# Conditional Causal Effect for Individual Attribution (Supplementary Material)

### A PROOF OF LEMMA 1

We can write the conditional probability as

$$P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1 \mid \mathbf{X}=\mathbf{x}\right) = \frac{P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1, \mathbf{X}=\mathbf{x}\right)}{P\left(\mathbf{X}=\mathbf{x}\right)}.$$

We first show the identifiability of the numerator.

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{X}=\mathbf{x}\right) \\ = & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},X_{k}=x_{k},\mathbf{D}_{k+1}=\mathbf{d}_{k+1}\right) \\ &= \sum_{c_{k}\leq x_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},(X_{k})_{\mathbf{a}_{k}}=x_{k},(X_{k})_{\mathbf{a}_{k},\mathbf{x}_{S}^{0}}=c_{k},\mathbf{D}_{k+1}=\mathbf{d}_{k+1}\right) \\ &= \sum_{c_{k}\leq x_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},\mathbf{D}_{k+1}=\mathbf{d}_{k+1}\right) \\ &= \sum_{(c_{k},c_{k+1})\leq(x_{k},x_{k+1})} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},(X_{k+1})_{\mathbf{a}_{k+1}}=x_{k+1},(X_{k+1})_{\mathbf{a}_{k},c_{k},\mathbf{x}_{S}^{0}}=c_{k+1},\mathbf{D}_{k+2}=\mathbf{d}_{k+2}\right) \\ &= \sum_{\mathbf{c}_{k:k+1}\leq\mathbf{x}_{k:k+1}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},C_{k+1}=c_{k+1},\mathbf{D}_{k+2}=\mathbf{d}_{k+2}\right), \end{split}$$

where for ease of presentation we use  $C_l = c_l$  to denote  $((X_l)_{\mathbf{a}_l}, (X_l)_{\mathbf{a}_k, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^0}) = (x_l, c_l)$  for  $k \leq l \leq p$  and  $x_l \geq c_l$ , and  $c_l = x_l^0$  if  $l \in \mathbf{S}$ . The second equality holds because of the consistency and the monotonicity assumptions.

Recursively, by the consistency and the composition, we have

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{X}=\mathbf{x}\right) \\ &= P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},\mathbf{D}_{k}=\mathbf{d}_{k}\right) \\ &= \sum_{\mathbf{c}_{k:p} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},\cdots,C_{p}=c_{p}\right) \\ &= \sum_{\mathbf{c}_{k:p} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},\cdots,C_{p}=c_{p}\right) \\ &= \sum_{\mathbf{c}_{k:p} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1,C_{k}=c_{k},\cdots,C_{p}=c_{p} \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \times P(\mathbf{A}_{k}=\mathbf{a}_{k}), \\ &= \sum_{\mathbf{c}_{k:p} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \times \prod_{l=k}^{p} P\left(C_{l}=c_{l} \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \times P(\mathbf{A}_{k}=\mathbf{a}_{k}), \end{split}$$

where the last equality holds as the potential outcomes  $C_{k:p} = (C_k, \dots, C_p)$  are conditionally independent given  $A_k$ . By the no confounding assumption, the first factor can be identified by

$$P\left(Y_{\mathbf{a}_k,\mathbf{c}_{k:p}}=1 \mid \mathbf{A}_k=\mathbf{a}_k\right) = P(Y=1 \mid \mathbf{A}_k=\mathbf{a}_k, X_k=c_k, \cdots, X_p=c_p).$$

Next, we consider the identifiability of  $P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$  for  $l = k + 1, \dots, p$ . For  $l \in \mathbf{S}$ , we have

$$P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$$

$$= P\left((X_l)_{\mathbf{a}_l} = x_l, (X_l)_{\mathbf{a}_k, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^1} = c_l | \mathbf{A}_k = \mathbf{a}_k\right)$$

$$= 1_{c_l = x_l} \cdot P\left((X_l)_{\mathbf{a}_l} = x_l, (X_l)_{\mathbf{a}_k, \mathbf{c}_{k:l-1}, x_l^1} = x_l^1 | \mathbf{A}_k = \mathbf{a}_k\right)$$

$$= 1_{c_l = x_l} \cdot P\left((X_l)_{\mathbf{a}_l} = x_l | \mathbf{A}_k = \mathbf{a}_k\right)$$

$$= 1_{c_l = x_l} \cdot P\left(X_l = x_l | \mathbf{A}_k = \mathbf{a}_k\right),$$

where the second equality holds by the definition of  $c_l$  and the third equality holds by the consistency. For  $l \notin \mathbf{S}$ , we have the following three cases according to the values of  $(x_l, c_l)$ :

•  $(x_l, c_l) = (0, 0)$ : for this case, we have

$$P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$$
  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 0, (X<sub>l</sub>)<sub>**a**<sub>k</sub>, **c**<sub>k:l-1</sub>, **x**<sup>0</sup><sub>S</sub></sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P(X<sub>l</sub> = 0 | **A**<sub>l</sub> = **a**<sub>l</sub>),

where the second and the third equalities hold because of the monotonicity and no confounding assumptions, respectively;

• For the case of  $(x_l, c_l) = (1, 1)$ , we have

$$P(C_{l} = c_{l} | \mathbf{A}_{k} = \mathbf{a}_{k})$$

$$= P\left((X_{l})_{\mathbf{a}_{l}} = 1, (X_{l})_{\mathbf{a}_{k}, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^{0}} = 1 | \mathbf{A}_{k} = \mathbf{a}_{k}\right)$$

$$= P\left((X_{l})_{\mathbf{a}_{k}, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^{0}} = 1 | \mathbf{A}_{k} = \mathbf{a}_{k}\right)$$

$$= P(X_{l} = 1 | \mathbf{A}_{k} = \mathbf{a}_{k}, \mathbf{X}_{k:l-1} = \mathbf{c}_{k+l-1});$$

• For the case of  $(x_l, c_l) = (1, 0)$ , we have

$$P(C_{l} = c_{l} | \mathbf{A}_{k} = \mathbf{a}_{k})$$

$$= P((X_{l})_{\mathbf{a}_{l}} = 1, (X_{l})_{\mathbf{a}_{k}, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^{0}} = 0 | \mathbf{A}_{k} = \mathbf{a}_{k})$$

$$= P((X_{l})_{\mathbf{a}_{l}} = 1 | \mathbf{A}_{k} = \mathbf{a}_{k}) - P((X_{l})_{\mathbf{a}_{l}} = 1, (X_{l})_{\mathbf{a}_{k}, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^{0}} = 1 | \mathbf{A}_{k} = \mathbf{a}_{k})$$

$$= P(X_{l} = 1 | \mathbf{A}_{l} = \mathbf{a}_{l}) - P(X_{l} = 1 | \mathbf{A}_{k} = \mathbf{a}_{k}, \mathbf{X}_{k:l-1} = \mathbf{c}_{k+l-1}).$$

Summarizing the identification equations for the three cases, we get

$$\prod_{l=k}^{p} P\left(C_{l} = c_{l} \mid \mathbf{A}_{k} = \mathbf{a}_{k}\right)$$

$$= \prod_{i \in \{k, \dots, p\} \setminus \mathbf{S}} \left\{ (1 - x_{i}) \times P(X_{i} = 0 \mid \mathbf{A}_{i} = \mathbf{a}_{i}) + x_{i}(1 - c_{i}) \times P(X_{i} = 1 \mid \mathbf{A}_{i} = \mathbf{a}_{i}) + x_{i}(-1)^{1 - c_{i}} \times P\left(X_{i} = 1 \mid \mathbf{A}_{k} = \mathbf{a}_{k}, \mathbf{X}_{k:i-1} = \mathbf{c}_{k:i-1}\right) \right\} \times \prod_{i \in \mathbf{S}} P(X_{i} = x_{i} \mid \mathbf{A}_{i} = \mathbf{a}_{i}) \times 1_{\mathbf{x}_{S} = \mathbf{c}_{S}}.$$

From the above results, the identification formula of  $P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 \mid \mathbf{X} = \mathbf{x}\right)$  can be derived as follows

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1 \mid \mathbf{X}=\mathbf{x}\right) = \frac{P(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1,\mathbf{X}=\mathbf{x})}{P(\mathbf{X}=\mathbf{x})} \\ &= \sum_{\mathbf{c}_{k:p} \preceq \mathbf{d}_{k}} \left[\frac{P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right)}{P(\mathbf{D}_{k}=\mathbf{d}_{k} \mid \mathbf{A}_{k}=\mathbf{a}_{k})} \times \prod_{l=k}^{p} P\left(C_{l}=c_{l} \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right)\right] \\ &= \sum_{\mathbf{c}_{k:p} \preceq \mathbf{d}_{k}} \left\{1_{\mathbf{x}_{S}=\mathbf{c}_{S}} \times \frac{P\left(Y=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{D}_{k}=\mathbf{c}_{k:p}\right)}{P(\mathbf{D}_{k}=\mathbf{d}_{k} \mid \mathbf{A}_{k}=\mathbf{a}_{k})} \times \prod_{i \in \mathbf{S}} P(X_{i}=x_{i} \mid \mathbf{A}_{i}=\mathbf{a}_{i}) \right. \\ &\times \prod_{i \in \{k, \dots, p\} \setminus \mathbf{S}} \left[(1-x_{i}) \times P(X_{i}=0 \mid \mathbf{A}_{i}=\mathbf{a}_{i}) + x_{i}(1-c_{i}) \times P(X_{i}=1 \mid \mathbf{A}_{i}=\mathbf{a}_{i}) \right. \\ &+ x_{i}(-1)^{1-c_{i}} \times P\left(X_{i}=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{X}_{k:i-1}=\mathbf{c}_{k:i-1}\right)\right] \right\} \\ &= \sum_{\mathbf{c}_{k:p} \preceq \mathbf{d}_{k}} \left\{1_{\mathbf{x}_{S}=\mathbf{c}_{S}} \times P(Y=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{D}_{k}=\mathbf{c}_{k:p}) \right. \\ &\times \prod_{i \in \{k, \dots, p\} \setminus \mathbf{S}} \left[1-x_{i}c_{i}+x_{i}(-1)^{1-c_{i}} \times \frac{P(X_{i}=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{X}_{k:i-1}=\mathbf{c}_{k:i-1})}{P(X_{i}=x_{i} \mid \mathbf{A}_{i}=\mathbf{a}_{i})}\right] \right\}, \end{split}$$

where the last equality holds because

$$(1 - x_i) \times \frac{P(X_i = 0 \mid \mathbf{A}_i = \mathbf{a}_i)}{P(X_i = x_i \mid \mathbf{A}_i = \mathbf{a}_i)} = \begin{cases} 0, & \text{if } x_i = 1; \\ 1 - x_i, & \text{if } x_i = 0; \end{cases}$$

and

$$x_i(1-c_i) \times \frac{P(X_i = 1 \mid \mathbf{A}_i = \mathbf{a}_i)}{P(X_i = x_i \mid \mathbf{A}_i = \mathbf{a}_i)} = \begin{cases} x_i(1-c_i), & \text{if } x_i = 1; \\ 0, & \text{if } x_i = 0. \end{cases}$$

### **B PROOF OF LEMMA 2**

We write the conditional probability as

$$P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1 \mid \mathbf{X}=\mathbf{x}\right) = \frac{P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{X}=\mathbf{x}\right)}{P\left(\mathbf{X}=\mathbf{x}\right)},$$

and we first show the identifiability of the numerator above.

$$P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{X}=\mathbf{x}\right)$$

$$=P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{A}_{k}=\mathbf{a}_{k}, X_{k}=x_{k}, \mathbf{D}_{k+1}=\mathbf{d}_{k+1}\right)$$

$$=\sum_{c_{k}\geq x_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{A}_{k}=\mathbf{a}_{k}, (X_{k})_{\mathbf{a}_{k}}=x_{k}, (X_{k})_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{1}}=c_{k}, \mathbf{D}_{k+1}=\mathbf{d}_{k+1}\right)$$

$$=\sum_{c_{k}\geq x_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{A}_{k}=\mathbf{a}_{k}, C_{k}=c_{k}, \mathbf{D}_{k+1}=\mathbf{d}_{k+1}\right)$$

$$=\sum_{(c_{k}, c_{k+1})\succeq(x_{k}, x_{k+1})} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{A}_{k}=\mathbf{a}_{k}, C_{k}=c_{k}, (X_{k+1})_{\mathbf{a}_{k+1}}=x_{k+1}, (X_{k+1})_{\mathbf{a}_{k}, c_{k}, \mathbf{x}_{\mathbf{S}}^{1}}=c_{k+1}, \mathbf{D}_{k+2}=\mathbf{d}_{k+2}\right)$$

$$=\sum_{\mathbf{c}_{k:k+1}\succeq \mathbf{x}_{k:k+1}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1, \mathbf{A}_{k}=\mathbf{a}_{k}, C_{k}=c_{k}, C_{k+1}=c_{k+1}, \mathbf{D}_{k+2}=\mathbf{d}_{k+2}\right),$$

where  $C_l = c_l$  denotes  $((X_l)_{\mathbf{a}_l}, (X_l)_{\mathbf{a}_k, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^1}) = (x_l, c_l)$  for any  $k \leq l \leq p$  satisfying  $x_l \leq c_l$  and  $c_l = x_l^1$  if  $l \in \mathbf{S}$ . The second equality holds because of the consistency and Assumption 2(a).

Recursively, by the consistency and the composition, we have

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1,\mathbf{X}=\mathbf{x}\right) \\ = & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},\mathbf{D}_{k}=\mathbf{d}_{k}\right) \\ &= \sum_{\mathbf{c}_{k:p}\succeq \mathbf{d}_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},\cdots,C_{p}=c_{p}\right) \\ &= \sum_{\mathbf{c}_{k:p}\succeq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1,\mathbf{A}_{k}=\mathbf{a}_{k},C_{k}=c_{k},\cdots,C_{p}=c_{p}\right) \\ &= \sum_{\mathbf{c}_{k:p}\succeq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1,C_{k}=c_{k},\cdots,C_{p}=c_{p} \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \times P(\mathbf{A}_{k}=\mathbf{a}_{k}), \\ &= \sum_{\mathbf{c}_{k:p}\succeq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \times \prod_{l=k}^{p} P\left(C_{l}=c_{l} \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \times P(\mathbf{A}_{k}=\mathbf{a}_{k}), \end{split}$$

where the last equality holds because of the conditional independencies between the potential outcomes  $C_{k:p} = (C_k, \dots, C_p)$  given  $A_k$ . By the no confounding assumption, the first factor above can be identified by

$$P(Y_{\mathbf{a}_k,\mathbf{c}_{k:p}} = 1 | \mathbf{A}_k = \mathbf{a}_k)$$
  
=P(Y = 1 |  $\mathbf{A}_k = \mathbf{a}_k, X_k = c_k, \cdots, X_p = c_p).$ 

Next, we consider the identifiability of  $P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$  for l = k + 1, ..., p. For  $l \in \mathbf{S}$ , we have

$$P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$$

$$= P\left((X_l)_{\mathbf{a}_l} = x_l, (X_l)_{\mathbf{a}_k, \mathbf{c}_{k:l-1}, \mathbf{x}_{\mathbf{S}}^1} = c_l | \mathbf{A}_k = \mathbf{a}_k\right)$$

$$= 1_{c_l = x_l} \cdot P\left((X_l)_{\mathbf{a}_l} = x_l, (X_l)_{\mathbf{a}_k, \mathbf{c}_{k:l-1}, x_l^1} = x_l^1 | \mathbf{A}_k = \mathbf{a}_k\right)$$

$$= 1_{c_l = x_l} \cdot P\left((X_l)_{\mathbf{a}_l} = x_l | \mathbf{A}_k = \mathbf{a}_k\right)$$

$$= 1_{c_l = x_l} \cdot P\left(X_l = x_l | \mathbf{A}_k = \mathbf{a}_k\right),$$

where the second equality holds by the definition of  $c_l$  and the third equality holds by the consistency. For  $l \notin \mathbf{S}$ , according to the value of  $(x_l, c_l)$  we discuss it for three cases.

• For the case of  $(x_l, c_l) = (0, 0)$ , we have

$$P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$$
  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 0, (X<sub>l</sub>)<sub>**a**<sub>k</sub>, **c**<sub>k:l-1</sub>, **x**<sup>1</sup><sub>S</sub></sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P((X<sub>l</sub>)<sub>**a**<sub>k</sub>, **c**<sub>k:l-1</sub>, **x**<sup>1</sup><sub>S</sub></sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P(X<sub>l</sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>, **X**<sub>k:l-1</sub> = **c**<sub>k:l-1</sub>),

where the second and the third equalities hold bacause of the monotonicity and no confounding assumptions, respectively;

• For the case of  $(x_l, c_l) = (1, 1)$ , we have

$$P(C_l = c_l | \mathbf{A}_k = \mathbf{a}_k)$$
  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 1, (X<sub>l</sub>)<sub>**a**<sub>k</sub>, **c**<sub>k:l-1</sub>, **x**<sup>1</sup><sub>**s**</sub></sub> = 1 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 1 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P(X<sub>l</sub> = 1 | **A**<sub>l</sub> = **a**<sub>l</sub>);

• For the case of  $(x_l, c_l) = (0, 1)$ , we have

$$P(C_{l} = c_{l} | \mathbf{A}_{k} = \mathbf{a}_{k})$$
  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 0, (X<sub>l</sub>)<sub>**a**<sub>k</sub>, **c**<sub>k:l-1</sub>, **x**\_{s}<sup>1</sup></sub> = 1 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>) - P((X<sub>l</sub>)<sub>**a**<sub>l</sub></sub> = 0, (X<sub>l</sub>)<sub>**a**<sub>k</sub>, **c**<sub>k:l-1</sub>, **x**\_{s}<sup>1</sup></sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>)  
=P(X<sub>l</sub> = 0 | **A**<sub>l</sub> = **a**<sub>l</sub>) - P(X<sub>l</sub> = 0 | **A**<sub>k</sub> = **a**<sub>k</sub>, **X**<sub>k:l-1</sub> = **c**<sub>k+l-1</sub>).

Summarizing the identification equations for the three cases, we get

$$\begin{split} &\prod_{l=k}^{p} \mathbf{P}\left(C_{l} = c_{l} \mid \mathbf{A}_{k} = \mathbf{a}_{k}\right) \\ = &\mathbf{1}_{\mathbf{x}_{S} = \mathbf{c}_{S}} \times \prod_{i \in \{k, \dots, p\} \setminus \mathbf{S}} \left\{ (1 - x_{i})c_{i} \times \mathbf{P}(X_{i} = 0 \mid \mathbf{A}_{i} = \mathbf{a}_{i}) + x_{i} \times \mathbf{P}(X_{i} = 1 \mid \mathbf{A}_{i} = \mathbf{a}_{i}) \\ &+ (1 - x_{i})(-1)^{c_{i}} \times \mathbf{P}\left(X_{i} = 0 \mid \mathbf{A}_{k} = \mathbf{a}_{k}, \mathbf{X}_{k:i-1} = \mathbf{c}_{k:i-1}\right) \right\} \times \prod_{i \in \mathbf{S}} \mathbf{P}(X_{i} = x_{i} \mid \mathbf{A}_{i} = \mathbf{a}_{i}). \end{split}$$

From the above results, the identification formula of  $P(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 1 | \mathbf{X} = \mathbf{x})$  can be derived as follows

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1 \mid \mathbf{X}=\mathbf{x}\right) = \frac{P(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=1,\mathbf{X}=\mathbf{x})}{P(\mathbf{X}=\mathbf{x})} \\ &= \sum_{\mathbf{c}_{k:p} \succeq \mathbf{d}_{k}} \left[ \frac{P\left(Y_{\mathbf{a}_{k},\mathbf{c}_{k:p}}=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right)}{P(\mathbf{D}_{k}=\mathbf{d}_{k} \mid \mathbf{A}_{k}=\mathbf{a}_{k})} \times \prod_{l=k}^{p} P\left(C_{l}=c_{l} \mid \mathbf{A}_{k}=\mathbf{a}_{k}\right) \right] \\ &= \sum_{\mathbf{c}_{k:p} \succeq \mathbf{d}_{k}} \left\{ 1_{\mathbf{x}_{S}=\mathbf{c}_{S}} \times \frac{P\left(Y=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{D}_{k}=\mathbf{c}_{k:p}\right)}{P(\mathbf{D}_{k}=\mathbf{d}_{k} \mid \mathbf{A}_{k}=\mathbf{a}_{k})} \times \prod_{i \in \mathbf{S}} P(X_{i}=x_{i} \mid \mathbf{A}_{i}=\mathbf{a}_{i}) \\ &\times \prod_{i \in \{k, \dots, p\} \setminus \mathbf{S}} \left[ (1-x_{i})c_{i} \times P(X_{i}=0 \mid \mathbf{A}_{i}=\mathbf{a}_{i}) + x_{i} \times P(X_{i}=1 \mid \mathbf{A}_{i}=\mathbf{a}_{i}) \\ &+ (1-x_{i})(-1)^{c_{i}} \times P\left(X_{i}=0 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{X}_{k:i-1}=\mathbf{c}_{k:i-1}\right) \right] \right\} \\ &= \sum_{\mathbf{c}_{k:p} \succeq \mathbf{d}_{k}} \left\{ 1_{\mathbf{x}_{S}=\mathbf{c}_{S}} \times P(Y=1 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{D}_{k}=\mathbf{c}_{k:p}) \\ &\times \prod_{i \in \{k, \dots, p\} \setminus \mathbf{S}} \left[ x_{i}+c_{i}-x_{i}c_{i}+(1-x_{i})(-1)^{c_{i}} \times \frac{P(X_{i}=0 \mid \mathbf{A}_{k}=\mathbf{a}_{k}, \mathbf{X}_{k:i-1}=\mathbf{c}_{k:i-1})}{P(X_{i}=x_{i} \mid \mathbf{A}_{i}=\mathbf{a}_{i})} \right] \right\}, \end{split}$$

where the last equality holds because

$$(1-x_i)c_i \times \frac{P(X_i=0 \mid \mathbf{A}_i = \mathbf{a}_i)}{P(X_i=x_i \mid \mathbf{A}_i = \mathbf{a}_i)} = \begin{cases} 0, & \text{if } x_i = 1; \\ (1-x_i)c_i, & \text{if } x_i = 0; \end{cases}$$

.

and

$$x_i \times \frac{\mathbf{P}(X_i = 1 \mid \mathbf{A}_i = \mathbf{a}_i)}{\mathbf{P}(X_i = x_i \mid \mathbf{A}_i = \mathbf{a}_i)} = \begin{cases} x_i, & \text{if } x_i = 1; \\ 0, & \text{if } x_i = 0. \end{cases}$$

### C PROOF OF THEOREM 1

The conclusion follows directly from Lemma 1, Lemma 2 and the definition of CCE.

### **D PROOF OF COROLLARY 1**

For any subset  $\mathbf{X}' \subset \mathbf{X}$ , we have

$$CCE (\mathbf{X}_{\mathbf{S}} \Rightarrow Y | \mathbf{X}' = \mathbf{x}')$$
  
=P  $\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 1 | \mathbf{X}' = \mathbf{x}'\right)$  - P  $\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 | \mathbf{X}' = \mathbf{x}'\right)$   
=  $\sum_{\mathbf{x}:\mathbf{x}\supset\mathbf{x}'} \left[P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 1 | \mathbf{X} = \mathbf{x}\right) - P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 | \mathbf{X} = \mathbf{x}\right)\right] \times P(\mathbf{X} = \mathbf{x} | \mathbf{X}' = \mathbf{x}')$   
=  $\sum_{\mathbf{x}:\mathbf{x}\supset\mathbf{x}'} CCE (\mathbf{X}_{\mathbf{S}} \Rightarrow Y | \mathbf{X} = \mathbf{x}) \times P(\mathbf{X} = \mathbf{x} | \mathbf{X}' = \mathbf{x}').$ 

Hence, CCE ( $\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X}' = \mathbf{x}'$ ) is identifiable if and only if CCE ( $\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}$ ) is identifiable, and its identification formula can be obtained by Theorem 1.

## E PROOF OF THEOREM 2

## **E.1** CCE( $\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = 1$ )

For Y = 1, we have

$$\begin{split} &\operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = 1) = \operatorname{E}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} - Y_{\mathbf{x}_{\mathbf{S}}^{0}} \mid \mathbf{X} = \mathbf{x}, Y = 1) \\ = &1 - \operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 \mid \mathbf{X} = \mathbf{x}, Y = 1) = 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1, \mathbf{X} = \mathbf{x}, Y = 1)}{\operatorname{P}(\mathbf{X} = \mathbf{x}, Y = 1)}. \end{split}$$

By consistency, composition and Assumption 2(a), we have

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1, \mathbf{X}=\mathbf{x}, Y=1\right) \\ &= \sum_{\mathbf{c}_{k} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}}=1, Y_{\mathbf{x}}=1, (\mathbf{A}_{k})_{\mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{a}_{k}, (\mathbf{D}_{k})_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{c}_{k}, \mathbf{X}=\mathbf{x}\right) \\ &= \sum_{\mathbf{c}_{k} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{0}, \mathbf{c}_{k}} = 1, Y_{\mathbf{x}} = 1, (\mathbf{A}_{k})_{\mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{a}_{k}, (\mathbf{D}_{k})_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{c}_{k}, \mathbf{X}=\mathbf{x}\right) \\ &= \sum_{\mathbf{c}_{k} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{0}, \mathbf{c}_{k}} = 1, (\mathbf{A}_{k})_{\mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{a}_{k}, (\mathbf{D}_{k})_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{c}_{k}, \mathbf{X}=\mathbf{x}\right) \\ &= \sum_{\mathbf{c}_{k} \leq \mathbf{d}_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1, (\mathbf{D}_{k})_{\mathbf{a}_{k}, \mathbf{x}_{\mathbf{S}}^{0}} = \mathbf{c}_{k}, \mathbf{X}=\mathbf{x}\right) \\ &= P\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1, \mathbf{X}=\mathbf{x}\right), \end{split}$$

where  $k = \min \mathbf{S}$ . Hence, we have

$$\begin{split} & \operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = 1) = 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1, \mathbf{X} = \mathbf{x}, Y = 1)}{\operatorname{P}(\mathbf{X} = \mathbf{x}, Y = 1)} \\ = & 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1, \mathbf{X} = \mathbf{x})}{\operatorname{P}(\mathbf{X} = \mathbf{x}, Y = 1)} = 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 \mid \mathbf{X} = \mathbf{x})}{\operatorname{P}(Y = 1 \mid \mathbf{X} = \mathbf{x})}. \end{split}$$

**E.2** CCE( $\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = 0$ )

For Y = 0, we have

$$\begin{aligned} & \operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = 0) = \operatorname{E}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} - Y_{\mathbf{x}_{\mathbf{S}}^{0}} \mid \mathbf{X} = \mathbf{x}, Y = 0) \\ & = \operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 1 \mid \mathbf{X} = \mathbf{x}, Y = 0) - 0 = 1 - \operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 0 \mid \mathbf{X} = \mathbf{x}, Y = 0) \\ & = 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 0, \mathbf{X} = \mathbf{x}, Y = 0)}{\operatorname{P}(\mathbf{X} = \mathbf{x}, Y = 0)}. \end{aligned}$$

By consistency, composition and Assumption 2(a), we have

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=0,\mathbf{X}=\mathbf{x},Y=0\right) \\ &= \sum_{\mathbf{c}_{k}\succeq\mathbf{d}_{k}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=0,Y_{\mathbf{x}}=0,(\mathbf{A}_{k})_{\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{a}_{k},(\mathbf{D}_{k})_{\mathbf{a}_{k},\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{c}_{k},\mathbf{X}=\mathbf{x}\right) \\ &= \sum_{\mathbf{c}_{k}\succeq\mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{x}_{\mathbf{S}}^{1},\mathbf{c}_{k}}=0,Y_{\mathbf{x}}=0,(\mathbf{A}_{k})_{\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{a}_{k},(\mathbf{D}_{k})_{\mathbf{a}_{k},\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{c}_{k},\mathbf{X}=\mathbf{x}\right) \\ &= \sum_{\mathbf{c}_{k}\succeq\mathbf{d}_{k}} P\left(Y_{\mathbf{a}_{k},\mathbf{x}_{\mathbf{S}}^{1},\mathbf{c}_{k}}=0,(\mathbf{A}_{k})_{\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{a}_{k},(\mathbf{D}_{k})_{\mathbf{a}_{k},\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{c}_{k},\mathbf{X}=\mathbf{x}\right) \\ &= \sum_{\mathbf{c}_{k}\succeq\mathbf{d}_{k}} P\left(Y_{\mathbf{x}_{\mathbf{s}}^{1}}=0,(\mathbf{D}_{k})_{\mathbf{a}_{k},\mathbf{x}_{\mathbf{S}}^{1}}=\mathbf{c}_{k},\mathbf{X}=\mathbf{x}\right) \\ &= P\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}}=0,\mathbf{X}=\mathbf{x}\right), \end{split}$$

where  $k = \min \mathbf{S}$ . Hence, we have

$$\begin{split} & \operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = 0) = 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 0, \mathbf{X} = \mathbf{x}, Y = 0)}{\operatorname{P}(\mathbf{X} = \mathbf{x}, Y = 0)} \\ = & 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 0, \mathbf{X} = \mathbf{x})}{\operatorname{P}(\mathbf{X} = \mathbf{x}, Y = 0)} = 1 - \frac{\operatorname{P}(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 0 \mid \mathbf{X} = \mathbf{x})}{\operatorname{P}(Y = 0 \mid \mathbf{X} = \mathbf{x})}. \end{split}$$

### F PROOF OF LEMMA 3

Using the notations in this lemma, we have

$$\begin{split} & P\left(Y_{\mathbf{x}_{\mathbf{S}}^{*}}=1, \mathbf{X}=\mathbf{x}, Y=y, \mathbf{Z}=\mathbf{z}\right) \\ &= \sum_{\mathbf{x}^{*}} P\left(Y_{\mathbf{x}_{\mathbf{S}}^{*}}=1, \mathbf{X}_{\mathbf{x}_{\mathbf{S}}^{*}}=\mathbf{x}^{*}, \mathbf{X}=\mathbf{x}, Y=y, \mathbf{Z}=\mathbf{z}\right) \\ &= \sum_{\mathbf{x}^{*}} P\left(Y_{\mathbf{x}^{*}}=1, \mathbf{X}_{\mathbf{x}_{\mathbf{S}}^{*}}=\mathbf{x}^{*}, \mathbf{X}=\mathbf{x}, Y=y, \mathbf{Z}=\mathbf{z}\right) \\ &= \sum_{\mathbf{x}^{*}} P\left(Y_{\mathbf{x}^{*}}=1, \mathbf{X}_{\mathbf{x}_{\mathbf{S}}^{*}}=\mathbf{x}^{*}, \mathbf{Z}=\mathbf{z} \mid \mathbf{X}=\mathbf{x}, Y=y\right) \times P(\mathbf{X}=\mathbf{x}, Y=y) \\ &= \sum_{\mathbf{x}^{*}} P\left(Y_{\mathbf{x}^{*}}=1, \mathbf{X}_{\mathbf{x}_{\mathbf{S}}^{*}}=\mathbf{x}^{*} \mid \mathbf{X}=\mathbf{x}, Y=y\right) \times P(\mathbf{Z}=\mathbf{z} \mid \mathbf{X}=\mathbf{x}, Y=y) \times P(\mathbf{X}=\mathbf{x}, Y=y) \\ &= \sum_{\mathbf{x}^{*}} P\left(Y_{\mathbf{x}^{*}}=1, \mathbf{X}_{\mathbf{x}_{\mathbf{S}}^{*}}=\mathbf{x}^{*} \mid \mathbf{X}=\mathbf{x}, Y=y\right) \times P(\mathbf{X}=\mathbf{x}, Y=y) \times P(\mathbf{X}=\mathbf{x}, Y=y) \end{split}$$

where the second and the fourth equalities hold because of the composition and Assumption 1(c), respectively. Hence, we have

$$P\left(Y_{\mathbf{x}_{\mathbf{S}}^{*}}=1 \mid \mathbf{X}=\mathbf{x}, Y=y, \mathbf{Z}=\mathbf{z}\right) = \frac{P\left(Y_{\mathbf{x}_{\mathbf{S}}^{*}}=1, \mathbf{X}=\mathbf{x}, Y=y, \mathbf{Z}=\mathbf{z}\right)}{P(\mathbf{X}=\mathbf{x}, Y=y, \mathbf{Z}=\mathbf{z})}$$
$$= \sum_{\mathbf{x}^{*}} P\left(Y_{\mathbf{x}^{*}}=1, \mathbf{X}_{\mathbf{x}_{\mathbf{S}}^{*}}=\mathbf{x}^{*} \mid \mathbf{X}=\mathbf{x}, Y=y\right)$$
$$= P\left(Y_{\mathbf{x}_{\mathbf{S}}^{*}}=1 \mid \mathbf{X}=\mathbf{x}, Y=y\right).$$

#### G PROOF OF COROLLARY 3

The conclusion follows directly from Lemma 3 and the definition of CCE.

#### H PROOF OF THEOREM 3

For any subset  $\mathbf{W} \subset (\mathbf{X}, Y, \mathbf{Z})$ , we have

$$\begin{split} & \operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{W} = \mathbf{w}) \\ = & \operatorname{P}\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 1 \mid \mathbf{W} = \mathbf{w}\right) - \operatorname{P}\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 \mid \mathbf{W} = \mathbf{w}\right) \\ = & \sum_{(\mathbf{x}, y, \mathbf{z}): (\mathbf{x}, y, \mathbf{z}) \supset \mathbf{w}} \operatorname{P}(\mathbf{X} = \mathbf{x}, Y = y, \mathbf{Z} = \mathbf{z} \mid \mathbf{W}) \times \left[\operatorname{P}\left(Y_{\mathbf{x}_{\mathbf{S}}^{1}} = 1 \mid \mathbf{X} = \mathbf{x}, Y = y, \mathbf{Z} = \mathbf{z}\right) \\ & - \operatorname{P}\left(Y_{\mathbf{x}_{\mathbf{S}}^{0}} = 1 \mid \mathbf{X} = \mathbf{x}, Y = y, \mathbf{Z} = \mathbf{z}\right)\right] \\ = & \sum_{(\mathbf{x}, y, \mathbf{z}): (\mathbf{x}, y, \mathbf{z}) \supset \mathbf{w}} \operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = y, \mathbf{Z} = \mathbf{z}) \times \operatorname{P}(\mathbf{X} = \mathbf{x}, Y = y, \mathbf{Z} = \mathbf{z} \mid \mathbf{W}) \\ = & \sum_{(\mathbf{x}, y, \mathbf{z}): (\mathbf{x}, y, \mathbf{z}) \supset \mathbf{w}} \operatorname{CCE}(\mathbf{X}_{\mathbf{S}} \Rightarrow Y \mid \mathbf{X} = \mathbf{x}, Y = y) \times \operatorname{P}(\mathbf{X} = \mathbf{x}, Y = y, \mathbf{Z} = \mathbf{z} \mid \mathbf{W}), \end{split}$$

where the last equality holds because of Corollary 3. Hence,  $CCE(\mathbf{X}_{\mathbf{S}} \Rightarrow Y | \mathbf{W} = \mathbf{w})$  is identifiable if and only if  $CCE(\mathbf{X}_{\mathbf{S}} \Rightarrow Y | \mathbf{X} = \mathbf{x}, Y = y)$  is identifiable for any  $(\mathbf{x}, y, \mathbf{z}) \supset \mathbf{w}$ , and under Assumption 1 and Assumption 2, the identification equations are given by Theorem 2.