

# IDS 702: MODULE 6.6

## PROPENSITY SCORES

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# PROPENSITY SCORES

- The **propensity score** (ps) is defined as the conditional probability of receiving a treatment given pre-treatment covariates  $X$ .
- That is,

$$e(X) = \Pr[W = 1|X] = \mathbb{E}[W|X],$$

where  $X = (X_1, \dots, X_p)$  is the vector of  $p$  covariates/predictors.

- Propensity score is a probability, analogous to a summary statistic.
- Propensity score has really nice properties which makes it desirable to use within our causal inference framework.

# BALANCING PROPERTY OF PROPENSITY SCORE

- **Property 1.** The propensity score  $e(X)$  balances the distribution of all  $X$  between the treatment groups:

$$W \perp X | e(X)$$

- Equivalently,

$$\Pr[W_i = 1 | X_i, e(X_i)] = \Pr[W_i = 1 | e(X_i)].$$

- The propensity score is NOT the only **balancing score**. Generally, a balancing score  $b(x)$  is a function of the covariates such that:

$$W \perp X | b(X)$$

# REMARKS ON THE BALANCING PROPERTY

- Rosenbaum and Rubin (1983) show that all balancing scores are a function of  $e(X)$ .
- If a subclass of units or a matched treatment-control pair are homogeneous in  $e(X)$ , then the treatment and control units have the same distribution of  $X$ .
- The balancing property is a statement on the distribution of  $X$ , NOT on assignment mechanism or potential outcomes.

# PROPENSITY SCORE: UNCONFOUNDEDNESS

- **Property 2.** If  $W$  is unconfounded given  $X$ , then  $W$  is unconfounded given  $e(X)$ , i.e.,
- That is, if

$$Y_i(0), Y_i(1) \perp W_i | X_i$$

holds, then

$$Y_i(0), Y_i(1) \perp W_i | e(X_i),$$

also holds.

- Given a vector of covariates that ensure unconfoundedness, adjustment for differences in propensity scores removes all biases associated with differences in the covariates.

# PROPENSITY SCORE: UNCONFOUNDEDNESS

- $e(X)$  can be viewed as a summary score of the observed covariates.
- This is great because causal inference can then be drawn through stratification, matching, regression, etc. using the scalar  $e(X)$  instead of the high dimensional covariates.
- The propensity score balances the **observed covariates**, but does not generally balance **unobserved covariates**.
- In most observational studies, the propensity score  $e(X)$  is unknown and thus needs to be estimated.
- However, since we always observe  $X$  and  $W$ , estimation can be done using models for binary outcomes.

# WHAT'S NEXT?

MOVE ON TO THE READINGS FOR THE NEXT MODULE!