

Math Notes Week 2

p. 475 Neyman-Rubin-Holland Formulation From Holland (1988)  
 potential outcome  $Y(u, s)$  response observed if unit  $u$  exposed to cause  $s \in K$   
 U units in pop  
 K causes (treatments)  
 S assignment rule  
 p. 476 actual exposure  $S(u)$ , observed response  $Y(u, S(u)) = \underline{Y}_S(u)$

Unit-level Causal effect ( $t, c$  in  $K$ )

$$Y(u, t) - Y(u, c) = T_{tc}(u)$$

unobserved counterfactual  
 can't observe both

p. 477 Unit homogeneity (lab setting)

$$Y(u, s) = Y(v, s) \text{ all } s \in K, u, v \in U$$

Causal Effect  $T_{tc}(u) = Y_t(u) - Y_c(u)$  constant effect

Fisher's  $H_0$   $T_{tc}(u) = 0$   $t, c \in K, u \in U$

p. 478 ACE (p. 478)  
 unobservable average over  $u$

$P(S=s)$  proportion units  $S(u)=s$   
 $E(Y_s | S=t)$  average value of  $Y$  for units  $S(u)=t$

$$ACE_{tc} = E(T_{tc}) \quad ACE_{tc}(Y) = E(Y_t) - E(Y_c)$$

aka ATE (econ)

observable

$$FACE_{tc}(Y) = E(Y_t | S=t) - E(Y_c | S=c)$$

p. 479 randomization makes independence assumption,  $S$  indep  $Y$ , plausible, so that  $FACE_{tc}(Y) = ACE_{tc}(Y)$

p. 480 more detail: for constant effect  $T_{tc}(u)$  same for all  $u = \tau_{tc}$

$$ACE_{tc}(Y) = \tau_{tc}$$

$$FACE_{tc}(Y) = \tau_{tc} + \{E(Y_c | S=t) - E(Y_c | S=c)\}$$

value of  $Y$  in control condition for treatment subjects

Covariates:  $X$   $X(u, s)$  does not depend on  $s$

$$ATE = E(Y_t | S=t) - E(Y_c | S=t)$$

"exogenous" counterfactual MIT p. 3

Observational Studies  
(week 5)SEW not arranged  
not index of  $Y_S$ pretest equivalence:  $P\{X=x | S=s\}$  does not depend  
on  $S$ conditional independence (strong ignorability)  
given  $X$ 

$$C-FACE_{tc}(Y) = E\{E(Y_S | S=t, X) - E(Y_S | S=c, X)\}$$

what conditions equals  $ACE_{tc}(Y)$ (regression, discontinuity, model selection)  
function

Freedman at Stanford

What would it take to make this stick?

Response schedules and invariance. Potential outcomes

There are two treatments (levels  $u$  and  $v$ ), and a response variable  $Y$ . Both treatments may be applied to subject  $i$ . There are three parameters,  $a$ ,  $b$ , and  $c$ . With no treatment at all, response level for subject  $i$  is  $a$ , up to random error. Each additional unit of treatment #1 adds  $b$  to the response. Likewise, each additional unit of treatment #2 adds  $c$  to the response. Constancy of parameters across subjects and levels of treatment is an assumption. If treatment #1 is applied at level  $u$  and #2 at level  $v$ , response is

$$Y_{i,u,v} = a + bu + cv + \epsilon_i.$$

"potential"

Invariance of  $a, b, c, \epsilon_i$ ? My response is unaffected by your treatment?? Manipulation???

Statistical assumptions

In order to make the transition from a hypothetical experiment to the actual observational study, and to justify OLS, we assume:

- (i)  $E(\epsilon_i) = 0$ .
- (ii)  $\epsilon_i$  are independent and identically distributed across subjects  $i$ .
- (iii) Exogeneity. Nature chooses  $U_i, V_i$  independently of the random errors  $\epsilon_i$ , and determines the response  $Y_i$  from the response schedule:

$$Y_i = Y_{i,U_i,V_i} = a + bU_i + cV_i + \epsilon_i.$$

Nature shows us  $U_i, V_i, Y_i$ . We're good to go. a) OLS works. b) Causal inferences justified—built into the response schedule. (With small samples, need to assume errors are normal.)