

# Algorithmic Recourse under incomplete causal graph

# Algorithmic Recourse: from Counterfactual Explanations to Interventions

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# Algorithmic Recourse

The systematic process of reversing unfavourable decisions by algorithms and bureaucracies across a range of counterfactual scenarios.

# Problem Formulation

$$x^{*\text{CFE}} \in \underset{x}{\operatorname{argmin}} \quad \operatorname{dist}(x, x^F) \quad s.t. \quad h(x) \neq h(x^F), x \in \mathcal{P}$$

Contd.

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$$\delta^* \in \underset{\delta}{\operatorname{argmin}} \quad \operatorname{cost}(\delta; x^F) \quad s.t. \quad h(x^{\text{CFE}}) \neq h(x^F),$$

$$x^{\text{CFE}} = x^F + \delta,$$

$$x^{\text{CFE}} \in \mathcal{P}, \delta \in \mathcal{F}$$

# Why does the above formulation fail?

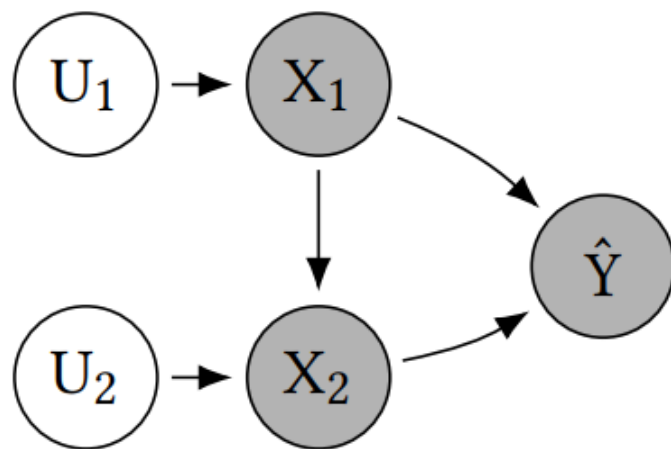
- Example: Consider an example where an individual has an annual salary of \$75,000 and an account balance \$25,000. Say the model is

$$h = \text{sgn}(X_1 + 5X_2 - \$225,000)$$

- The counterfactual explanation could be
  - i. Annual salary: \$100,000 or
  - ii. Bank Balance: \$30,000

In a world where home-seekers save %30 of their salary, a salary increase of 14% would result in positive decision of loan-granting algorithm.

# Structural Causal Model



$$\left. \begin{array}{l} X_1 := U_1 \\ X_2 := f_2(X_1) + U_2 \end{array} \right\} \mathcal{M}$$
$$\hat{Y} = h(X_1, X_2)$$

## Proposition-1

A CFE-based action,  $\mathbf{A}^{CFE}$ , where  $I = \{i | \delta_i^* \neq 0\}$ , performed by individual  $x^F$ , in general results in the structural counterfactual,  $x^{SCF} = x^{*CFE} := x^F + \delta^*$ , if and only if, the set of descendants of the acted upon variables, determined by  $I$ , is the empty set.

## Corollary-1

If the true world  $\mathcal{M}$  is independent, i.e. all the observed features are root-nodes, then CFE-based actions always guarantee recourse.



# Causal Persepective

$$\begin{aligned} \mathbf{A}^* &\in \underset{A}{\operatorname{argmin}} \quad \operatorname{cost}(\mathbf{A}; x^F) \\ s.t. \quad &h(x^{\text{SCF}}) \neq h(x^F), \\ &x^{\text{CFE}} = \mathbb{F}_{\mathbf{A}}(\mathbb{F}^{-1}(\mathbb{X}^F)) \quad x^{\text{CFE}} \in \mathcal{P}, \mathbf{A} \in \mathcal{F} \end{aligned}$$

## Proposition-2

Given an individual  $x^F$  observed in world  $\mathcal{M}$ , a family of feasible actions  $\mathcal{F}$ . Assume that there exists CFE-based actions  $A^{CFE} \in \mathcal{F}$  that achieves recourse, i.e.,  $h(x^F) \neq h(x^{*CFE})$ . Then,  $cost(A^*; x^F) \leq cost(A^{CFE}; x^F)$

# Algorithm

1. **Abduction:** uniquely determines the values of all exogenous variables.
2. **Action:** modify the SCM according to the hypothetical interventions. ( $\mathbb{F}_A$ )
3. **Prediction:** Determine the values of of all endogenous variables.

The assignment of structural counterfactual values can generally be written as

$$\begin{aligned} x_i^{\text{SCF}} = & [i \in I] \cdot (x_i^{\text{F}} + \delta_i) \\ & + [i \notin I] \cdot (x_i^{\text{F}} + f_i(pa_i^{\text{SCF}}) + f_i(pa_i^{\text{F}})) \end{aligned}$$

