

11/3/2011

Given $\theta = (L(0), A(0), L(1), A(1), Y \in \mathcal{L}(2)) \sim P_0 \in \mathcal{M}$

our $\mathcal{M} = \{Q_{L(2)}, Q_{L(1)}, Q_{A(0)}, g_{A(0)}, g_{A(1)} : g : \mathcal{G}\}$, nonparam.

$$P^{a=(a(0), a(1))} (l(0), l(1), l(2)) = Q_{L(0)}(l(0)) Q_{L(1)}(l(1) | A(0) = a(0), L(0) = l(0)) \\ Q_{L(2)}(l(2) | l(1), A(1) = a(1))$$

$$L^q = (L(0), L(1), L(2)) \sim P^q$$

$$\text{Our param } \psi(P) = E_P Y^q$$

① We have to find EIC:

$$\mathcal{M}(g) = \{Q; g : \mathcal{G}\}, \text{ i.e. } g \text{ is known}$$

the EIC for $\psi : \mathcal{M}(g) \rightarrow \mathbb{R}$ equals EIC for $\psi : \mathcal{M} \rightarrow \mathbb{R}$
 i.e. where g is unknown

Need to look at pathwise derivative

$$\psi(Q(\epsilon)) - \psi(Q) = \sum_l \ell(l) [Q(\epsilon)(l; a) - Q(l, a)]$$

$\uparrow \quad \underbrace{\epsilon}_{\text{Note: } Q(l, a) = P(L^q = l)}$

$$E_P[Y^q] \quad E_P[Y^q]$$

$$= P^q(l)$$

$$= \sum_{l, a} \frac{\ell(l) I(a = a)}{g(a; l)} \frac{[Q(\epsilon)(l; a) - Q(l, a)]}{\epsilon Q(l, a)} Q(l, a) g(a; l)$$

$$\text{as } \epsilon \rightarrow 0, \Rightarrow \sum_{a'} \sum_l \left(\frac{\ell(l) I(a' = a)}{g(a'; l)} \right), \text{ & score } S = \frac{\frac{\partial}{\partial \epsilon} Q(\epsilon) |_{\epsilon=0}}{Q}$$

$$\Rightarrow S(a', l) Q(l, a') g(a', l)$$

$$= E_P \left[\underbrace{\left(Y \frac{I(A=a)}{g(A, l)} - \psi \right)}_{\text{is a gradient}} \cdot S(A, l) \right]$$

i.e. $D(\theta) = \left[Y \frac{I(A=a)}{g(A, l)} - \psi \right]$ is a gradient of the

pathwise derivative

↪ cont.

cont. the canonical gradient,

$$\Delta^* \nabla = \nabla (\Delta |_{T_Q}) \text{, where } T_Q \text{ is the tangent space of } Q = (Q_1, Q_2, Q_3)$$

$$= \nabla \left[\Delta |_{T_{Q_0} \oplus T_{Q_1} \oplus T_{Q_2}} \right]$$

↑ orthogonal sum

$$T_{Q_j} = \text{tang. space of } Q_{L(j)} = \{ S(L(j)) / A_a(L(j)) : E(S) \cap A_a(L(j)) = \emptyset \}$$

$$= \sum_{k=0}^2 \nabla \left[\Delta |_{T_{Q_k}} \right]$$

$$= \sum_{k=0}^2 \left[E \left[\Delta |_{L(k)}, P_a(L(k)) \right] - E \left[E \left[\Delta |_{L(k)}, P_a(L(k)) \right] | P_a(L(k)) \right] \right]$$

$$\left(E \left[\frac{Y I(A=a)}{g(A, L)} - \psi |_{L(0)} \right] = E \left[\frac{Y I(A=a)}{g(A, L)} | L(0) \right] - \psi \right)$$

$$= \left(E \left[\frac{Y I(A=a)}{g(A, L)} | L(0) \right] - \psi \right) + E \left[\frac{Y I(A=a)}{g(L(1))} | P_a(L(1)) \right] - E \left[E \left(\frac{Y I(A=a)}{g(L(1))} | L(1), P_a(L(1)) \right) \right]$$

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$$P_a(L(1)) \Big]$$

$$+ \frac{Y I(A=a)}{g(0)} - E \left[Y \frac{I(A=a)}{g(0)} | P_a(L(2)) \right]$$

$$\text{Lemma: } E \left[Y \frac{I(\bar{A}(k) = \bar{a}(k))}{g(\bar{A}(k), \bar{L})} | \bar{L}(j), \underbrace{\bar{A}(j-1)}_{\text{fixed}} \right], j \leq k$$

$$= \frac{I(\bar{A}(j-1) = \bar{a}(j-1))}{g_{j-1}(0)} E \left[Y^a | \bar{L}(j), \bar{A}(j-1) = \bar{a}(j-1) \right]$$

$$\bar{Q}_j(\bar{L}(j)) = E[Y^a | \bar{L}(j)]$$

Now we can rewrite $E[C(\omega | \text{Lemma in mind})]$

$$\Delta^*(0) = \left[E \left[Y^a | L(0) \right] - \psi \right] + \frac{I(A(0) = a(0))}{g_0(0)} \left[E \left[Y^a | \bar{L}(1) P_a(L(1)) \right] - \right]$$

$$\left[-E \left[E \left[Y^a | \bar{L}(1), A(0) = a(0) \right] | L(0), A(0) = a(0) \right] \right]$$

$$+ \frac{I(A(0) = a(0))}{g_0(0)} \left[\bar{Q}_1(\bar{L}(1)) - E \left[\bar{Q}_1(\bar{L}(1)) | L(0), A(0) = a(0) \right] \right]$$

$$+ \frac{I(\bar{A} = \bar{a})}{g(0)} \left[Y - \underbrace{E[Y^a | \bar{L}(1), \bar{A}(1) = \bar{a}]}_{\bar{Q}_1^a(\bar{L}(1))} \right]$$

As, summary result,

$$\begin{aligned} D^*(0) &= [E[Y^a | L(0)] - \psi] + \frac{\pm(A=a)}{g_0(0)} [\bar{Q}_1^a(\bar{L}(0))] - \mathbb{E}[\bar{Q}_1^a(\bar{L}(1)) | L(0), A(0) = a(0)] \\ &\quad + \frac{\pm(A=\bar{a})}{g(0)} [Y - E[Y^a | \bar{L}(1), \bar{A}(1) = \bar{a}]] \end{aligned}$$

$$\text{Our form is } \psi(Q) = \mathbb{E}_{P^a}[Y^a] = E[E[E(Y^a | \bar{L}(1)) | L(0)]]$$

$$\begin{aligned} \text{Note: } P^a &= Q_0^a Q_1^a Q_2^a \\ &= \cancel{P(0)} P(\bar{Q}(0)) \cdot P(\bar{L}(1) | L(0), A(0)) \cdot P(\bar{L}(2) | \bar{L}(1), \bar{A}(1) = \bar{a}(1)) \end{aligned}$$

$$= E \left[E \left[E[Y | \bar{L}(1), \bar{A}(1) = \bar{a}(1)] | L(0), A(0) = a(0) \right] \right]$$

Algorithm

- ① $\bar{Q}_1^a = E[Y | \bar{L}(1), \bar{A}(1) = \bar{a}(1)]$
- ② $\bar{Q}_2^a = E[\bar{Q}_1^a | L(0), A(0) = a(0)]$
- ③ $\bar{Q}_0^a = E[\bar{Q}_1^a] \rightarrow \psi(Q) = \bar{Q}_0^a$

Read his paper!

11/15/2011 Note: Skipped last few classes to prepare for GRE
 * Taught mainly out of slides today

$$\begin{aligned} L_1(t+1) &= L_1(t) \exp(m_1(t)\Delta t) \Rightarrow S_1(t) = \exp(m_1(t)\Delta t) \\ L_2(t+1) &= L_2(t) \exp(m_2(t)\Delta t) \quad S_2(t) = \exp(m_2(t)\Delta t) \end{aligned}$$

↓

$$\frac{\log[S_1(t)]}{\log[S_2(t)]} = \beta$$

Read H Bang, J Robins paper

↳ Doubly robust estimation in missing data & causal inference models