado
- sugar 2 large eggs

#### Method

- 1. Heat the oven to 160C/140C fan/gas 3. Grease and base line a 1 litre heatproof glass pudding basin and a 450g loaf tin with baking parchment.
- 2. Put the butter and chocolate into a saucepan and melt over a low heat, stirring. When the chocolate has all melted remove from the heat.

#### estimator



#### estimate

Image source: Simon Grund

# **Propensity scores**

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#### **Average Treatment Effect (ATE)**

$$\tau = E[Y(1) - Y(0)]$$

		Treated			Т	reated		
	Tre	ated Treated	Trea	ted Treated				
Treate	ed	Treate	ed					
					Treated			
	Control Cor	ntrol	(	Control Control	Co	ontrolCont	rol	
				Cont	rol	Control		Control
			Control					

Estimand	Target population	Example Research Question
ATE	Full population	Should we decide whether to have extra magic hours all mornings to change the wait time for Seven Dwarfs Mine Train between 9-10 AM?
		Should a specific policy be applied to all eligible observations?

# Average Treatment Effect among the Treated (ATT)

$$\tau = E[Y(1) - Y(0)|Z = 1]$$

Estimand	Target population	Example Research Question
ATT	Exposed (treated) observations	Should we stop extra magic hours to change the wait time for Seven Dwarfs Mine Train between 9-10 AMpm?
		Should we stop our marketing campaign to those currently receiving it?
		Should medical providers stop recommending treatment for those currently receiving it?

## Matching in R (ATT)

```
1 library(MatchIt)
2 m <- matchit(
3  qsmk ~ sex +
4   race + age + I(age^2) + education +
5   smokeintensity + I(smokeintensity^2) +
6   smokeyrs + I(smokeyrs^2) + exercise +
7   active + wt71 + I(wt71^2),
8  data = nhefs_complete
9 )
10 m</pre>
```

```
A `matchit` object
- method: 1:1 nearest neighbor matching without replacement
- distance: Propensity score
- estimated with logistic regression
- number of obs.: 1566 (original), 806 (matched)
- target estimand: ATT
- covariates: sex, race, age, I(age^2), education,
smokeintensity, I(smokeintensity^2), smokeyrs, I(smokeyrs^2),
exercise, active, wt71, I(wt71^2)
```

## Matching in R (ATT)

matched\_data <- get\_matches(m, id = "i")</pre>

```
as_tibble(matched_data)
```

# Average Treatment Effect among the Controls (ATC)

$$\tau = E[Y(1) - Y(0)|Z = 0]$$

Estimand	Target population	Example Research Question
ATU	Unexposed (control) observations	Should we add extra magic hours for all days to change the wait time for Seven Dwarfs Mine Train between 9-10 AMpm?
		Should we extend our marketing campaign to those not receiving it?
		Should medical providers extend treatment to those not currently receiving it?

#### Matching in R (ATC)

```
1 m <- matchit(
2   qsmk ~ sex +
3    race + age + I(age^2) + education +
4   smokeintensity + I(smokeintensity^2) +
5   smokeyrs + I(smokeyrs^2) + exercise +
6   active + wt71 + I(wt71^2),
7   data = nhefs_complete,
8   estimand = "ATC"
9  )
10 m</pre>
```

Average Treatment Effect among the Matched (ATM)

Estimand	Target population	Example Research Question
ATM	Evenly matchable	Are there some days we should change whether we are offering extra magic hours in order to change the wait time for Seven Dwarfs Mine Train between 9-10 AMpm?
		Is there an effect of the exposure for some observations?
		Should those at clinical equipoise receive treatment?

## Matching in R (ATM)

```
1 m <- matchit(
2   qsmk ~ sex +
3     race + age + I(age^2) + education +
4     smokeintensity + I(smokeintensity^2) +
5     smokeyrs + I(smokeyrs^2) + exercise +
6     active + wt71 + I(wt71^2),
7     data = nhefs_complete,
8     link = "linear.logit",
9     caliper = 0.1
10 )
11 m</pre>
```

Observations with propensity scores (on the linear logit scale) within 0.1 standard errors (the caliper) will be discarded

## Matching in R (ATM)

```
A `matchit` object
- method: 1:1 nearest neighbor matching without replacement
- distance: Propensity score [caliper]

- estimated with logistic regression and linearized
- caliper: <distance> (0.063)
- number of obs.: 1566 (original), 780 (matched)
- target estimand: ATT
- covariates: sex, race, age, I(age^2), education,
smokeintensity, I(smokeintensity^2), smokeyrs, I(smokeyrs^2),
exercise, active, wt71, I(wt71^2)
```

## Matching in R (ATM)

```
matched_data <- get_matches(m, id = "i")</pre>
 2 as_tibble(matched_data)
# A tibble: 780 × 71
          subclass weights seqn qsmk death yrdth modth
   <chr> <fct>
                       <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
 1 11
                                428
                                                 \odot
                                                      NA
                                                             NA
 2 1220
                            1 23045
                                                      NA
                                                             NA
 3 15
                                446
                                                      88
 4 1082
                            1 22294
                                                 0
                                                      NA
                                                             NA
   18
                                596
                                                      NA
                                                             NA
 6 534
                            1 14088
                                                             NA
                                                 0
                                                      NA
 7 23
                                618
                                                 0
                                                      NA
                                                             NA
 8 697
                            1 18085
                                                             NA
                                                 0
                                                      NA
   27
                                806
                                                 0
                                                      NA
                                                             NA
10 879
                            1 21128
                                                 0
                                                       NA
                                                             NA
```

#### **Your Turn 1**

Using the propensity scores you created in the previous exercise, create a "matched" data set using the ATM method with a caliper of 0.2.

06:00

# **Propensity scores**

Matching

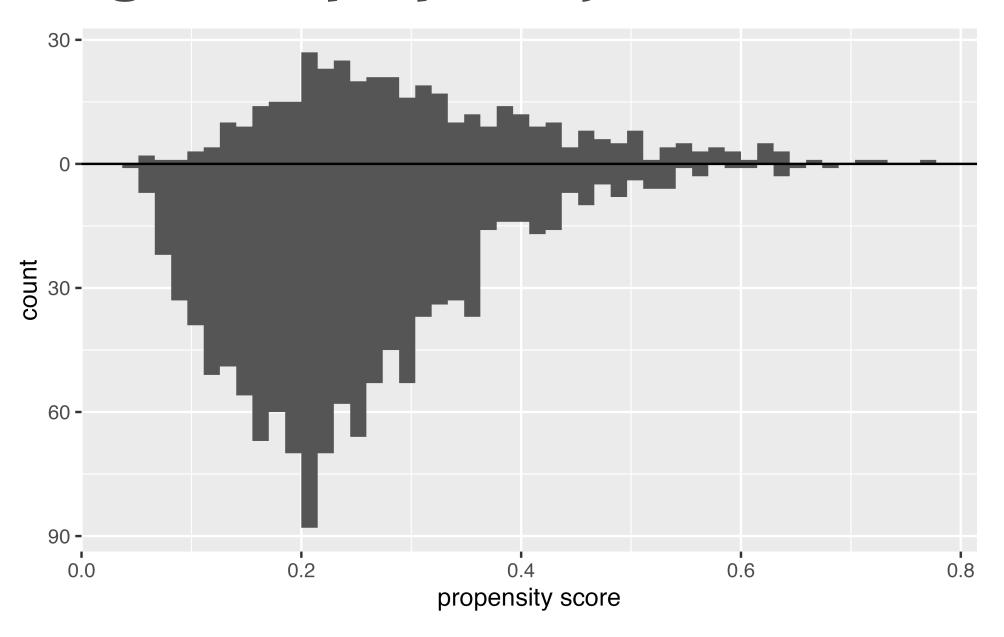
Weighting

Stratification

Direct Adjustment

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## Histogram of propensity scores



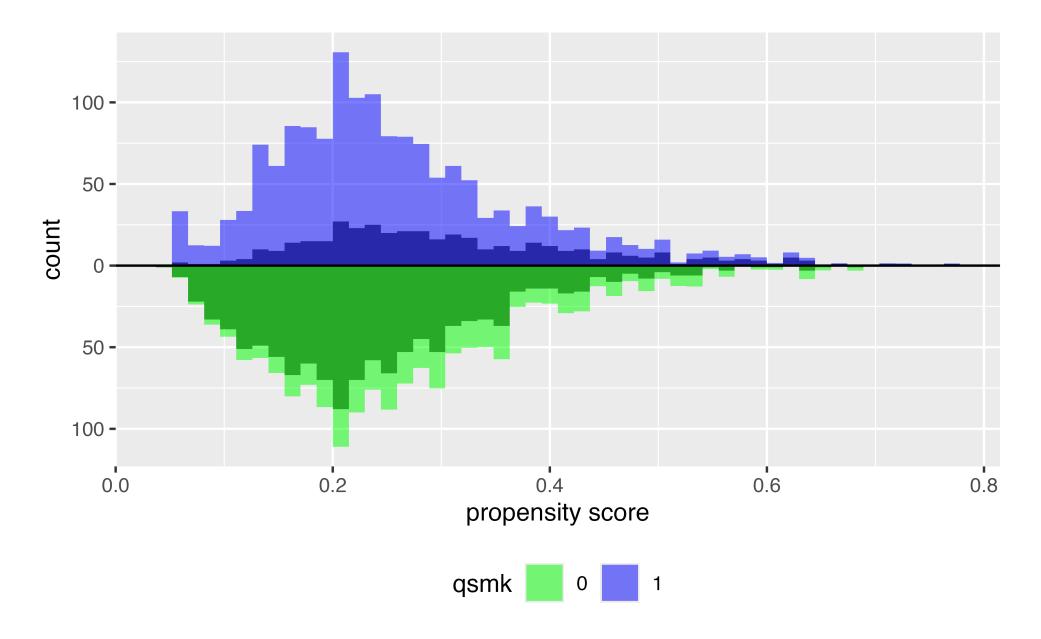
#### Target estimands: ATE

Average Treatment Effect (ATE)

$$w_{ATE} = \frac{Z_i}{p_i} + \frac{1 - Z_i}{1 - p_i}$$

$$1 (Z / p) + ((1 - Z) / (1 - p))$$

#### **ATE**



#### Target estimands: ATT & ATC

Average Treatment Effect Among the Treated (ATT)

$$w_{ATT} = \frac{p_i Z_i}{p_i} + \frac{p_i (1)}{1}$$

1 
$$((p * Z) / p) + ((p * (1 - Z)) / (1 - p))$$

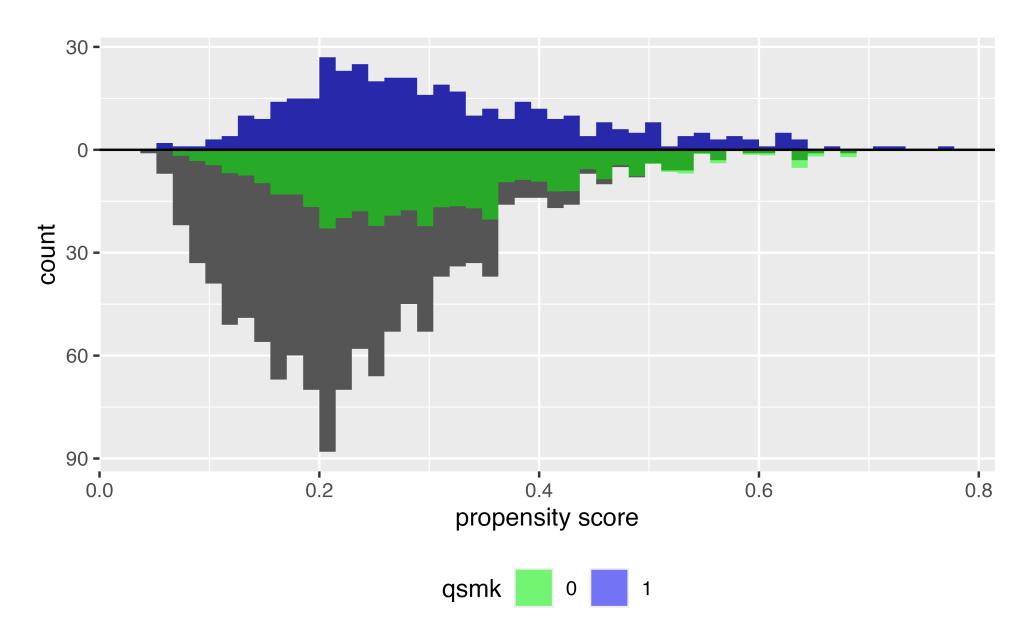
#### Target estimands: ATT & ATC

Average Treatment Effect Among the Controls (ATC)

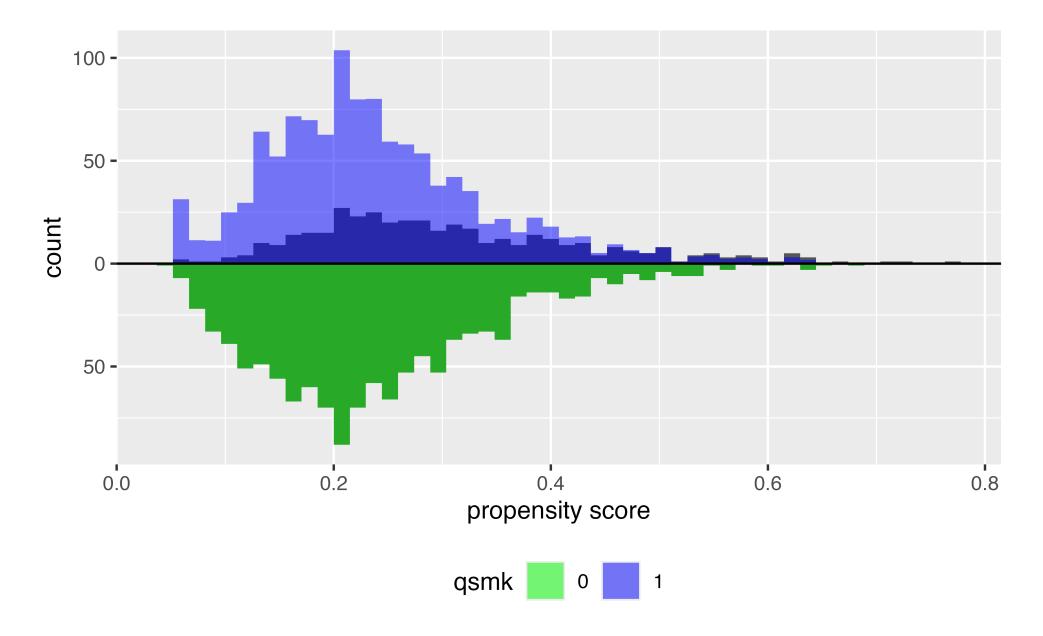
$$w_{ATC} = \frac{(1 - p_i)Z_i}{p_i} +$$

```
1 (((1 - p) * Z) / p) + (((1 - p) * (1 - Z)) / (1 - p))
```

#### **ATT**



#### **ATC**



#### Target estimands: ATM & ATO

Average Treatment Effect Among the Evenly Matchable (ATM)

$$w_{ATM} = \frac{\min\{p_i, Z_i, p_i + (1 - 1)\}}{Z_i p_i + (1 - 1)}$$

```
1 pmin(p, 1 - p) / (Z * p + (1 - Z) * (1 - p))
```

#### Target estimands: ATM & ATO

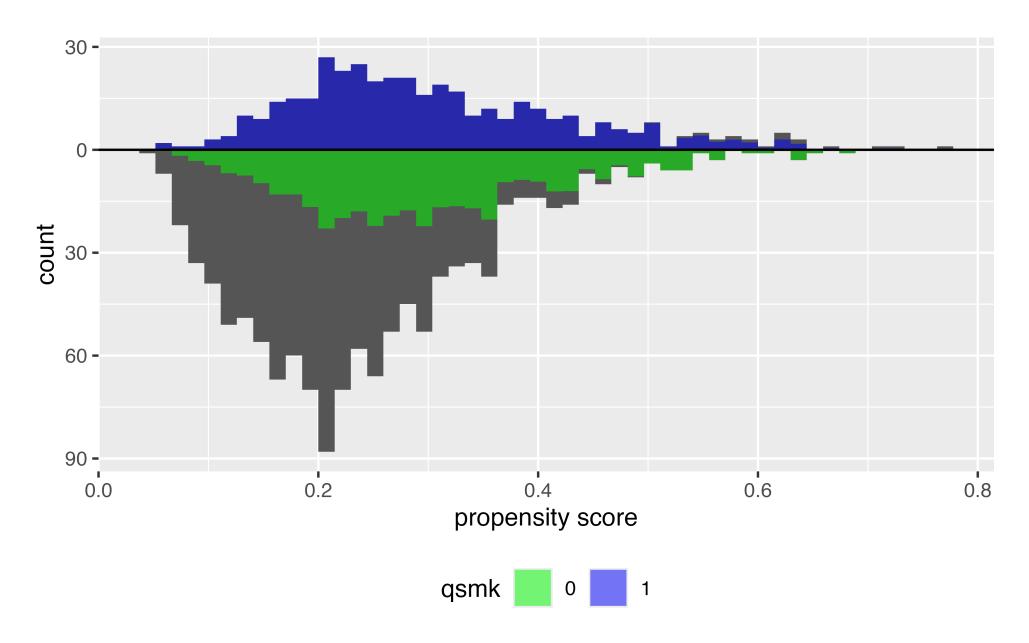
Average Treatment Effect Among the Overlap Population

$$w_{\Delta T \Omega} = (1 - p_i)Z_i +$$

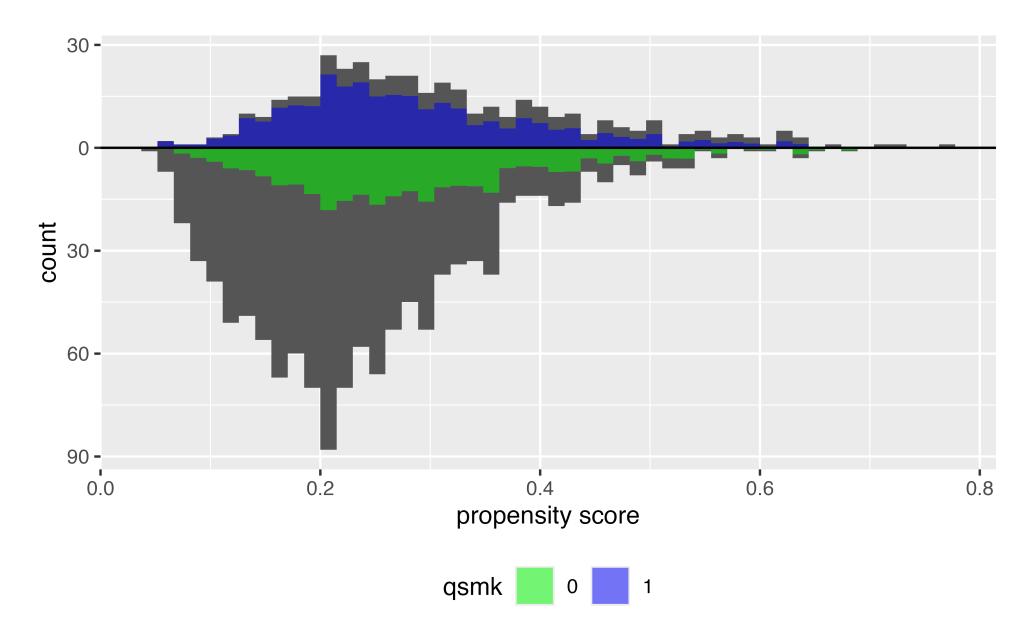
1 
$$(1 - p) * Z + p * (1 - Z)$$

<b>Estimand</b>	<b>Target population</b>	<b>Example Research Question</b>
ATO	Overlap population	Same as ATM

#### **ATM**



#### **ATO**



#### ATE in R



#### Average Treatment Effect (ATE) $w_{ATE} = \frac{Z_i}{p_i} + \frac{1-Z_i}{1-p_i}$

```
1 library(propensity)
2 df <- propensity_model |>
3   augment(type.predict = "response", data = nhefs_complete) |>
4   mutate(w_ate = wt_ate(.fitted, qsmk))
```

#### **Your Turn 2**

Using the propensity scores you created in the previous exercise, add the ATE weights to your data frame

**Stretch**: Using the same propensity scores, create ATM weights

06:00