Marginal structural models (MSM)

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Motivation:

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PFIMARY DIFFERENCE: MSMs model potential outnomes Y" as opposed to observed outnomes Y.

- 19 TW estimation is used for simple causal effects s.a.

average coursal effect (ACE)

Lo BUT we can also use IPTW methods to estimate causal effect params, from models

Marginal: model that is not conditional on the confounders (thus allowing use to infer population average estimates)

Structural: model for potential outcomes, NOT observed outcomes

1) Linear MSM: params. we want to estimate

(continuous outwome)

- E[Ya] = 40 + 41 a a = 0, 1

then: $E[Y^0] = \psi_0$ and $E[Y^1] = \psi_0 + \psi_1$

thus: Y is the average causal effect E[Y']- E[Y'] = (Y,+Y,)-(Y,)=Y,

MSM: is a model for the mean of the potential outcomes Ya. (i.e. ELYa])

2) Logistic MSM: (equivalent to P(Y=1) blc Ya is binary.)

- logit (E[Ya]) = Yo + Y1 a a = 0,1

- so exp (4,) is the causal odds ratio

- the mean of a binary outwork is a PROBABILITY of the outworke being equal to 1.

entire pop.

odds ratio

$$\frac{P(Y^1=1)}{1-P(Y^1=1)} \text{ odds that } Y^2=1$$

$$\frac{P(Y^0=1)}{1-P(Y^0=1)} \text{ odds that } Y^0=1$$

MSM with Effect Modification (a.k.a. heterogeneity of treatment effect):

- Suppose V is a variable that modifies the effect of treatment A

[e.g. V=diabetes, where V(diabetes=Yes)=1

V(diabetes=No)=0)

instead of cond. on all confounders X, only cond. on V $E[Y^{\alpha}|V] = Y_{0} + Y_{1} \alpha + Y_{3} V \cdot Y_{4} \cdot \alpha \cdot V \qquad (\alpha = 0, 1) \qquad \text{a.V}$ then $E[Y^{1}|V] - E[Y^{0}|V] = Y_{1} + Y_{4} V \qquad \text{main effect for modifier } V$

The of effect modifier: At different values of V, there might be different treatment effects

- V... effect modifier

General MSM:

$$- g\{E[Y^{\alpha}|V]\} = h(a,V;\Psi)$$

g { · 3 ... linh function

h (.) ... is a fext. specifying parametric form of a and V (typically additive, linear)