

The Causality Lab

Application: www.phil.cmu.edu/projects/causality-lab

Web Version: <http://oli.web.cmu.edu>

- Create New Account
- Course admit code: **csrdemo05**

Outline

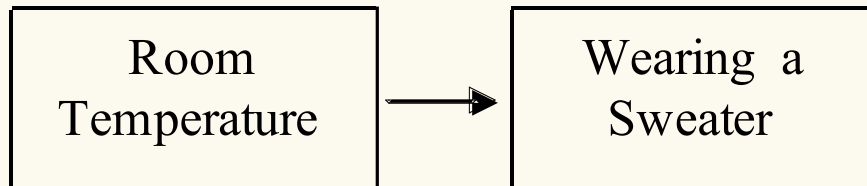
Yesterday

1. The Curriculum
2. The Online Course
 - Modules
 - Causality Lab
 - Case Studies
3. Learning Studies

Today

1. The Causality Lab in Detail
2. Hands On Work
 - Doing Exercises
 - Authoring Exercises
3. Pilot Studies

Simple Bayes Net



$$P(RT = <55) = .1$$

$$P(RT = 55 - 85) = .8$$

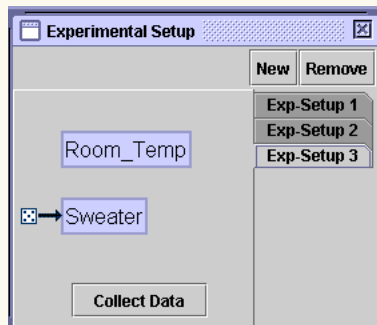
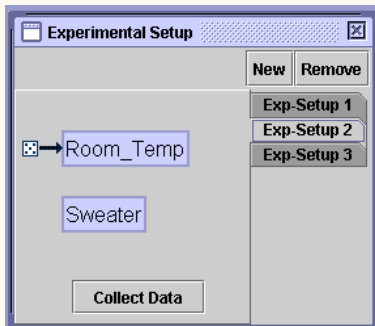
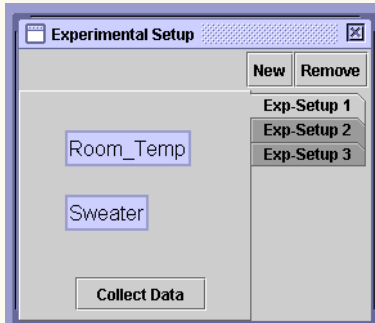
$$P(RT = >85) = .1$$

$$P(\text{Wearing a Sweater} \mid RT < 55) = .98$$

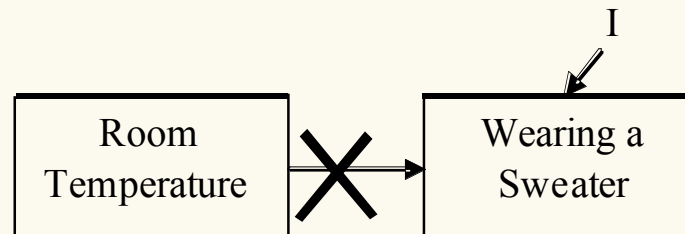
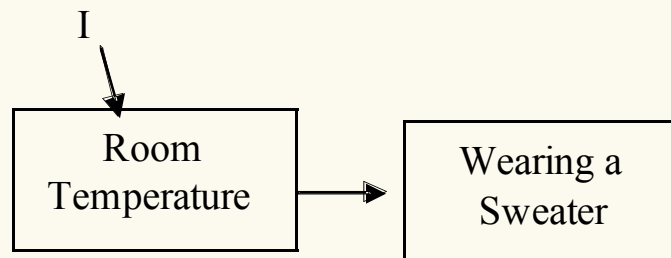
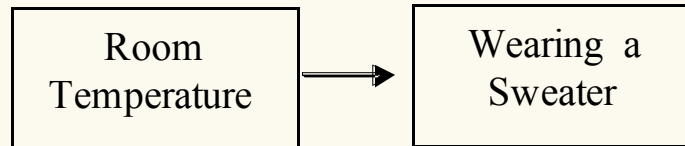
$$P(\text{Wearing a Sweater} \mid RT = 55 - 85) = .5$$

$$P(\text{Wearing a Sweater} \mid RT > 85) = .04$$

Exp. Setup



Manipulated Graph



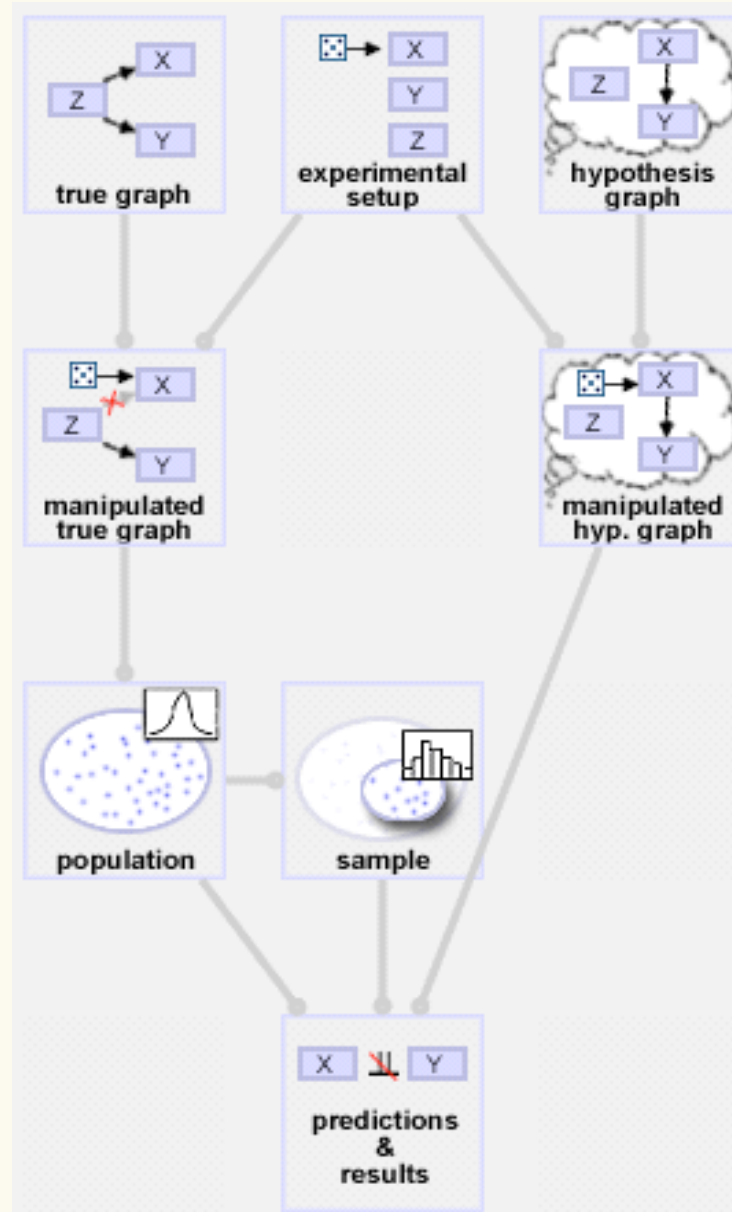
Population

$$\begin{aligned}
 P(RT = <55) &= .1 \\
 P(RT = 55-85) &= .8 \\
 P(RT = >85) &= .1 \\
 P(\text{Wearing a Sweater} \mid RT < 55) &= .98 \\
 P(\text{Wearing a Sweater} \mid RT = 55-85) &= .5 \\
 P(\text{Wearing a Sweater} \mid RT > 85) &= .04
 \end{aligned}$$

$$\begin{aligned}
 P(RT = <55 \mid \mathbf{I}) &= .33 \\
 P(RT = 55-85 \mid \mathbf{I}) &= .33 \\
 P(RT = >85 \mid \mathbf{I}) &= .33 \\
 P(\text{Wearing a Sweater} \mid RT < 55) &= .98 \\
 P(\text{Wearing a Sweater} \mid RT = 55-85) &= .5 \\
 P(\text{Wearing a Sweater} \mid RT > 85) &= .04
 \end{aligned}$$

$$\begin{aligned}
 P(RT = <55) &= .1 \\
 P(RT = 55-85) &= .8 \\
 P(RT = >85) &= .1 \\
 P(\text{Wearing a Sweater} \mid \mathbf{I}) &= .5
 \end{aligned}$$

Causality Lab: Navigation Panel



Causality Lab – Pilot Studies

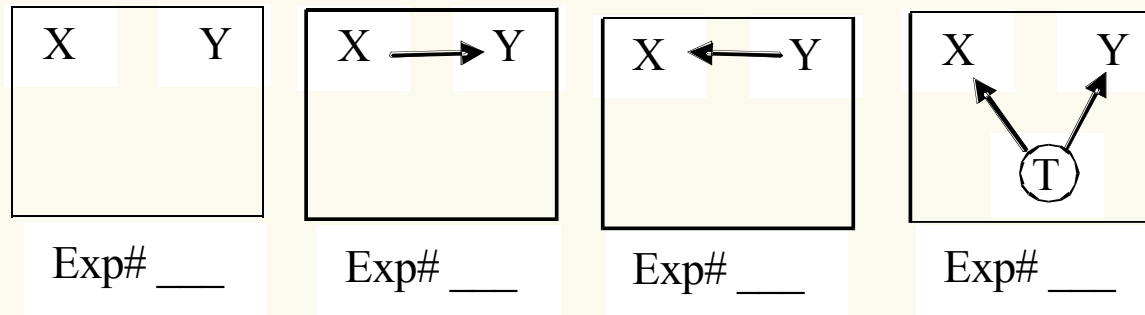
- Carnegie Mellon class (15 students): Causation and Social Policy
- 4-weeks so far through online course
 - Causal graphs,
 - Independence and Conditional Independence
 - d-separation
- No exposure to equivalence classes or methods for discovering structure

Causality Lab – Pilot Studies

- Students given set of possible true models
- Students setup experiments
- Given independence results directly
 - no sample data
 - no statistical inference needed
- Tasks:
 - Infer correct model in minimum number of experiments
 - Infer set of models consistent with experiments so far

Training Experiment

True Model one of:

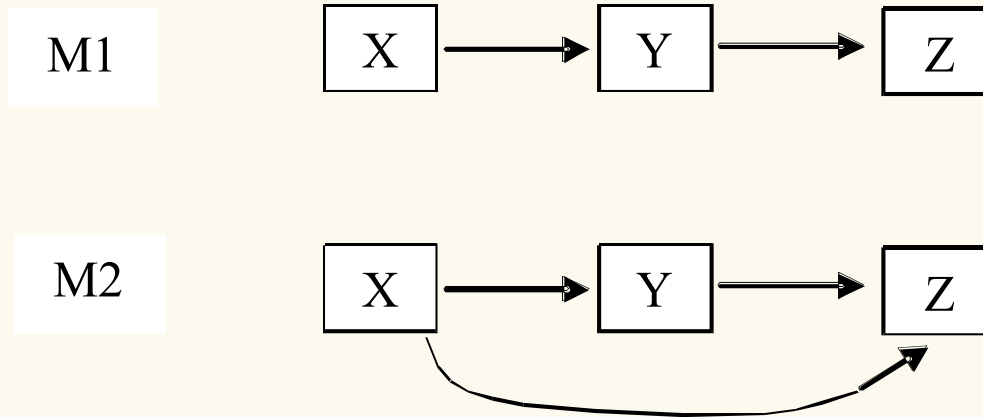


Student Task:

- Do Passive Observation First – eliminate inconsistent models
- Discover True model in fewest possible experiments after that

Experiment 1

True Model one of:



Student Task:

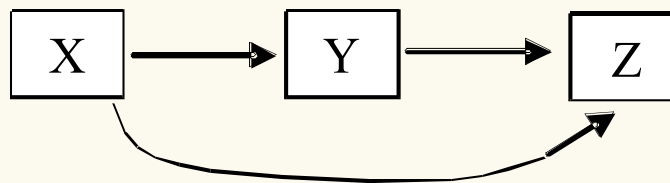
Discover True model in fewest possible experiments

Experiment 1

M1



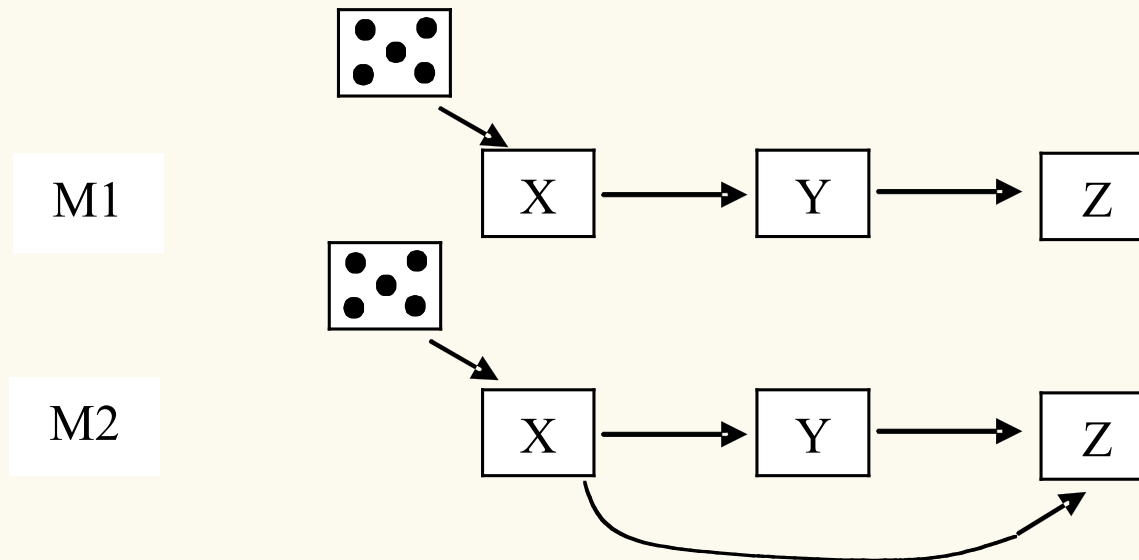
M2



Passive Observation:

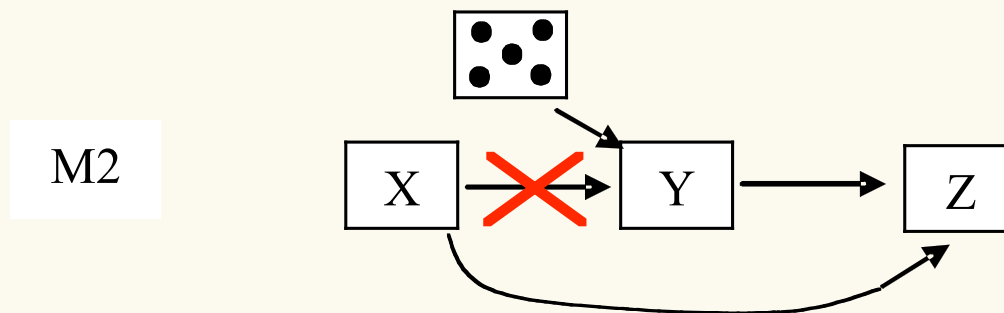
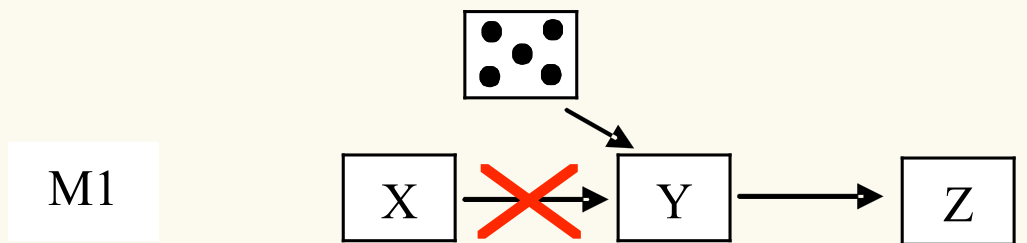
$X \perp\!\!\!\perp Z \mid Y ?$

Experiment 1



Intervene on X: $X \perp\!\!\!\perp Z \mid Y ?$

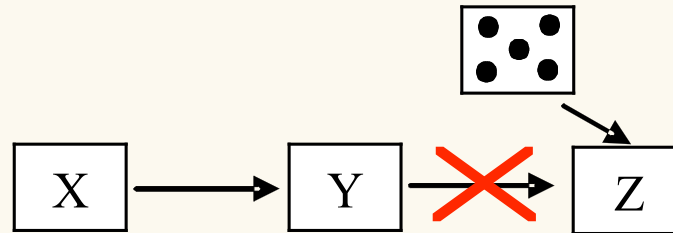
Experiment 1



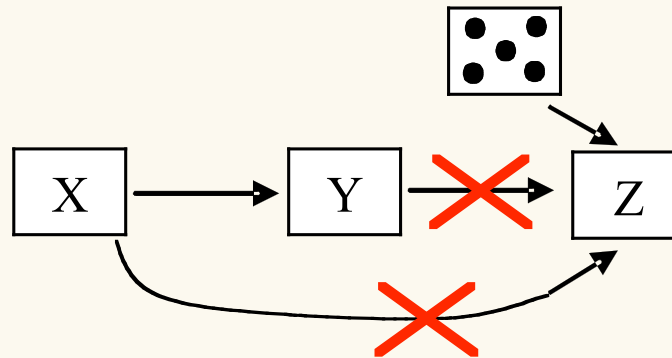
Intervene on Y: $X _||_ Z \ ?$

Experiment 1

M1



M2

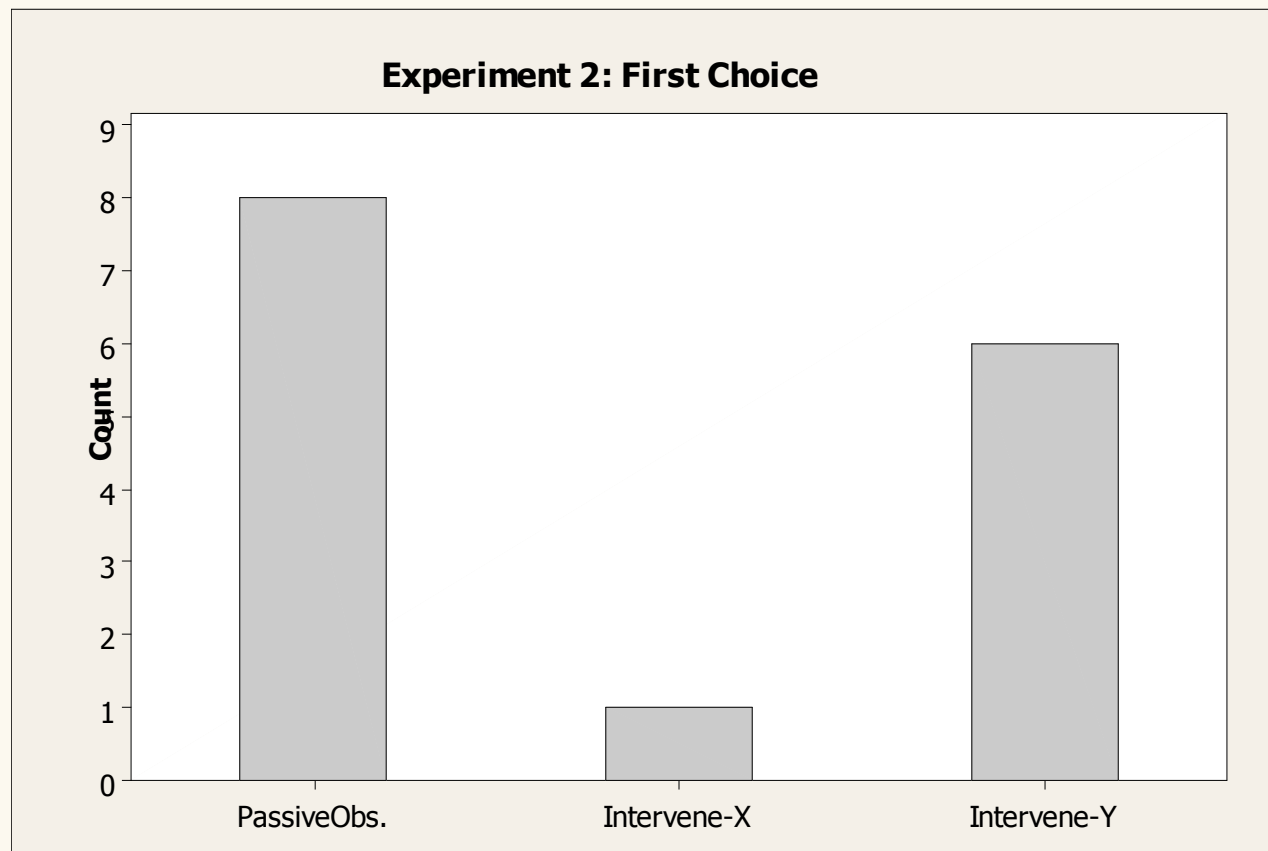


Intervene on Z:

Indistinguishable Models

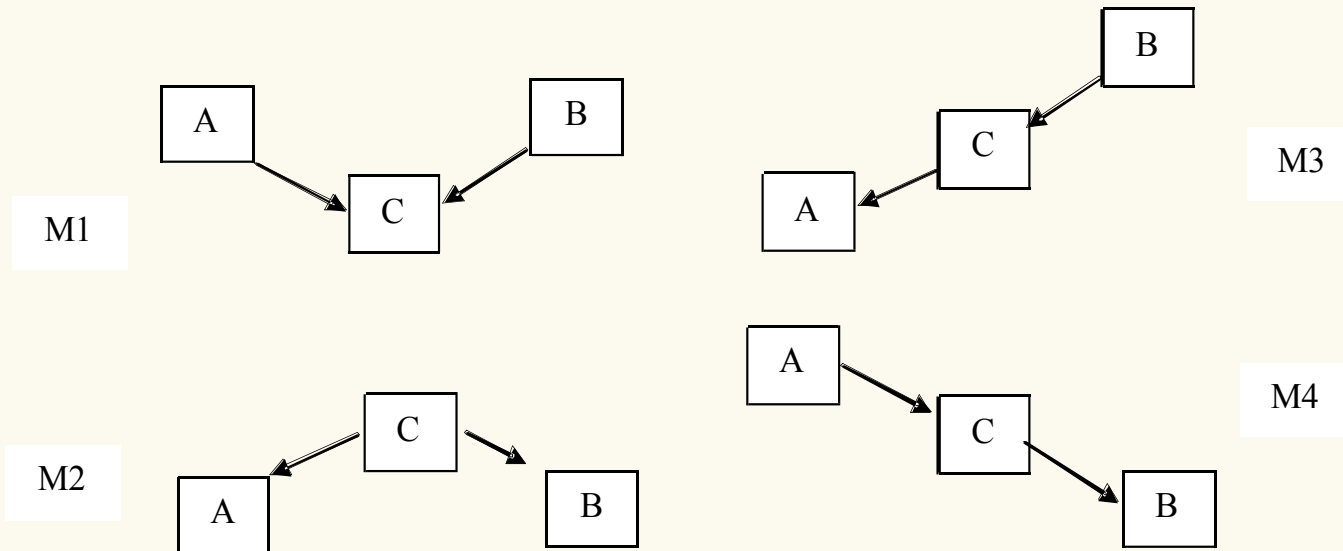
Experiment 1: Results

Correct Answer: 15/15



Experiment 2

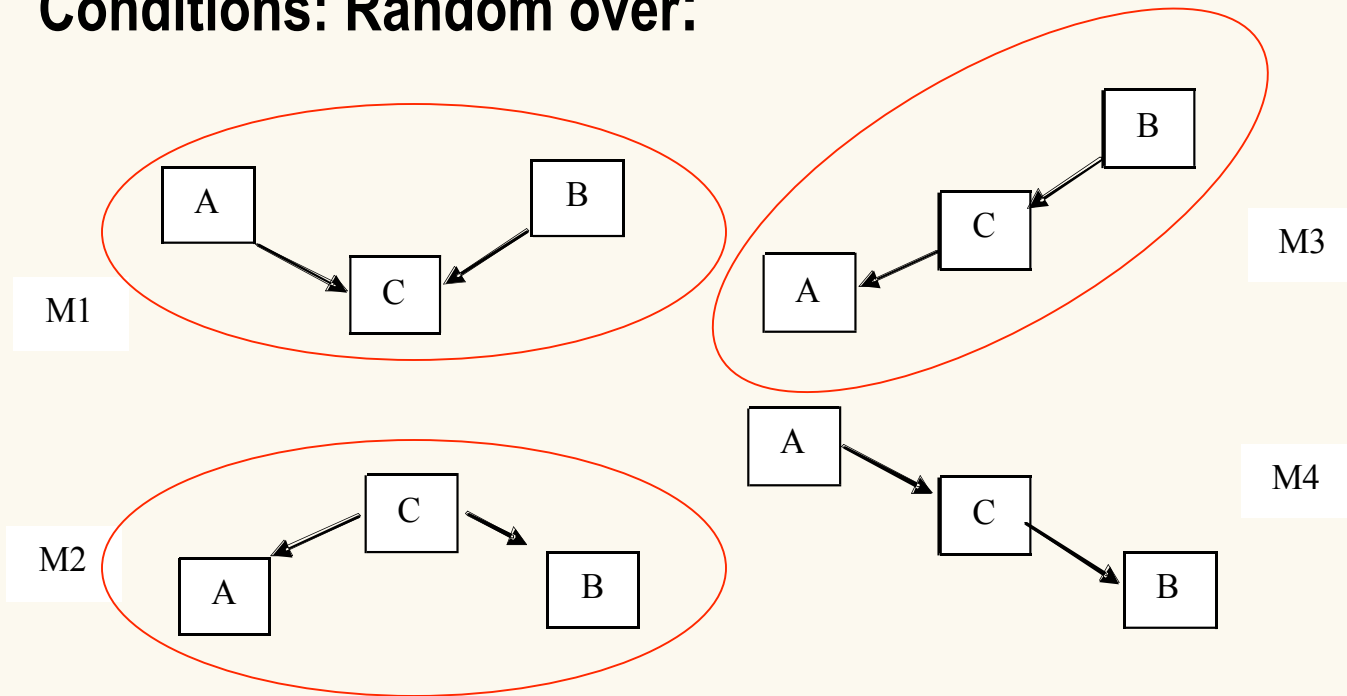
True Model one of:



Task: Discover True model in fewest possible experiments

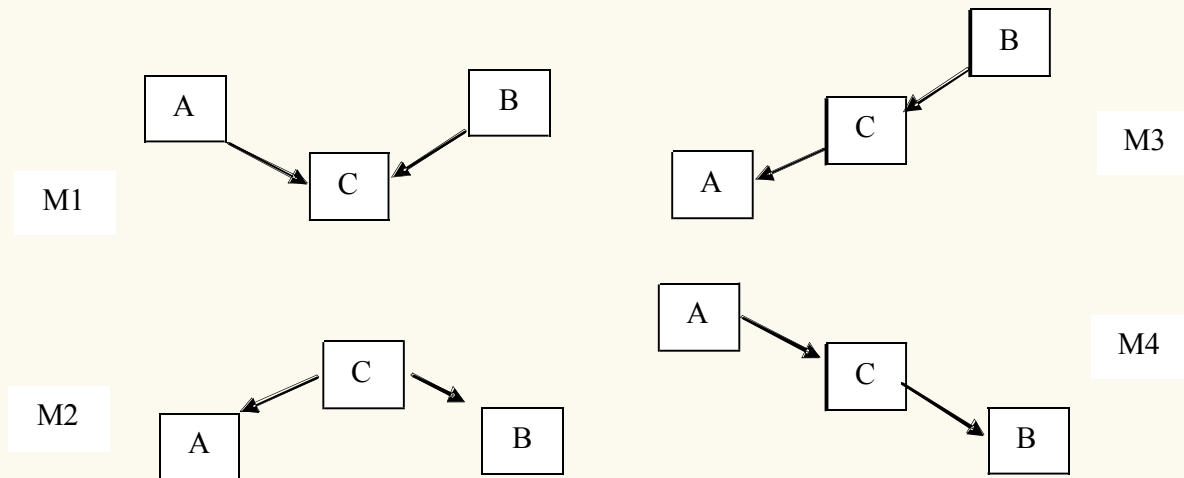
Experiment 1

Conditions: Random over:



Experiment 2

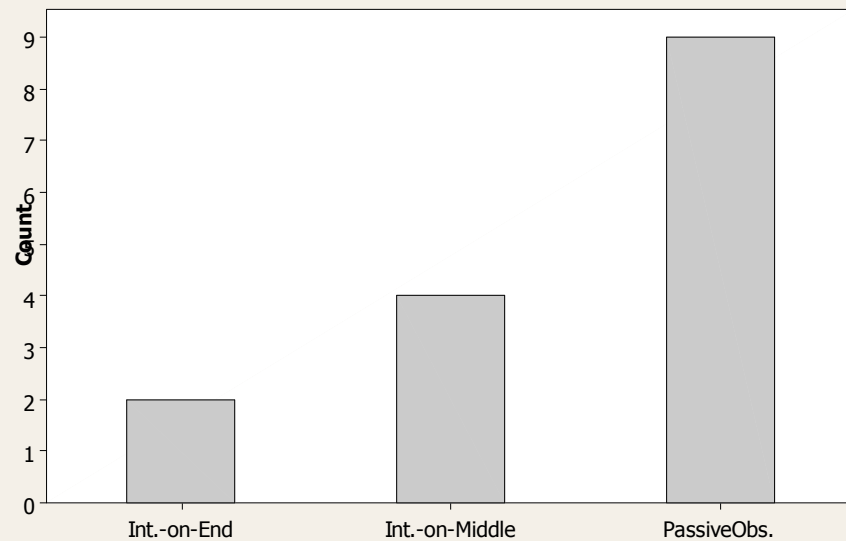
Experimental Setup	Distinguishable?
Passive Observation	M1 from {M2, M3, M4}
Randomize A	M1 from {M2, M3} from M4
Randomize B	M1 from {M2, M4} from M3
Randomize C	M1 from M2 from M3 from M4



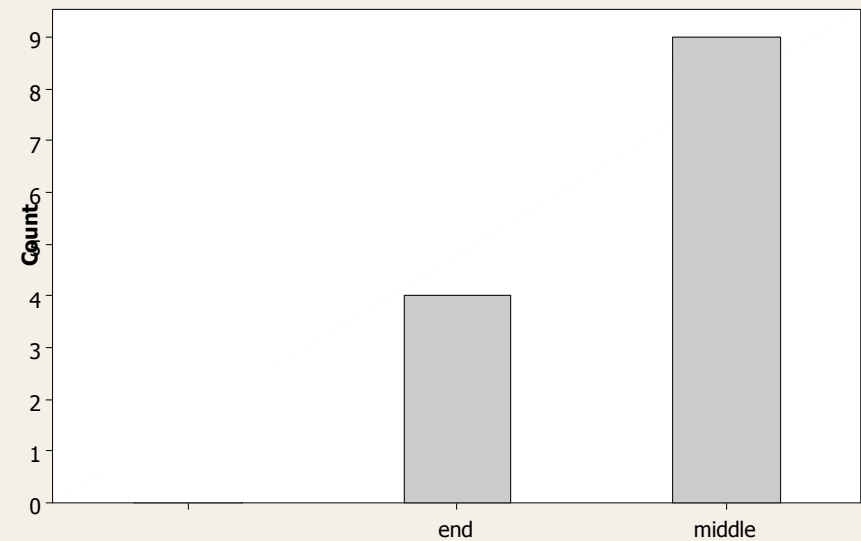
Experiment 2: Results

Correct Answer: 14/15

Experiment 2: First Choice

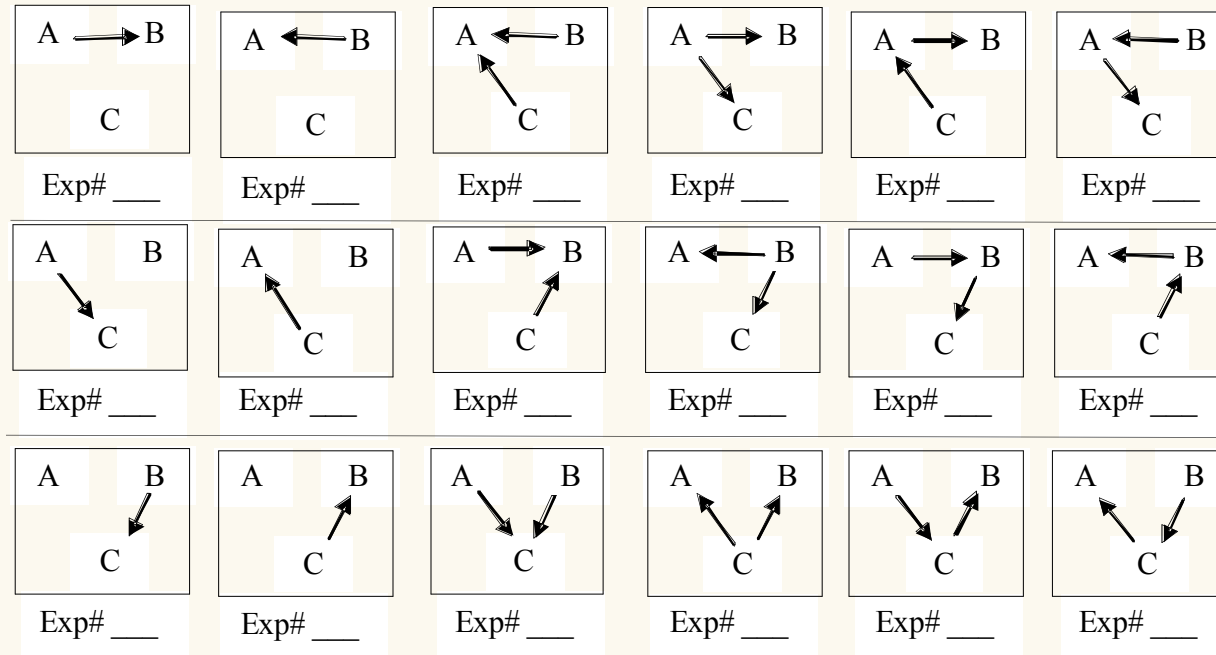


Experiment 2: First Intervention Choice



Experiment 3

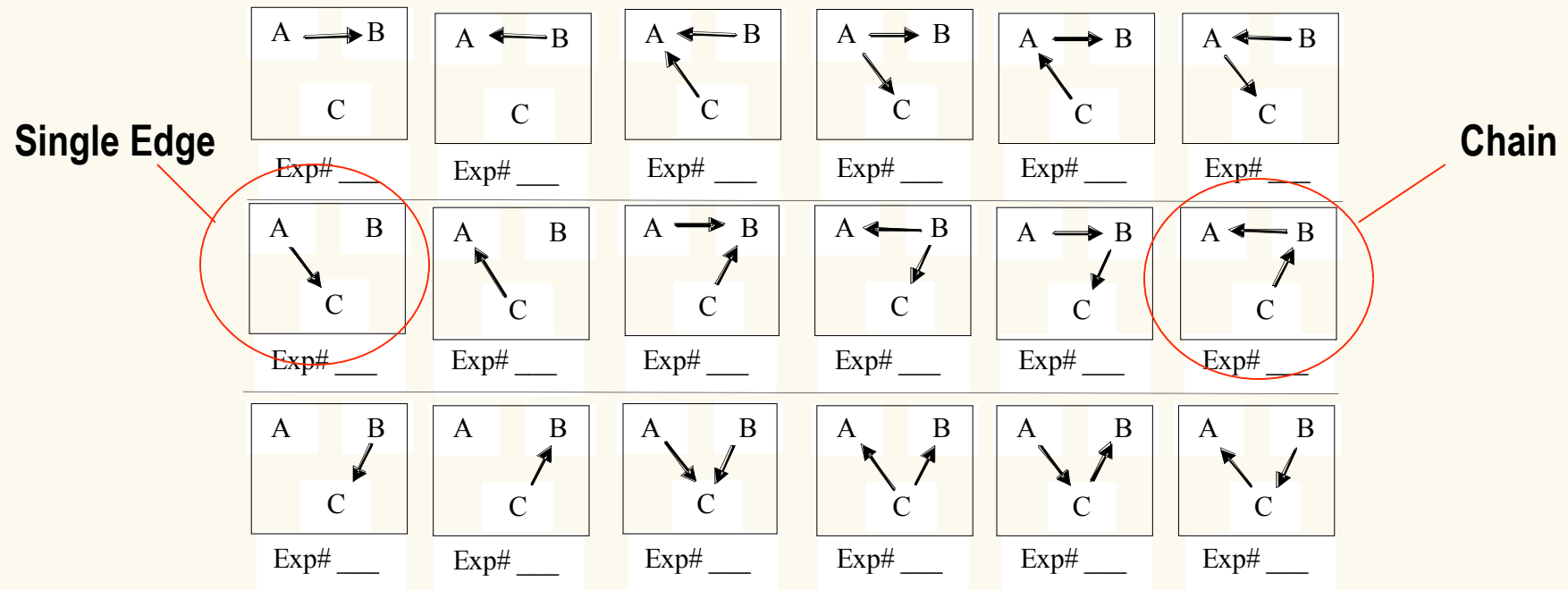
True Model one of:



Student Task:

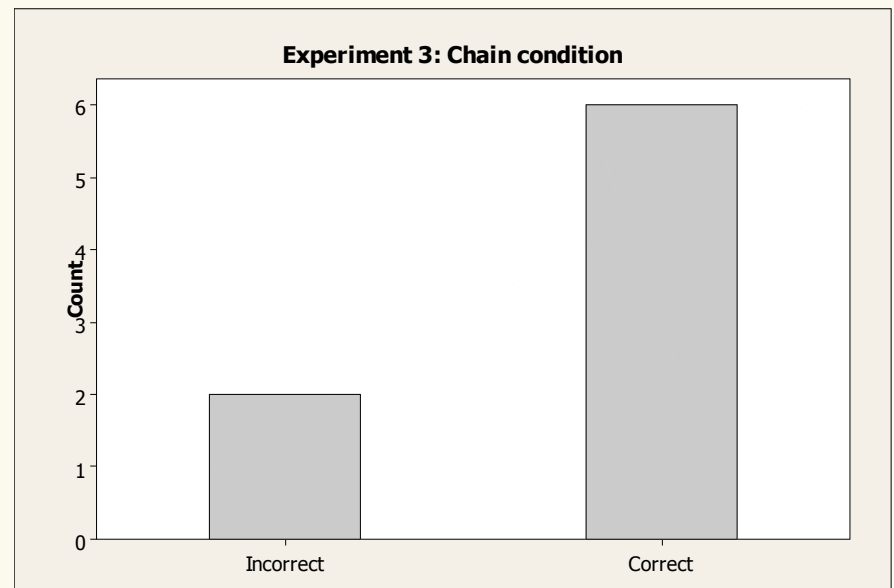
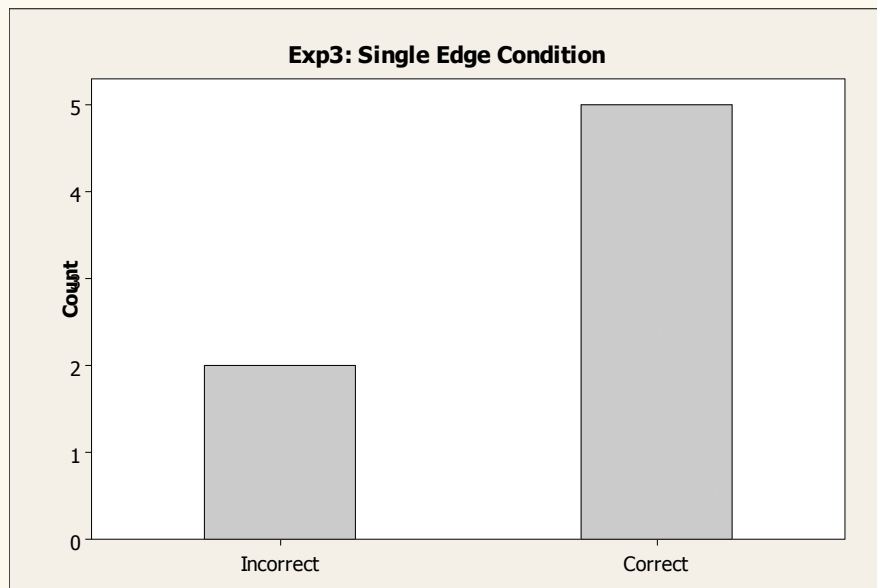
- 1) Passive Observation: eliminate inconsistent models
- 2) Discover True model in fewest possible experiments thereafter

Experiment 3: Conditions

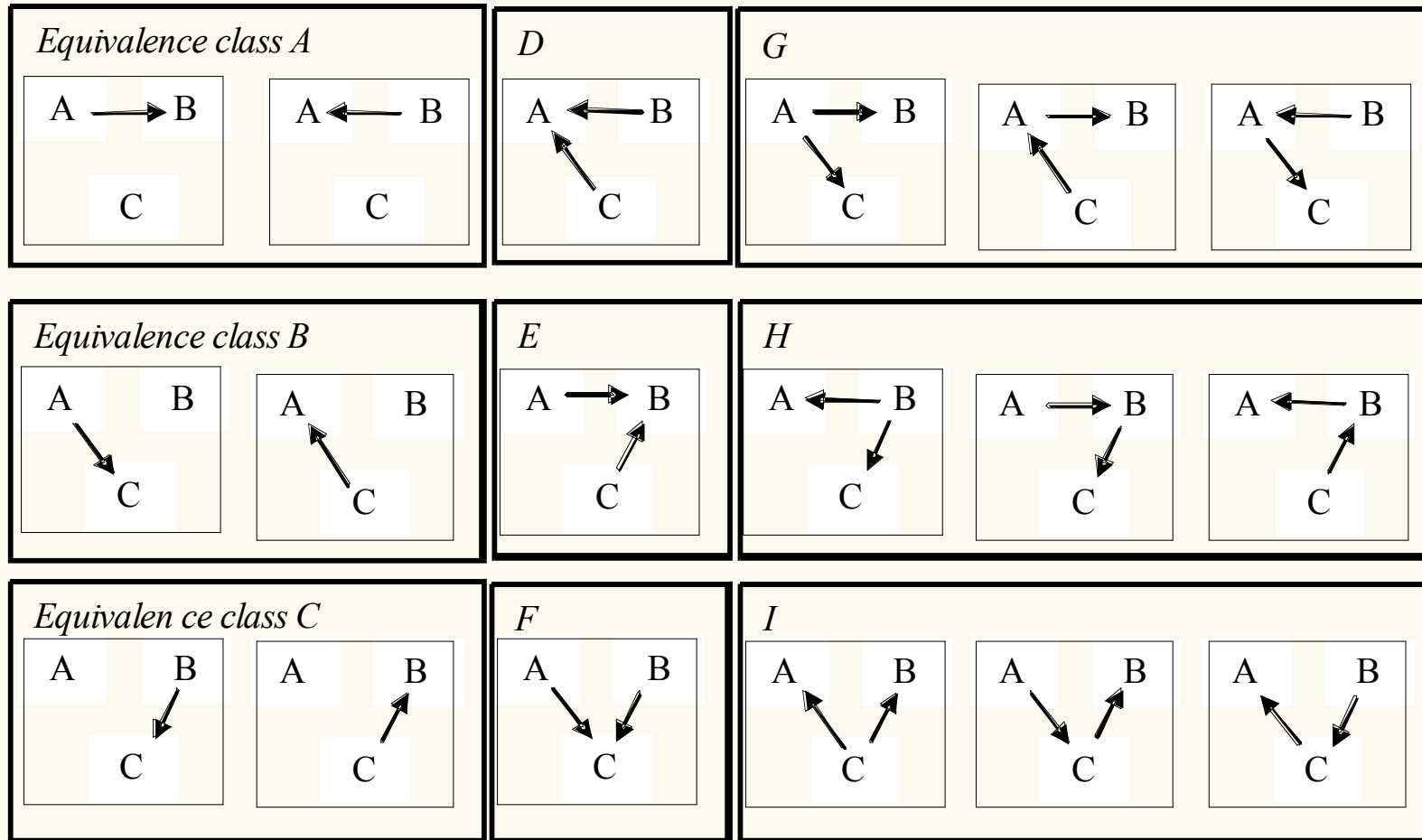


Experiment 3: Results

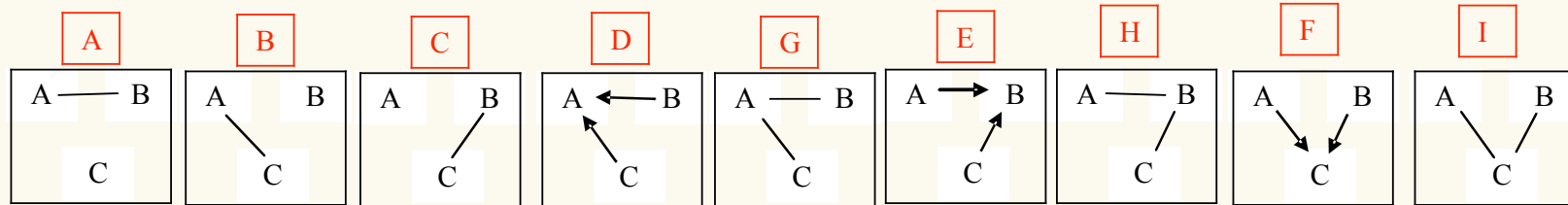
Overall Success



Experiment 3 : Equivalence under Passive Observation



Experiment 3: Equivalence Class Integrity



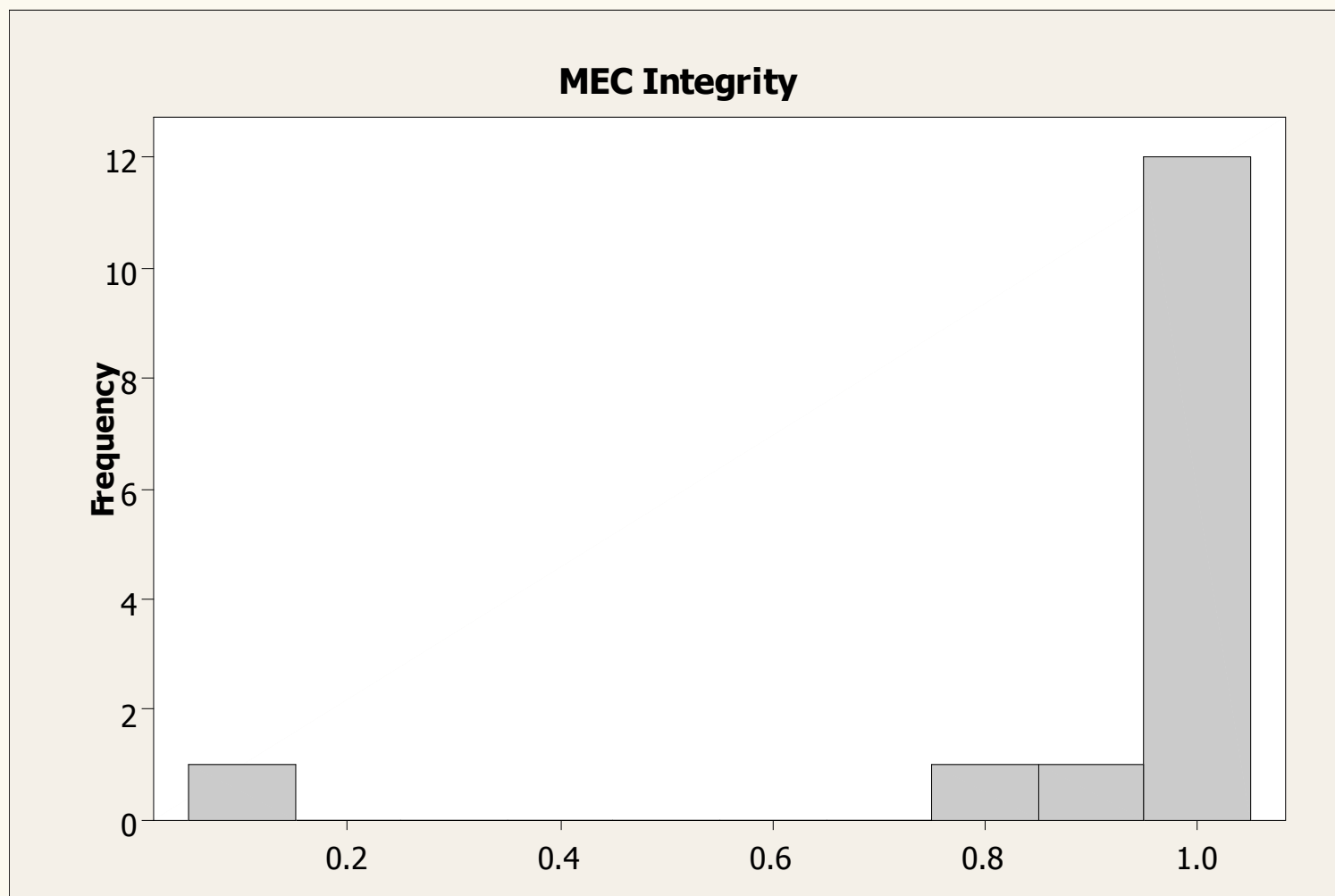
After Passive Observation:

Students who understand equivalence should either

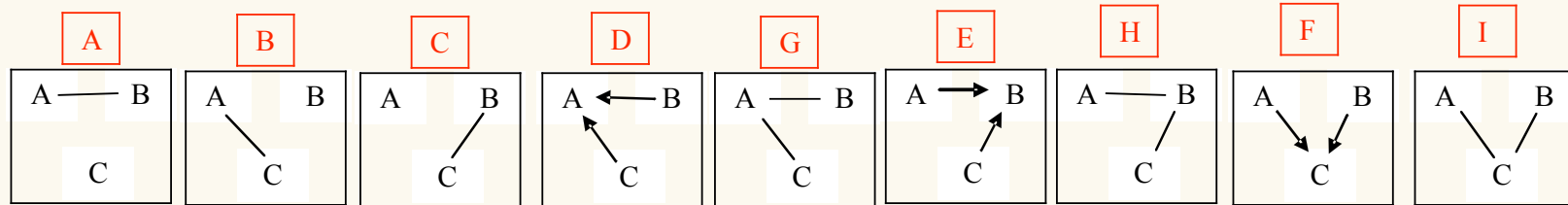
- Keep all models in an equivalence class, or
- Remove all models in an equivalence class

$$MEC-Integrity = \sum_{mec \in \{A,B,C,G,H,I\}} \begin{cases} |mec| & : \text{if all models in } mec \text{ were} \\ & \text{included or all excluded:} \\ 0 & : \text{otherwise} \end{cases}$$

Experiment 3: Equivalence Class Integrity Results



Experiment 3: Commission and Omission



$$\text{Commission Error} = \frac{\text{\# of graphs retained by student, but not in correct MEC}}{\text{\# of graphs not in correct MEC}}$$

$$\text{Omission Error} = \frac{\text{\# of graphs in the correct MEC omitted by the student}}{\text{\# of graphs in the correct MEC}}$$

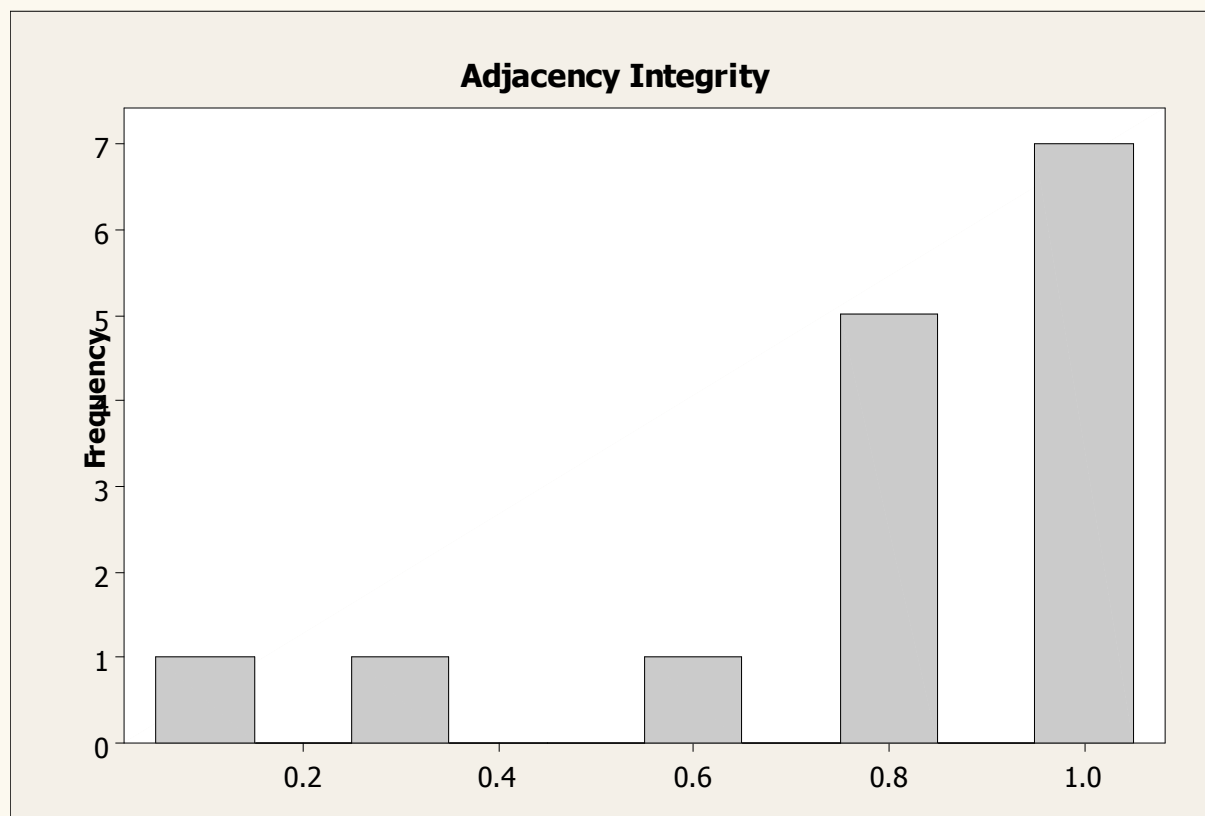
Experiment 3: Equivalence Class Integrity Results



Experiment 3: Adjacency Integrity

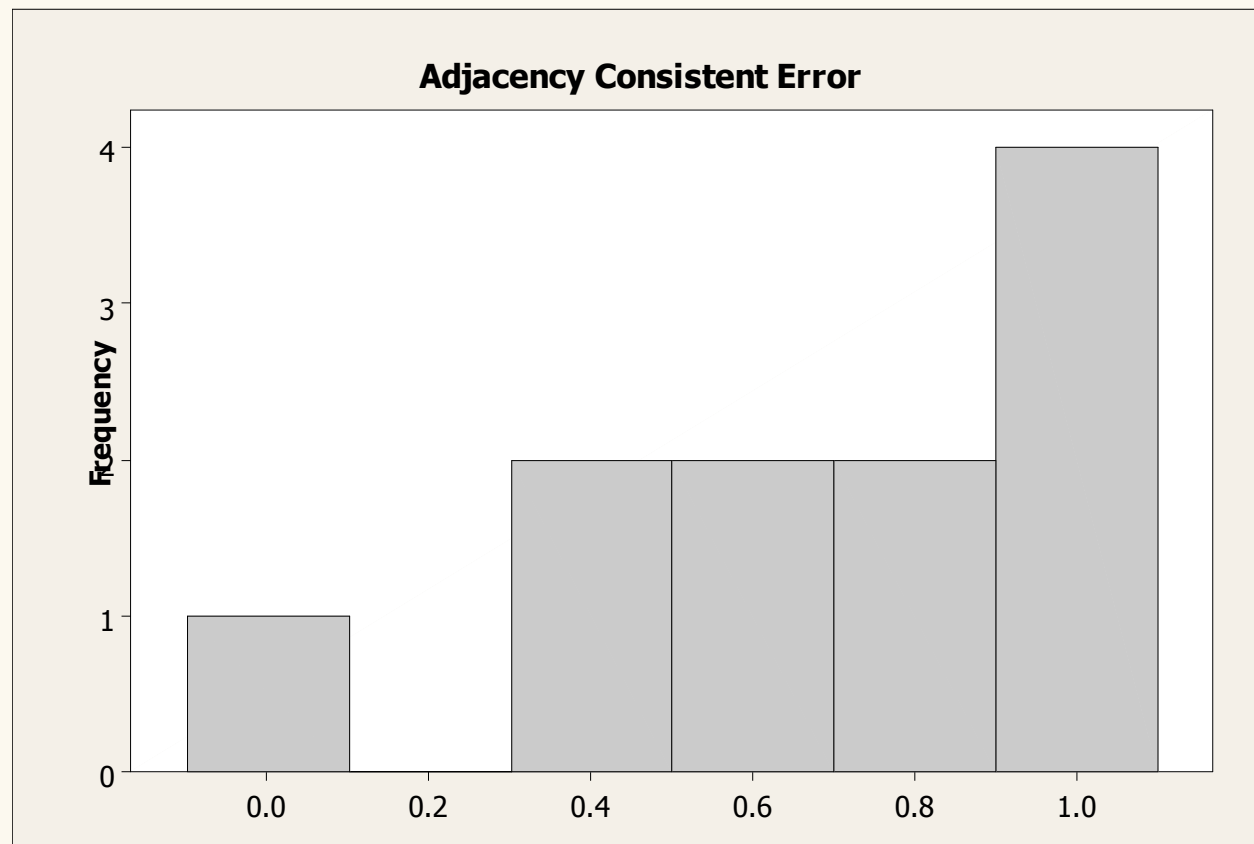
$$Adjacency-Integrity = \sum_{adj \in \{A,B,C, D+G,E+H,F+I\}} \begin{cases} |adj| & : \text{if all models in } adj \text{ were} \\ & \text{included or all excluded:} \\ 0 & : \text{otherwise} \end{cases}$$

18



Experiment 3: Adjacency Integrity

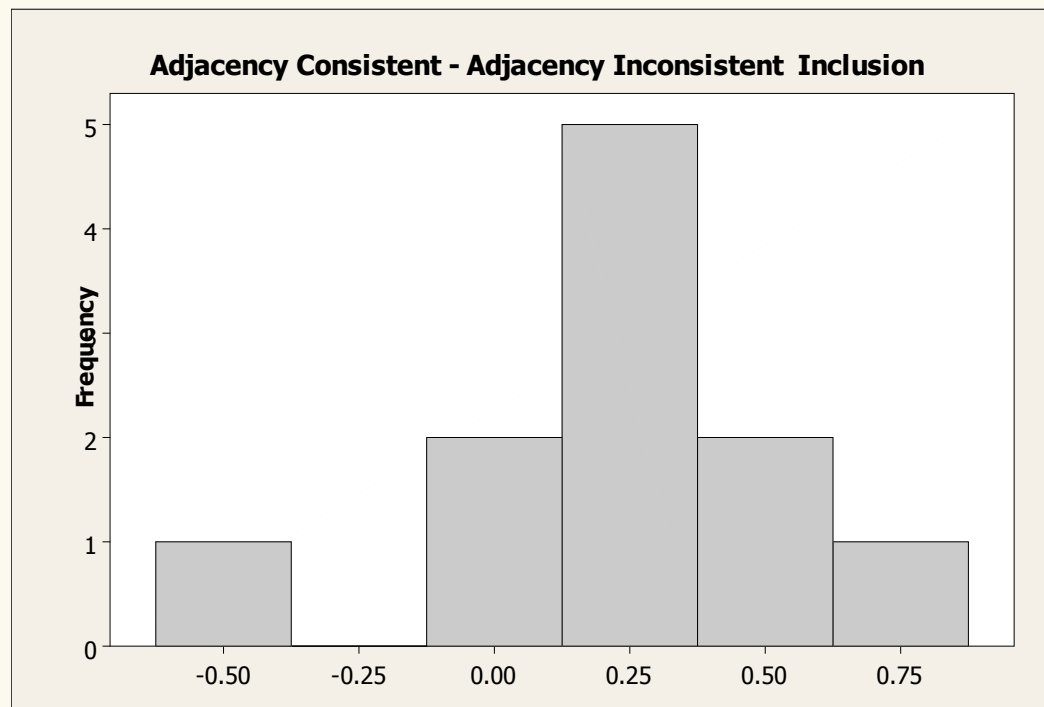
$$\text{Adjacency Consistent Error} = \frac{\text{\# of graphs committed that are adjacency consistent}}{\text{\# of graphs committed}}$$



Experiment 3: Adjacency Integrity

$$\text{Adjacency Consistent Inclusion} = \frac{\text{\# of committed graphs that are adjacency consistent}}{\text{\# of committable graphs that are adjacency consistent}}$$

$$\text{Adjacency Inconsistent Inclusion} = \frac{\text{\# of committed graphs that are adjacency inconsistent}}{\text{\# of committable graphs that are adjacency inconsistent}}$$



Tentative Hypotheses from Pilot Study

- Students can distinguish direct from indirect causation
- Students prefer passive observation
- Students often choose optimal experiments to orient a chain
- Students act as if they understand adjacency, but not orientation within an adjacency class reliably