

# Back door criterion

To estimate the causal effect of  $X$  on  $Y$ , identify a set of control variables such that no descendants of  $X$  are in the control set, and all paths between  $X$  and  $Y$  that contain an arrow into  $X$  are blocked.

# Not all estimates are causal

## The Table 2 Fallacy: Presenting and Interpreting Confounder and Modifier Coefficients FREE

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### Abstract

It is common to present multiple adjusted effect estimates from a single model in a single table. For example, a table might show odds ratios for one or more exposures and also for several confounders from a single logistic regression. This can lead to mistaken interpretations of these estimates. We use causal diagrams to display the sources of the problems. Presentation of exposure and confounder effect estimates from a single model may lead to several interpretative difficulties, inviting confusion of direct-effect estimates with total-effect estimates for covariates in the model. These effect estimates may also be confounded even though the effect estimate for the main exposure is not confounded. Interpretation of these effect estimates is further complicated by heterogeneity (variation, modification) of the exposure effect measure across covariate levels. We offer suggestions to limit potential misunderstandings when multiple effect estimates are presented, including precise distinction between total and direct effect measures from a single model, and use of multiple models tailored to yield total-effect estimates for covariates.