STANDARDIZATION AND G-FORMULA

Group meeting: *Causal Inference What If* reading

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STANDARDIZATION

• Basic idea: for a time-fixed setting, given Y, A, L, and identifiability conditions hold; then the marginal counterfactual outcome

$$E[Y^a] = \sum_{l} E[Y|A = a, L = l] \Pr[L = l]$$

or

$$\int E[Y|A = a, L = l]f_L(l) \, \mathrm{d}l$$

• Rationale and link to Chapter 2 (Conditionally randomized trial setting): The treatment A is randomized with respect to the probability Pr[L = l]or density $f_L(l)$.



RATIONALE 2

- IPTW and Standardization deal confounders in different ways:
- **IPTW:** create a pseudo-population in which the confounders are balanced. Require estimating treatment probability Pr[A = a|L = l]; or $f_A(a|l)$

for each treatment level.

• **Standardization:** create a trial in which the treatment is assigned w.r.t. the joint distribution of confounders. Require estimating conditional mean of the outcome

$$E[Y|A = a, L = l]$$

for each individual in the study population.



IDENTIFIABILITY CONDITIONS

- Consistency: $Y = Y^a$ among those whose A = a
- (Structural) positivity: $f_A(a|l) > 0$ where $f_L(l) > 0$
- Conditional exchangeability: $Y^{a} \coprod A | L$
- Note: structural and random violation of positivity for variable space containing continuous random variables.



SIMPLE ESTIMATION

- 1. In time-fixed settings, only conditional mean part is needed to be estimated. Empirical distribution of *L* will be used in the second part of the *g*-formula.
- 2. Nonparametric estimation: See section 2 example, for single discrete *L*.
- 3. Parametric estimation: Require parametrization of the conditional mean part.



ILLUSTRATION EXAMPLE

- 4 steps: Expansion of dataset, outcome modelling, prediction, and standardization by averaging
- Parametrization choices: be careful about model assumptions



| Table 2.2 | | | |
|------------|---|---|---|
| | L | A | Y |
| Rheia | 0 | 0 | 0 |
| Kronos | 0 | 0 | 1 |
| Demeter | 0 | 0 | 0 |
| Hades | 0 | 0 | 0 |
| Hestia | 0 | 1 | 0 |
| Poseidon | 0 | 1 | 0 |
| Hera | 0 | 1 | 0 |
| Zeus | 0 | 1 | 1 |
| Artemis | 1 | 0 | 1 |
| Apollo | 1 | 0 | 1 |
| Leto | 1 | 0 | 0 |
| Ares | 1 | 1 | 1 |
| Athena | 1 | 1 | 1 |
| Hephaestus | 1 | 1 | 1 |
| Aphrodite | 1 | 1 | 1 |
| Polyphemus | 1 | 1 | 1 |
| Persephone | 1 | 1 | 1 |
| Hermes | 1 | 1 | 0 |
| Hebe | 1 | 1 | 0 |
| Dionysus | 1 | 1 | 0 |

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| Second block: All untreated | | | | | | |
|-----------------------------|---|---|---|--|--|--|
| | L | A | Y | | | |
| Rheia | 0 | 0 | | | | |
| Kronos | 0 | 0 | | | | |
| Demeter | 0 | 0 | | | | |
| Hades | 0 | 0 | | | | |
| Hestia | 0 | 0 | | | | |
| Poseidon | 0 | 0 | | | | |
| Hera | 0 | 0 | | | | |
| Zeus | 0 | 0 | | | | |
| Artemis | 1 | 0 | | | | |
| Apollo | 1 | 0 | | | | |
| Leto | 1 | 0 | | | | |
| Ares | 1 | 0 | | | | |
| Athena | 1 | 0 | | | | |
| Hephaestus | 1 | 0 | | | | |
| Aphrodite | 1 | 0 | | | | |
| Polyphemus | 1 | 0 | | | | |
| Persephone | 1 | 0 | | | | |
| Hermes | 1 | 0 | | | | |
| Hebe | 1 | 0 | | | | |
| Dionysus | 1 | 0 | | | | |

| Third block: All treated | | | | | |
|--------------------------|---|---|---|--|--|
| | L | A | Y | | |
| Rheia | 0 | 1 | | | |
| Kronos | 0 | 1 | | | |
| Demeter | 0 | 1 | | | |
| Hades | 0 | 1 | | | |
| Hestia | 0 | 1 | | | |
| Poseidon | 0 | 1 | | | |
| Hera | 0 | 1 | | | |
| Zeus | 0 | 1 | | | |
| Artemis | 1 | 1 | | | |
| Apollo | 1 | 1 | | | |
| Leto | 1 | 1 | | | |
| Ares | 1 | 1 | | | |
| Athena | 1 | 1 | | | |
| Hephaestus | 1 | 1 | | | |
| Aphrodite | 1 | 1 | | | |
| Polyphemus | 1 | 1 | | | |
| Persephone | 1 | 1 | | | |
| Hermes | 1 | 1 | | | |
| Hebe | 1 | 1 | | | |
| Dionysus | 1 | 1 | | | |





 Difference between standardization and outcome regression: Standardization: marginal mean Outcome regression: conditional mean (conditioning on *L*).



SOME OTHER TOPICS

About g-formula

- 1. Big g-formula
- 2. front door formula
- 3. g-formula in time-varying settings
- 4. Estimation choices



QUESTIONS?

And suggestions?

REFERENCES

