Suppose that experiment E_1 showed that x_2 and x_1 were independent, such that the ordering between x_1 and x_2 remains underdetermined.

The only experiment that establishes whether $x_1 \rightarrow x_2$ are experiments E_2 with x_1 in J_2 and x_2 not in J_2 .

Experiments E_1 and E_2 resolve the order between x_1 and x_2 , suppose without loss of generality that it is $x_1 \to x_2$. In the worst case this required two experiments.

Now for the remainder: the only experiment that establishes whether $x_1 \rightarrow x_3$ are experiments E_3 with x_1 and x_2 in J_3 and x_3 not in J_3 . Note that none of the previous experiments could have been an E_3 .

The only experiment that establishes whether $x_1 \to x_4$ are experiments E_4 with x_1 , x_2 , x_3 in J_4 and x_4 not in J_4 . None of the previous experiments could have been an E_4 .

. . .

The only experiment that establishes whether $x_1 \to x_n$ is an experiment E_n with x_1, \ldots, x_{n-1} in J_n and x_n not in J_n . None of the previous experiments could have been an E_n . It follows that n experiments are in the worst case necessary to discover the causal structure. QED

The above proof shows that in the worst case a sequence of n experiments is necessary that have intervention sets that intervene on at least n-i variables simultaneously for each integer i in 1 < i < n.

Appendix B

Parameterization PM1 for structure 1 in figure 2 (all variables are binary):

```
p(u = 1) = .5
                                     p(z = 1|u = 1, v = 1, x = 1, y = 1) = .8
                                     p(z = 1|u = 1, v = 1, x = 1, y = 0) = .8
                                     p(z = 1 | u = 1, v = 1, x = 0, y = 1) = .84
p(v = 1) = .5
                                     p(z = 1|u = 1, v = 1, x = 0, y = 0) = .8
                                     p(z = 1|u = 1, v = 0, x = 1, y = 1) = .8
                                     p(z = 1|u = 1, v = 0, x = 1, y = 0) = .8
p(x = 1|u = 1) = .8
                                     p(z = 1|u = 1, v = 0, x = 0, y = 1) = .64
p(x = 1|u = 0) = .2
                                     p(z = 1|u = 1, v = 0, x = 0, y = 0) = .8
                                     p(z = 1|u = 0, v = 1, x = 1, y = 1) = .8
                                     p(z = 1|u = 0, v = 1, x = 1, y = 0) = .8
p(y = 1|v = 1, x = 1) = .8
                                     p(z = 1|u = 0, v = 1, x = 0, y = 1) = .79
p(y = 1|y = 1, x = 0) = .8
                                     p(z = 1 | u = 0, v = 1, x = 0, y = 0) = .8
p(y = 1|y = 0, x = 1) = .8
                                     p(z = 1|u = 0, v = 0, x = 1, y = 1) = .8
p(y = 1|y = 0, x = 0) = .2
                                     p(z = 1|u = 0, v = 0, x = 1, y = 0) = .2
                                     p(z = 1 | u = 0, v = 0, x = 0, y = 1) = .84
                                     p(z = 1|u = 0, v = 0, x = 0, y = 0) = .2
```

Parameterization PM2 for structure 2 in figure 2:

$$\begin{array}{l} p(u=1)=.5\\ p(v=1)=.5\\ p(x=1|u=1)=.8\\ p(x=1|u=0)=.2\\ p(y=1|v=1, x=1)=.8\\ p(y=1|v=0, x=1)=.8\\ p(y=1|v=0, x=0)=.8\\ p(y=1|v=0, x=0)=.8\\ p(y=1|v=0, x=0)=.8\\ p(y=1|v=0, x=0)=.2\\ \end{array}$$

Passive observational distribution:

PM1:
$$P(X, Y, Z) = \sum_{uv} P(U)P(V)P(X|U)P(Y|V, X)P(Z|U, V, X, Y)$$

PM2: $P(X, Y, Z) = \sum_{uv} P(U)P(V)P(X|U)P(Y|V, X)P(Z|U, V, Y)$

Experimental distribution when x is subject to an intervention (I write P(A|B||B) to mean the conditional probability of A given B in an experiment where B has been subject to a surgical intervention):

PM1:
$$P(Y,Z|X||X) = \sum_{uv} P(U)P(V)P(Y|V,X)P(Z|U,V,X,Y)$$

PM2: $P(Y,Z|X||X) = \sum_{uv} P(U)P(V)P(Y|V,X)P(Z|U,V,Y)$

Experimental distribution when y is subject to an intervention:

PM1:
$$P(X,Z|Y||Y) = \sum_{uv} P(U)P(V)P(X|U)P(Z|U,V,X,Y)$$

PM2: $P(X,Z|Y||Y) = \sum_{uv} P(U)P(V)P(X|U)P(Z|U,V,Y)$

Experimental distribution when z is subject to an intervention:

PM1:
$$P(X, Y|Z||Z) = \sum_{uv} P(U)P(V)P(X|U)P(Y|V,X)$$

PM2: $P(X, Y|Z||Z) = \sum_{uv} P(U)P(V)P(X|U)P(Y|V,X)$

By substituting the terms of PM1 and PM2 in the above equations it can be verified that PM1 and PM2 have identical passive observational and single-intervention distributions but that they differ for the following double-intervention distribution on *x* and *y*.

Experimental distribution when x and y are subject to an intervention:

PM1:
$$P(Z|X,Y||X,Y) = \sum_{uv} P(U)P(V)P(Z|U,V,X,Y)$$

PM2: $P(Z|X,Y||X,Y) = \sum_{uv} P(U)P(V)P(Z|U,V,Y)$

PM1 and PM2 (unsurprisingly) have identical distributions for the other two double-intervention distributions, since the $x \to z$ edge is broken and the remaining parameters are identical in the parameterizations:

Experimental distribution when x and z are subject to an intervention:

PM1:
$$P(Y|X,Z||X,Z) = \sum_{v} P(V)P(Y|V,X)$$

PM2: $P(Y|X,Z||X,Z) = \sum_{v} P(V)P(Y|V,X)$

Experimental distribution when y and z are subject to an intervention:

PM1:
$$P(X|Y,Z||Y,Z) = \sum_{u} P(U)P(X|U)$$

PM2: $P(X|Y,Z||Y,Z) = \sum_{u} P(U)P(X|U)$

Appendix C

Parameterization PM3 for structure 1 in figure 2:

$$p(u = 1) = .5$$

$$p(z = 1 | u = 1, v = 1, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 1, v = 1, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 1, v = 1, x = 0, y = 1) = .8$$

$$p(z = 1 | u = 1, v = 1, x = 0, y = 1) = .8$$

$$p(z = 1 | u = 1, v = 0, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 1, v = 0, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 1, v = 0, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 1, v = 0, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 1, v = 0, x = 0, y = 1) = .8$$

$$p(z = 1 | u = 1, v = 0, x = 0, y = 1) = .8$$

$$p(z = 1 | u = 0, v = 1, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 0, v = 1, x = 1, y = 0) = .8$$

$$p(z = 1 | u = 0, v = 1, x = 0, y = 0) = .8$$

$$p(z = 1 | u = 0, v = 1, x = 0, y = 0) = .8$$

$$p(z = 1 | u = 0, v = 0, x = 1, y = 1) = .9$$

$$p(z = 1 | u = 0, v = 0, x = 1, y = 0) = .2$$

$$p(z = 1 | u = 0, v = 0, x = 0, y = 1) = .8$$

$$p(z = 1 | u = 0, v = 0, x = 0, y = 1) = .8$$

$$p(z = 1 | u = 0, v = 0, x = 0, y = 0) = .2$$

Substituting the parameters of PM3 in the equations for the passive observational or any experimental distributions of PM1 in appendix 2, it can be