The Back Door Criterion

by Carlos Rubio

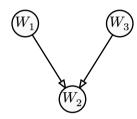
In order to understand the backdoor criterion (or admisible sets), let's review the concept of d-separation.

Consider a directed acyclic graph that contains a path p like the following:



Define S as any of set of nodes that blocks the path p: meaning, one of the arrow emiting nodes of p is in S, or S excludes at least one of the collision nodes of p and their descendant. For p, any $S \in 2^{\{W_1,W_2,W_3\}} = \{\{W_1\}, \{W_2\}, ... \{W_1, W_3\}, ...\}$ blocks p.

Another example: let p be the following path.



For $p, S = \emptyset$ blocks p, as p contains the collision node W_2 .

Two things to note about this definition:

- 1. Note that an "arrow emiting node in a path" includes collider nodes that emit arrows out of the path as well as nodes that have descendants in the path. Therefore, if a node is emiting an arrow there are only to possibilities: it has descendants in the path, or it's a collider of two other nodes.
- 2. The reasoning behind the second part of the definition is that, as long as the path includes a collider, the propagation of the effect will be restricted. I like to think about this preposition as, given a collider in a path, such collider *implicitly blocks* the path, and for purposes explained later, we will not include it.

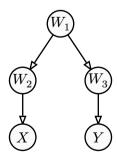
Now, consider a causal model G, were $X, Y \in G$ are considered the exposure variable and the outcome variables, respectively. A set S adjusts the G, or to be admissible for adjustment, if

- 1. None of its nodes are descendants of X, and
- 2. It blocks all the backdoor paths from Y "to" X. By backdoor, we mean all paths that include nodes X and Y, and X has an arrow point to it.

In practice, the nodes/variables in S must be observable variables. If not, it's impossible to condition on them in practise.

Note that the set S must block all paths in order to be admissible. If the set doesn't block all paths (which includes opening a path), it is not an admissible set.

Consider the path p:



Note that $p = X \leftarrow W_2 \leftarrow W_1 \rightarrow W_3 \rightarrow Y$ is the only backdoor path.

The sets S that can be consider admissible for adjustment are

- 1. $\{W_1\}$,
- 2. $\{W_2\}$

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3. \{W_3\}
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- 4. $\{W_1, W_2\}$
- 5. $\{W_1, W_3\}$
- 6. $\{W_2, W_3\}$
- 7. $\{W_1, W_2, W_3\}$

While in theory all these sets are admisible, in practice the **best option is to condition for either of the minimal sets.** Therefore, either of the first three sets is admissible.

Now, let's consider Pearl's classic butterfly causal model. I reproduced it using the causal graph simulator hosted in *dagitty.net*. I also include some of the metadata in the screenshot in order to compare my arguments with the results from the algorithm.

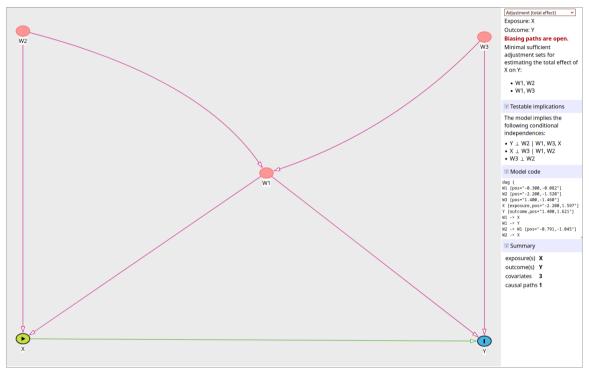


Figure 1: No adjustments made.

In this model, the backdoor baths are

- 1. $X \leftarrow W_1 \rightarrow Y$
- $2. \ X \leftarrow W_2 \to W_1 \to Y$
- 3. $X \leftarrow W_1 \leftarrow W_3 \rightarrow Y$

Let's consider we might consider $S=\{W_2\}$ or $S=\{W_3\}$ for adjustment. This sets only block 2 out of 3 backdoor paths. Then, we might consider the set $S=\{W_1\}$ for adjustment, as it blocks all the foremention paths. But in doing so it opens the path $p=X\leftarrow W_2\to W_1\leftarrow W_3\to Y$ that was being *implictly blocked by* the collider W_1 ,

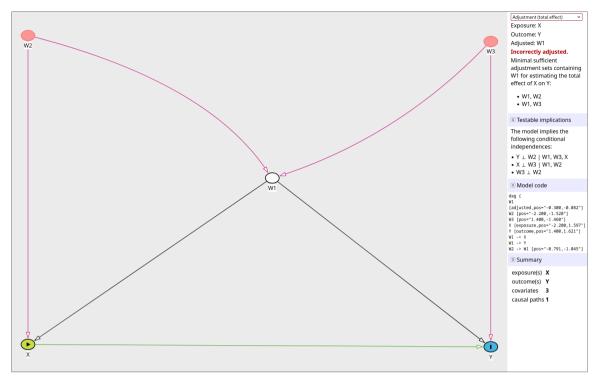


Figure 2: W_1 is insufficient.

The solution is blocking p without having to exclude W_1 from S. We can do this by including either W_2 or W_3 .

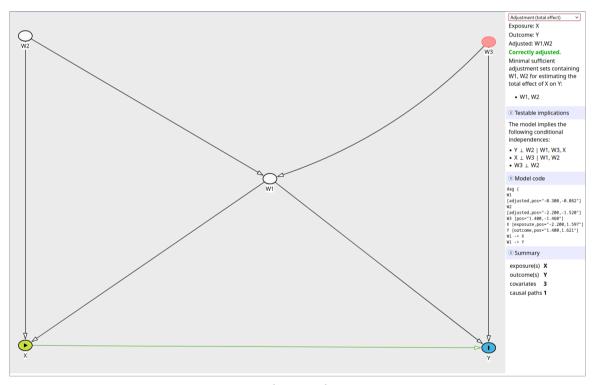


Figure 3: $S=\{W_1,W_2\}$ is admissible.