

Adjustment Criteria for Generalizing Experimental Findings

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1 Introduction

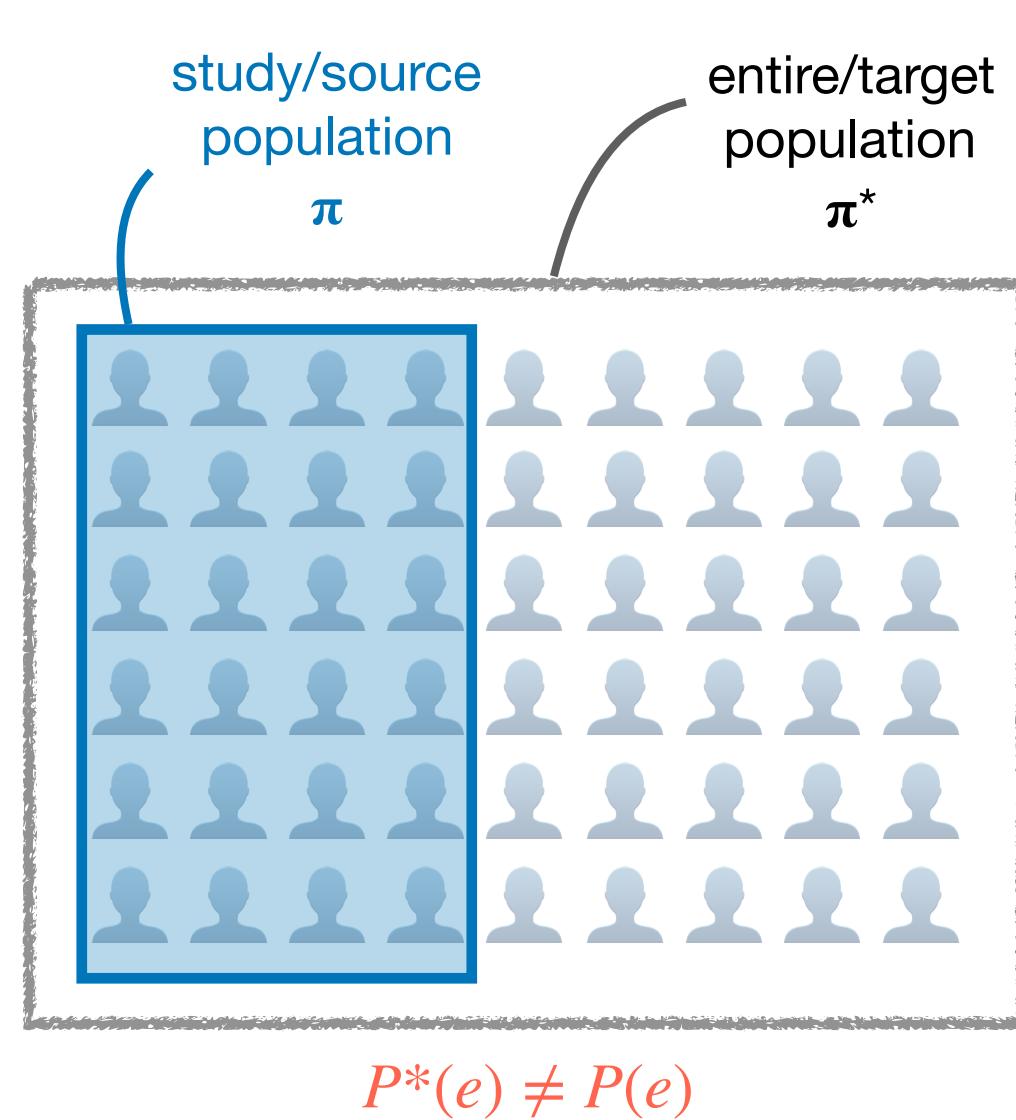
Proper experimentation guarantees **internal validity**, that is, causal conclusions obtained from it hold true for the domain from where the sample was drawn.

However, most experimental findings are intended to be **generalized** to a broader, or even different, **target domain** (in other words, population, setting, or environment).

3 Challenges

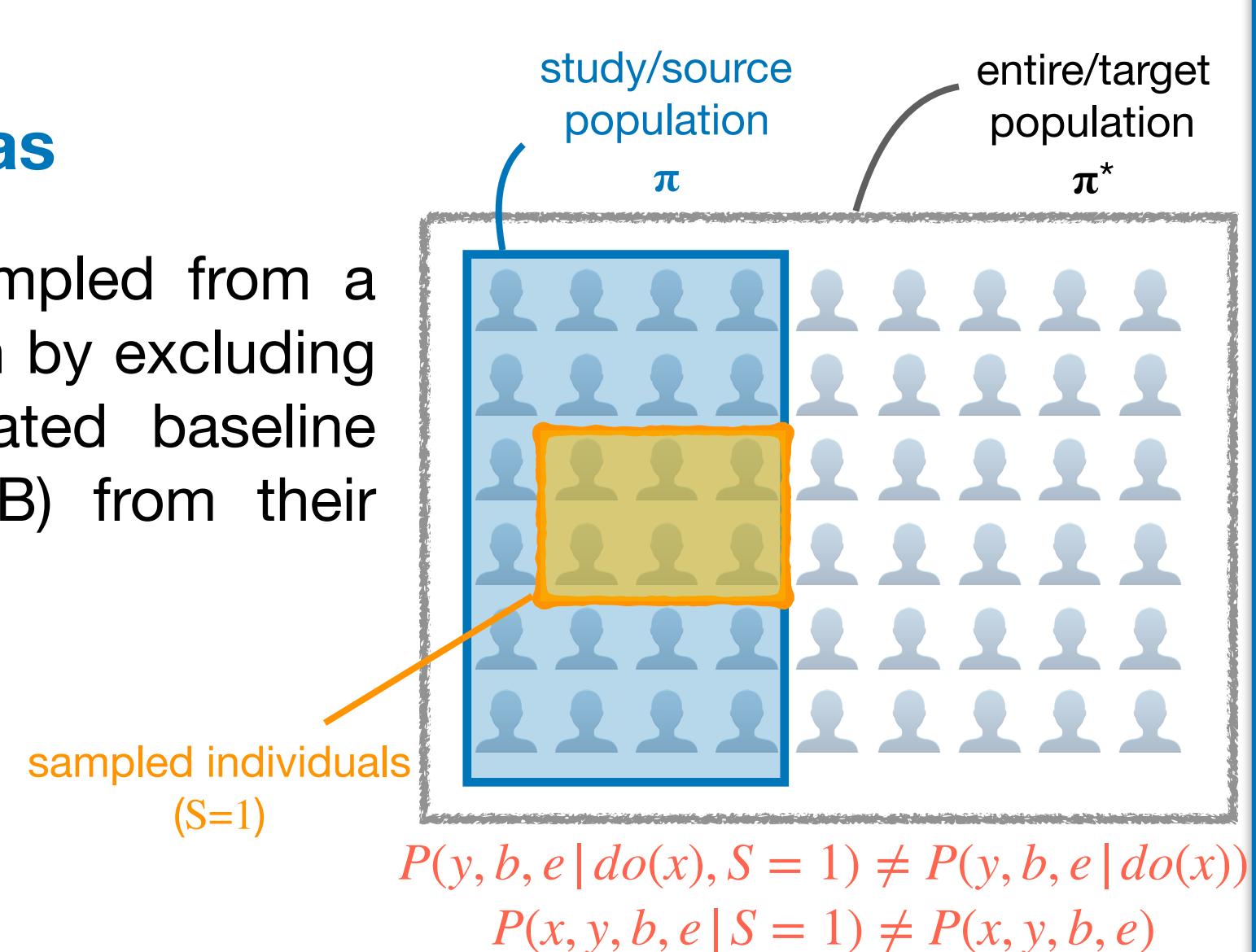
1. Transportability

There is a mismatch between the study population π and the general clinical population π^* regarding ethnicity, race, and income (covariates named E).



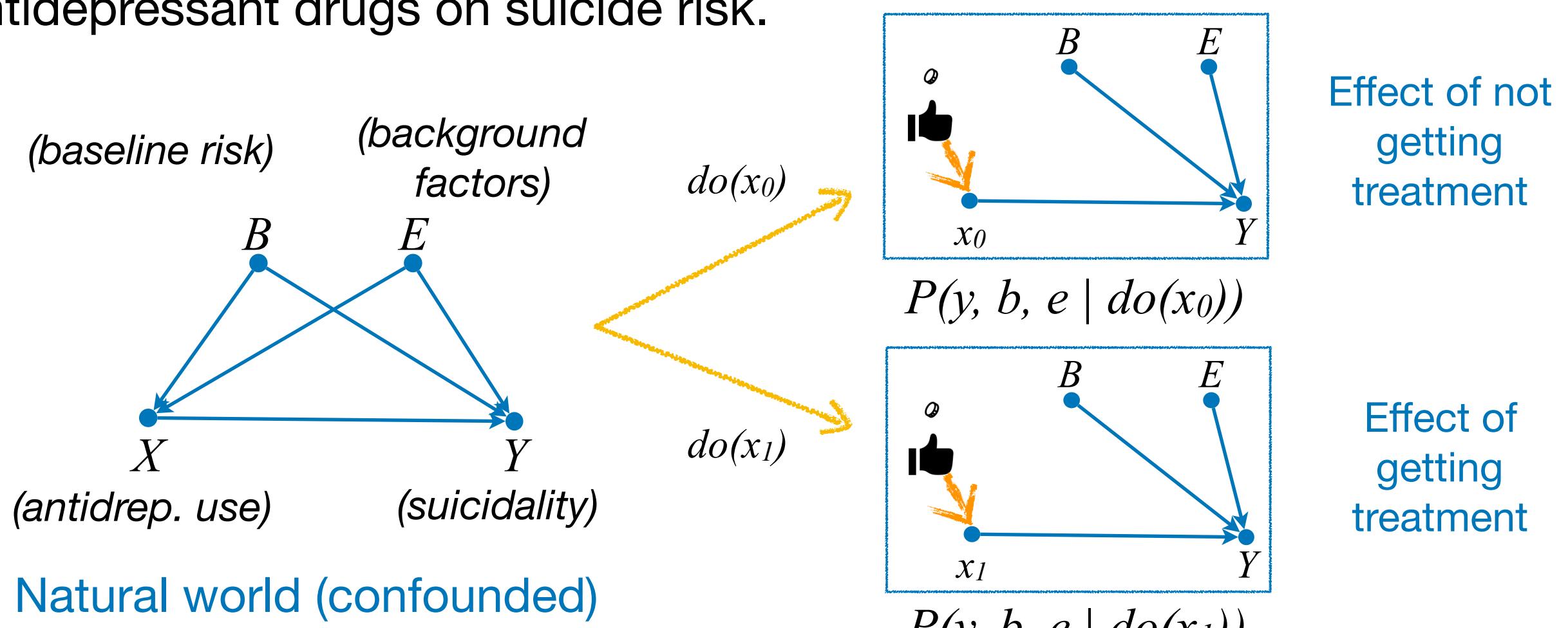
2. Selection Bias

FDA's studies sampled from a distinct population by excluding youths with elevated baseline risk for suicide (B) from their cohorts.



2 Motivating Example [Greenhouse et al. 2008]

- The FDA performed randomized controlled trials to assess the effect of antidepressant drugs on suicide risk.



- RCTs suggested that youths receiving antidepressants **had approximately twice the amount** of suicidal thoughts and behaviors compared to the control groups.

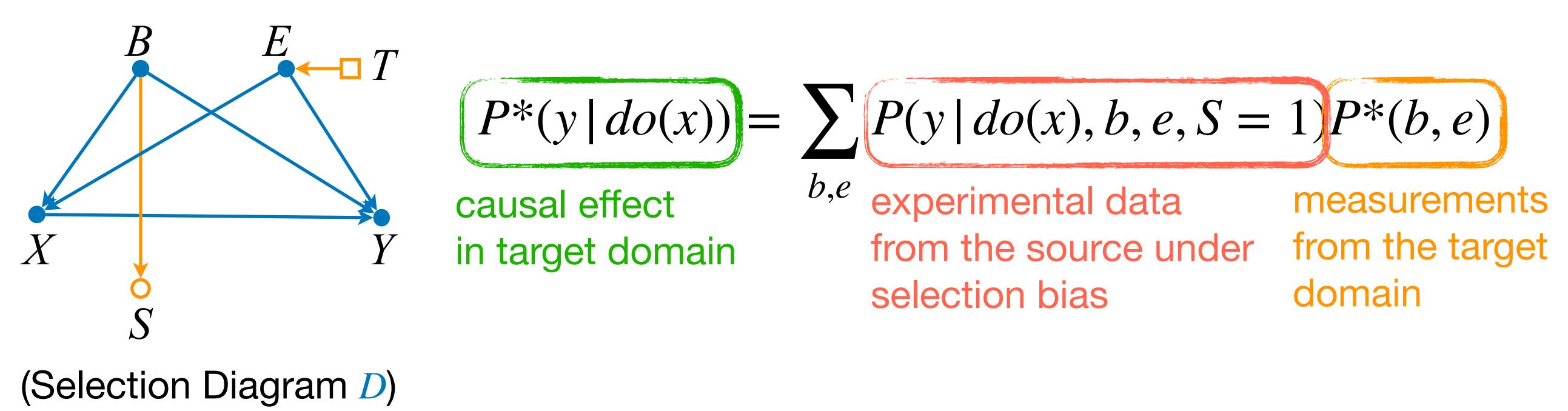
$$P(Y = 1 | do(x_1)) > P(Y = 1 | do(x_0))$$

- Results led to the addition of a **strict warning** to the drug's label.
- Surprisingly, following the warning, a **decrease in prescriptions** was reported together with **an increase of suicidal events** in the corresponding age group.

$$P^*(Y = 1 | do(x_1)) < P^*(Y = 1 | do(x_0))$$

- Several **observational studies** reported **positive results** for patients using the same antidepressants, even after accounting for access to mental health-care and other confounding factors.

$$P^*(Y = 1 | do(x_1)) < \tilde{P}^*(Y = 1 | do(x_0))$$



4 st-Adjustment

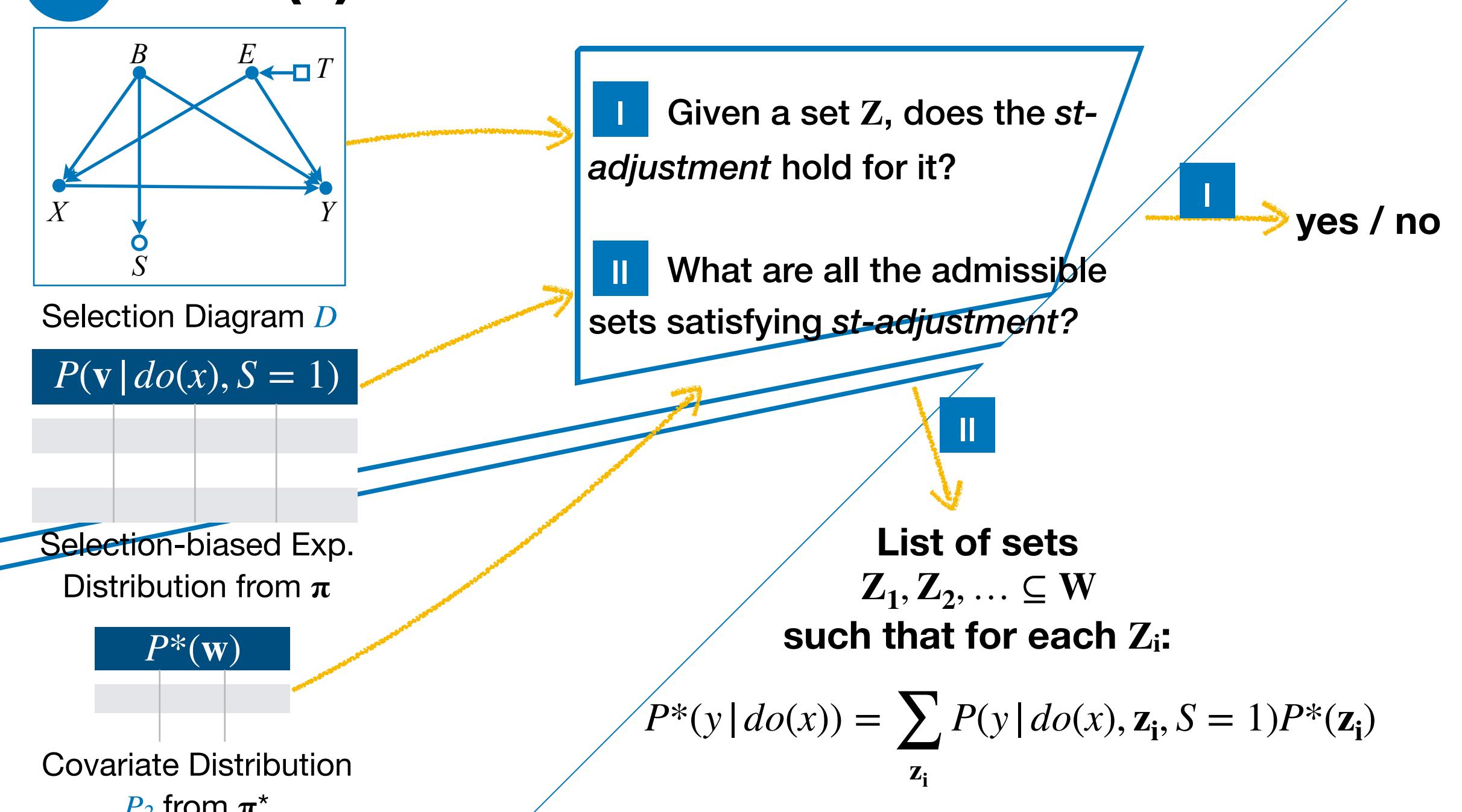
Strategy. Recalibrate the observed effect of the treatment on the outcome in the studied population using observations from the target population.

$$P^*(y | do(x)) = \sum_z P(y | do(x), z, S=1) P^*(z)$$

unbiased target z experiment results observations from in source domain π in the target domain π^*

Given a candidate set Z , special attention needs to be paid to those **affected by the treatment** that are **correlated with the outcome** given pre-treatment covariates. Call them Z_p .

5 Task(s) Overview



6 Result I: Graphical Criterion

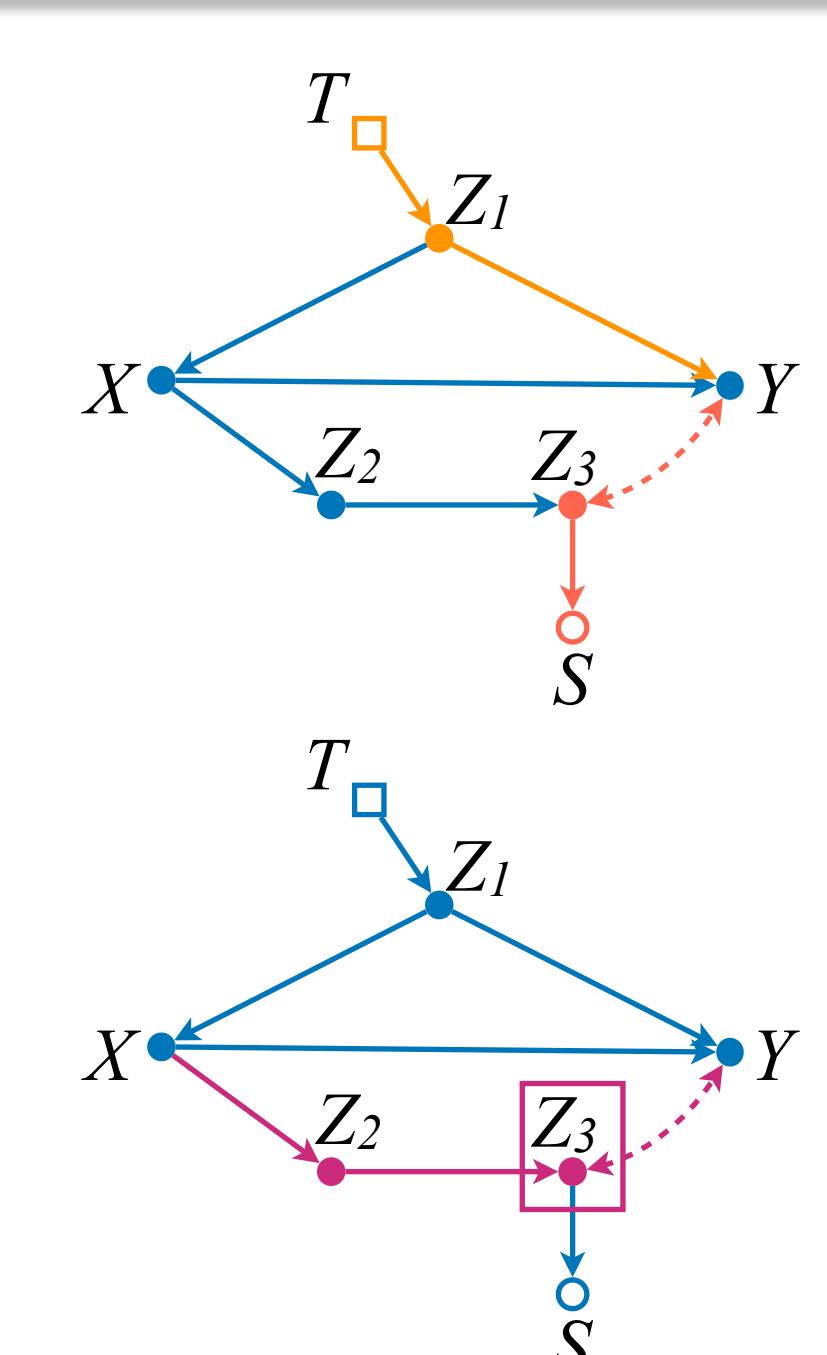
A set of covariates Z is admissible for **st-adjustment** in D relative to treatment X and outcome Y if:

- Variables in Z_p are independent of the treatment given all other covariates, and
- The outcome Y is independent of all the transportability (T) and selection bias nodes (S) given the covariates Z and the treatment X .

Thm. The causal effect $P^*(y | do(x))$ is identifiable by **st-adjustment** on a set Z with D if and only if the conditions above hold for Z relative to X and Y .

Task: Compute $P^*(y | do(x))$

- Consider the set $Z = \{Z_1, Z_2, Z_3\}$, where $Z_p = \{Z_3\}$.
- The outcome Y is affected by differences in the distribution of Z_1 between the source and target domains.
- The variable Z_3 affects the likelihood of units being sampled.
- If we adjust for Z_3 to control for selection bias, we introduce spurious correlation. Hence, we should also control for Z_2 .



7 Result II: Search Algorithm

The paper gives a procedure to list valid adjustment sets given what variables can be measured.

$$P^*(y | do(x)) = \sum_{z_1, z_2, z_3} P(y | do(x), z_1, z_2, z_3, S=1) P^*(z_1, z_2, z_3)$$