# Babies and Bayes Nets II: Observations, Interventions and Prior Knowledge

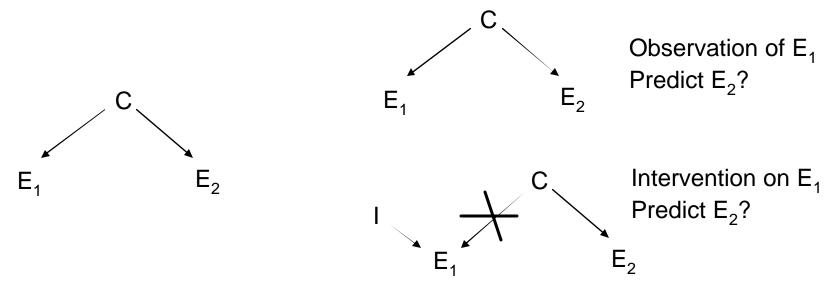
Tamar Kushnir Alison Gopnik

# **Outline**

- Part 1: Interventions vs Observations
  - Causal strength judgments
  - Causal structure learning
- Part 2: Intervention Source
  - Causal strength judgments
  - Causal structure learning
- Part 3: Prior Knowledge
  - Integrating prior knowledge
  - Overriding prior knowledge

# Part 1: Interventions vs. Observations

 Do we distinguish between observations and interventions in making inferences about causal strength? (Waldmann & Hagmayer, in press; Sloman & Lagnado, 2005)

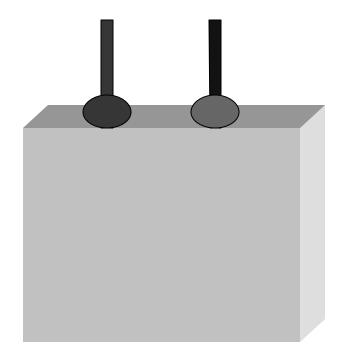


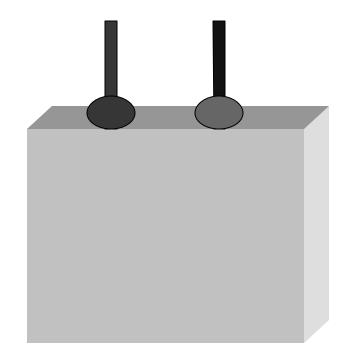
- Given structure
- Given observational data
- Adults make different inferences about hypothetical interventions than about hypothetical observations
  T. Kushnir, Few 2005

# Part 1: Interventions vs. Observations

- Do we distinguish between observations and interventions in causal structure learning?
- Stick ball machine: Gopnik et al 2004; Kushnir et al, 2003
  - No special instructions about interventions
  - Matched (between conditions) for pattern of movement.
  - Study 1 (Adults): No interventions
  - Study 2 (Children): No observations

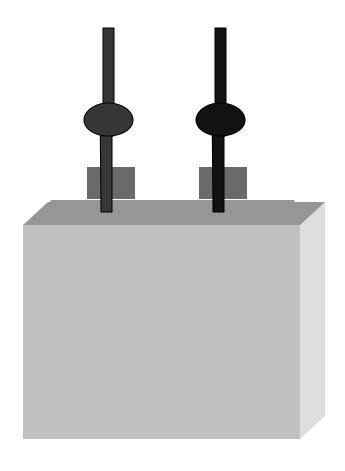
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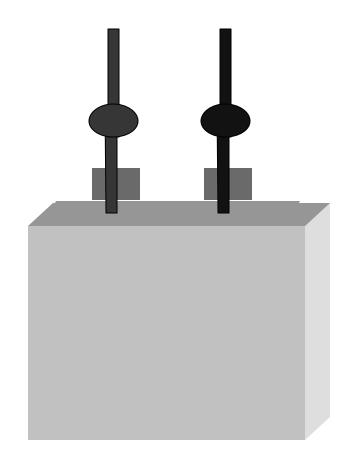




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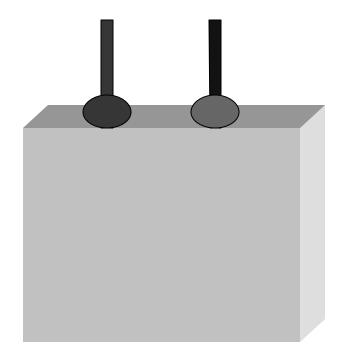
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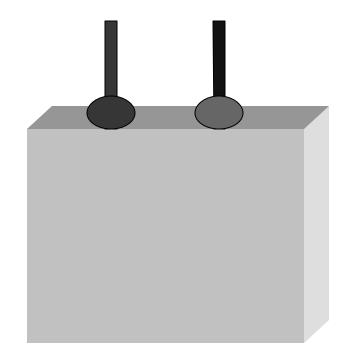




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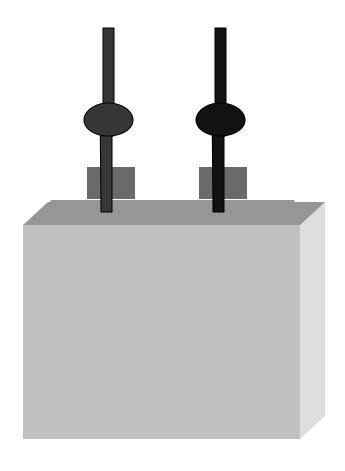
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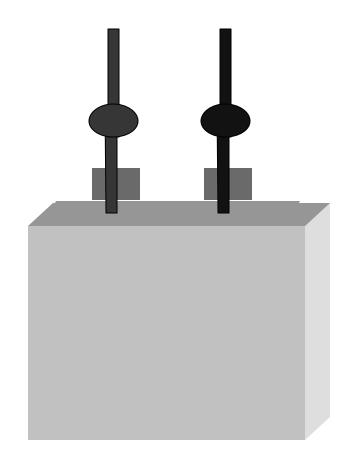




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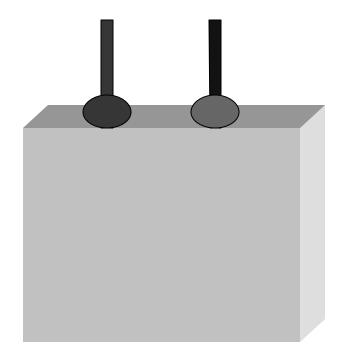
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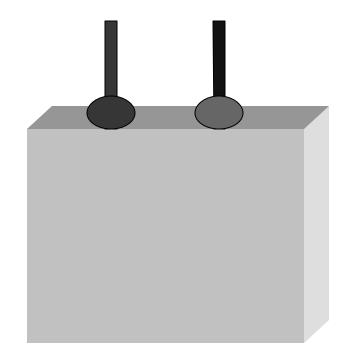




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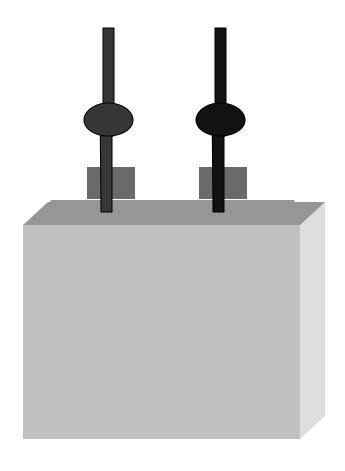
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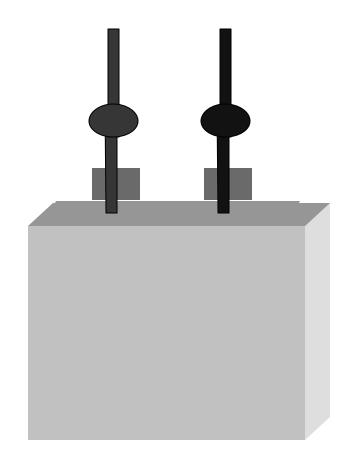




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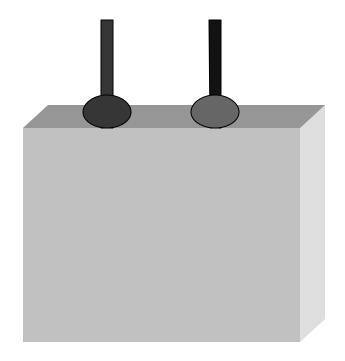
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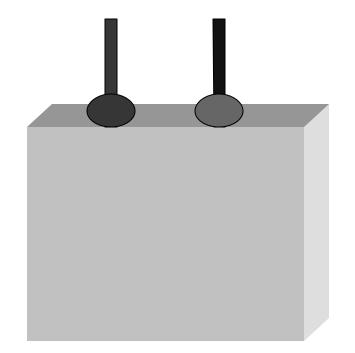




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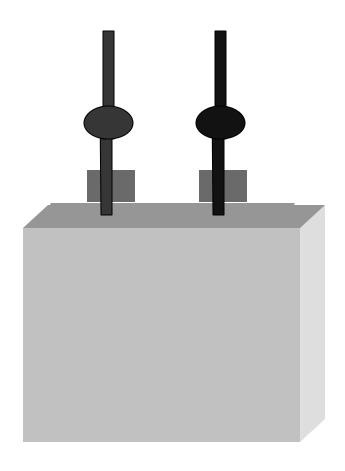
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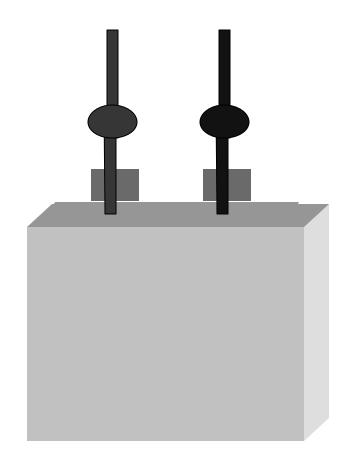




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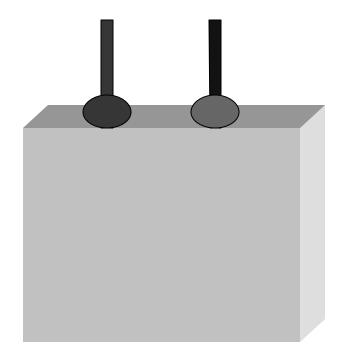
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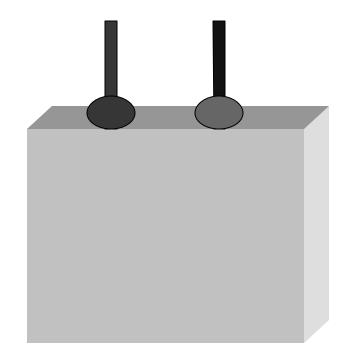




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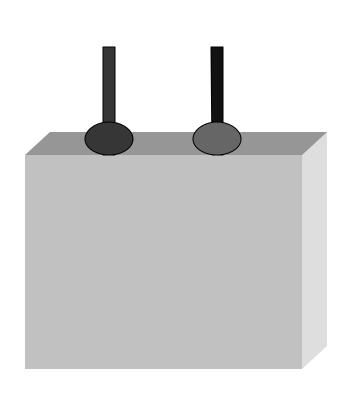
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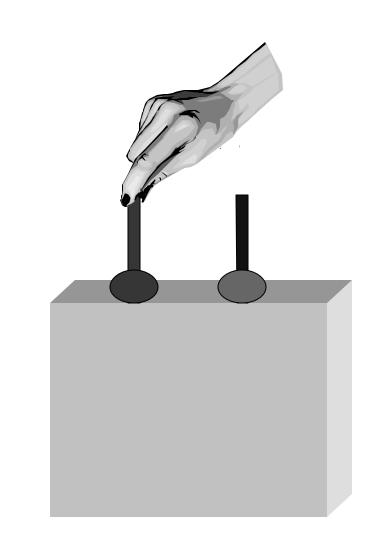




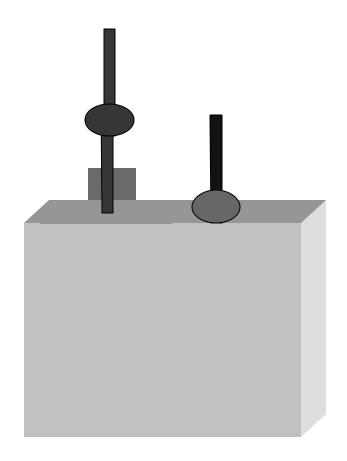
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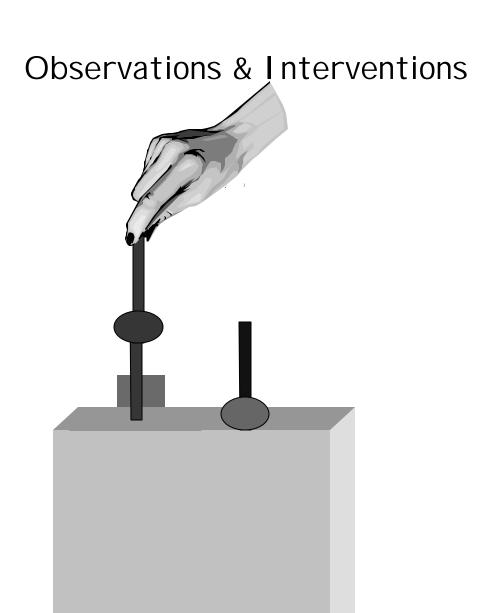
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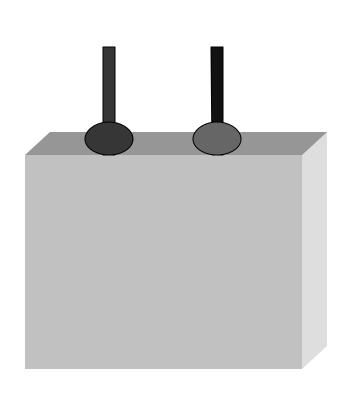
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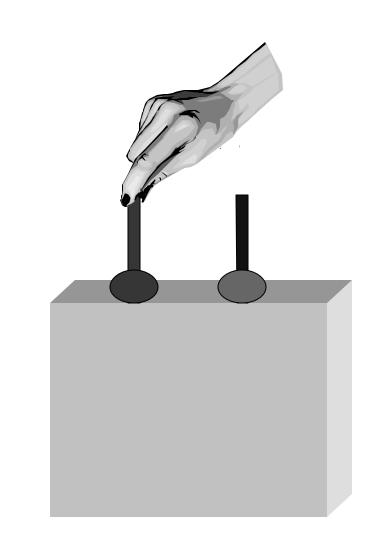




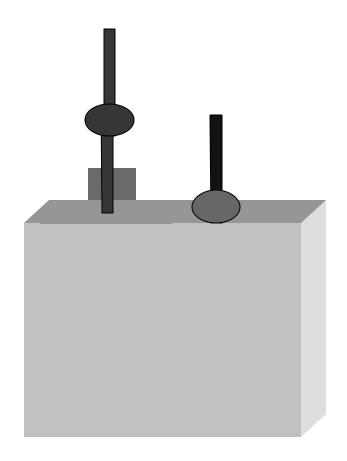
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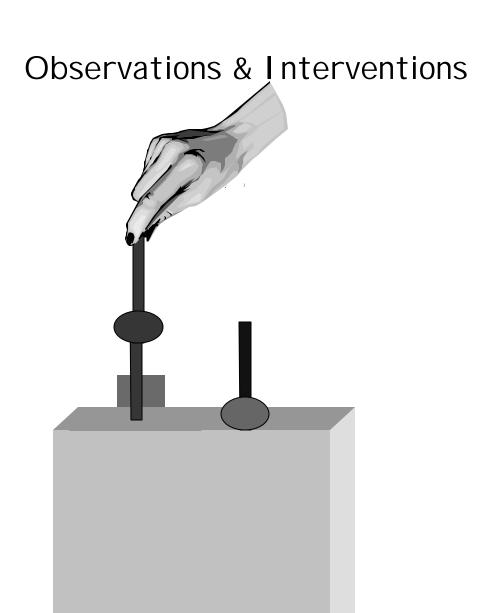
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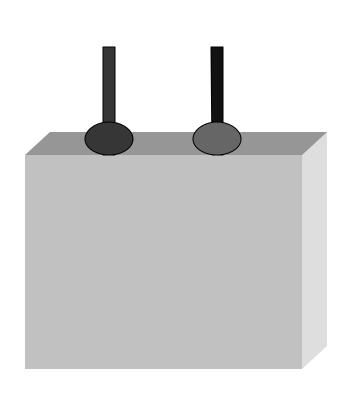
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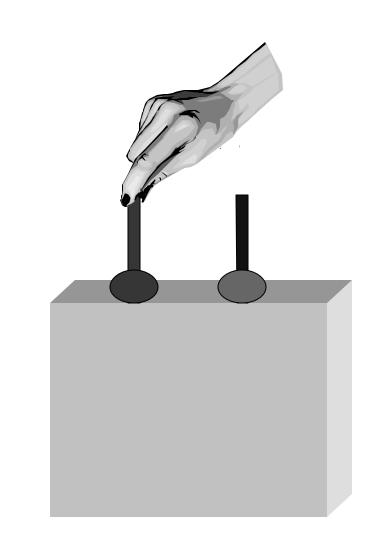




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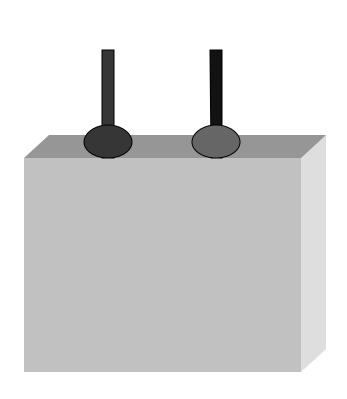
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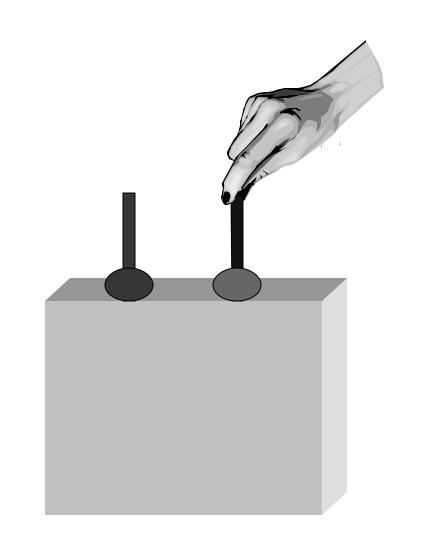




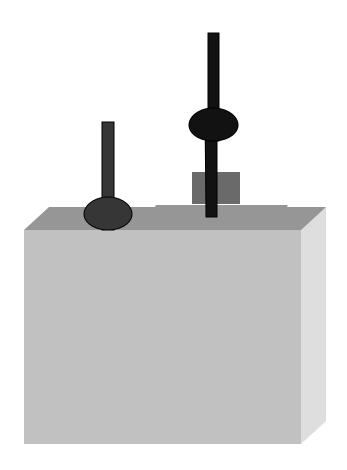
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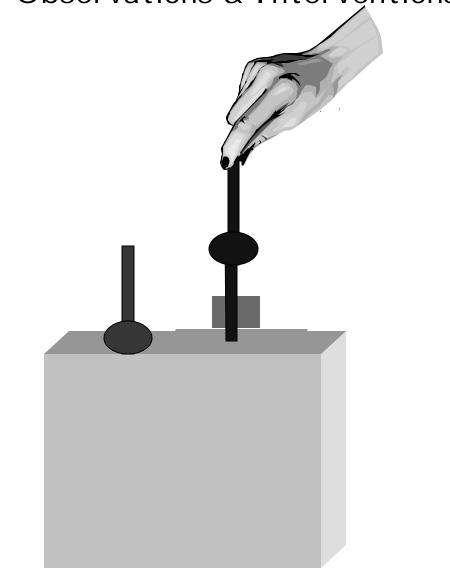




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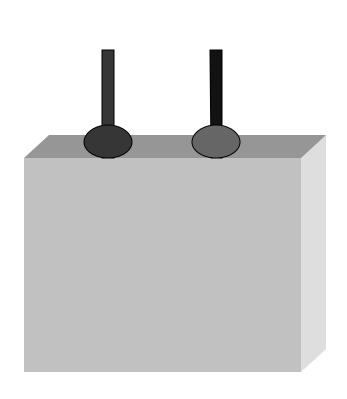


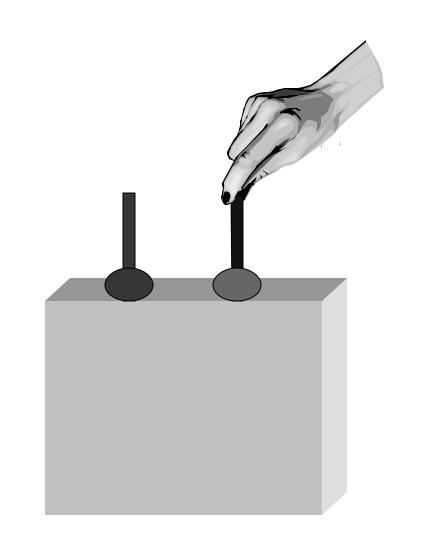




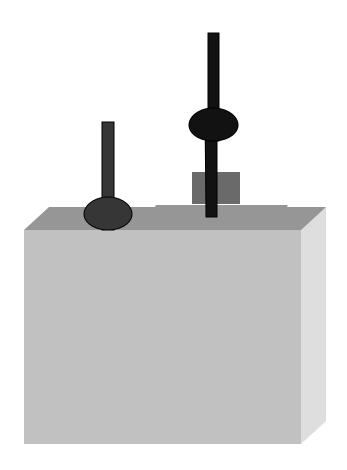
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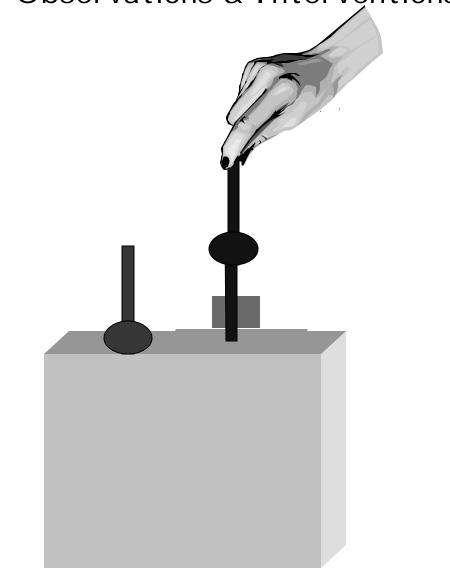




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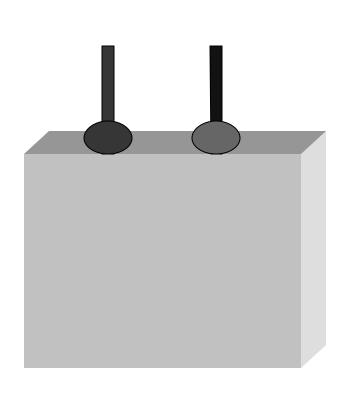


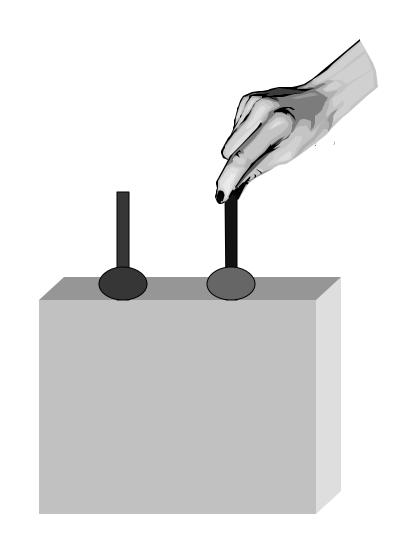




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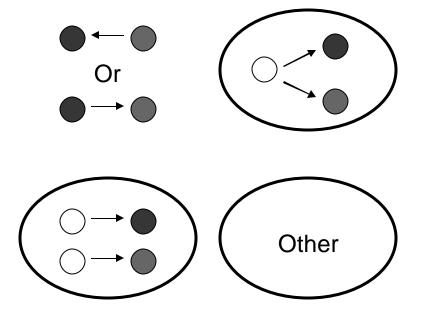


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A & B - 4 times

A alone – 2 times

B alone – 2 times

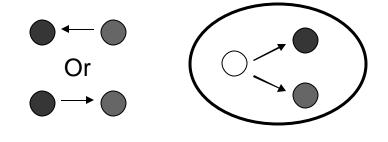


#### Observations & Interventions

A & B - 4 times

Intervene on A: A alone – 2 times

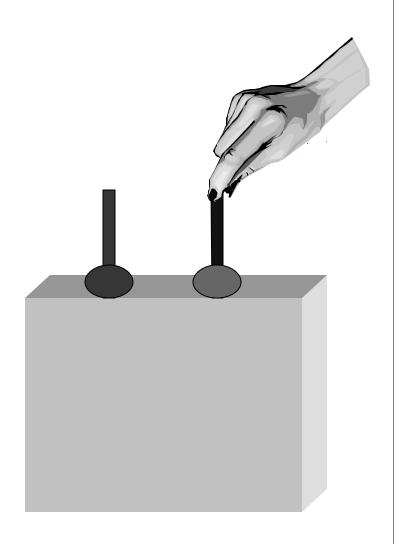
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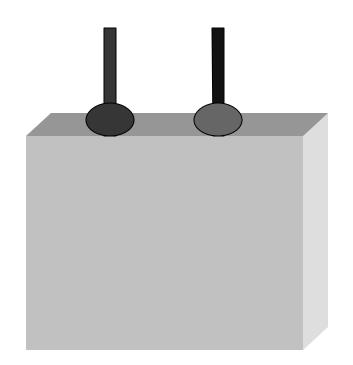




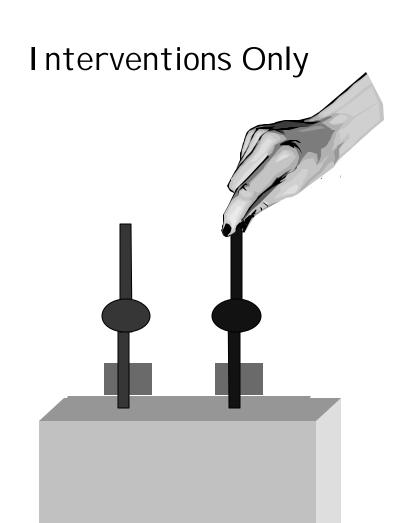
## Interventions Only

#### Observations & Interventions

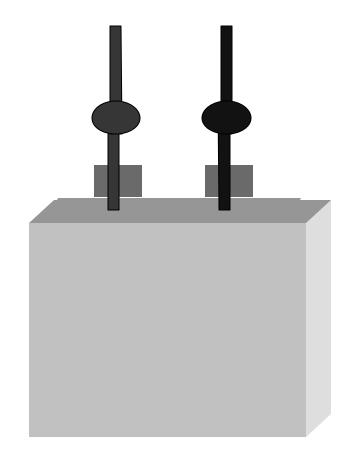




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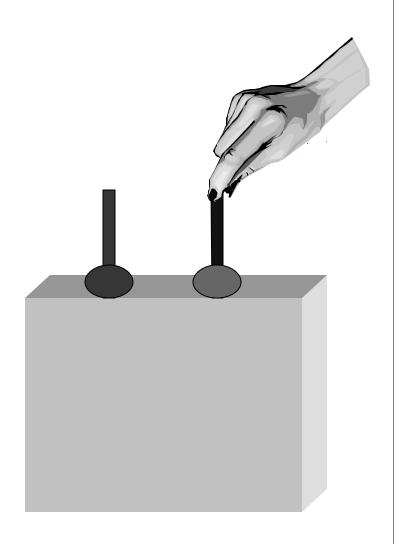
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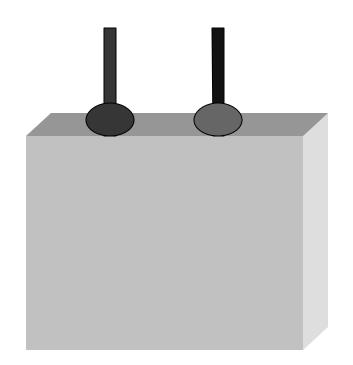


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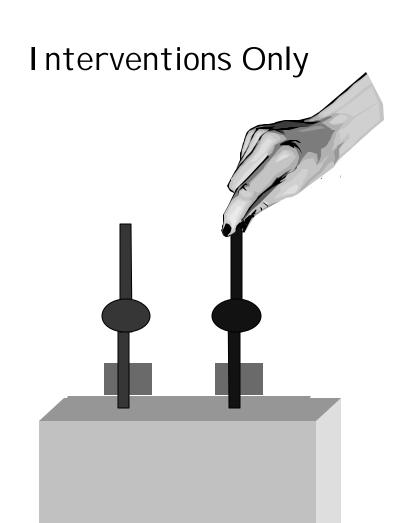
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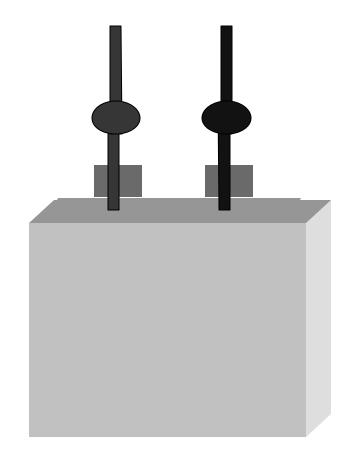




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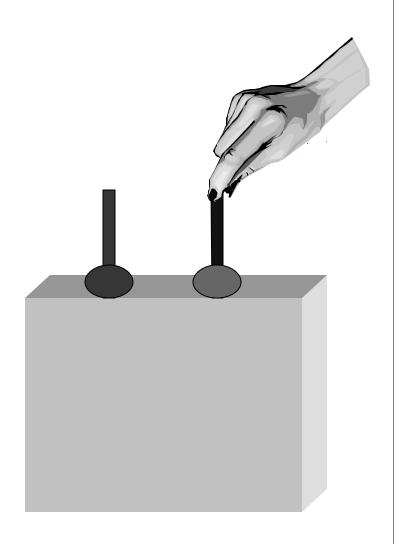
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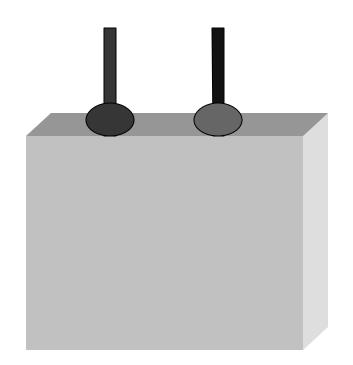


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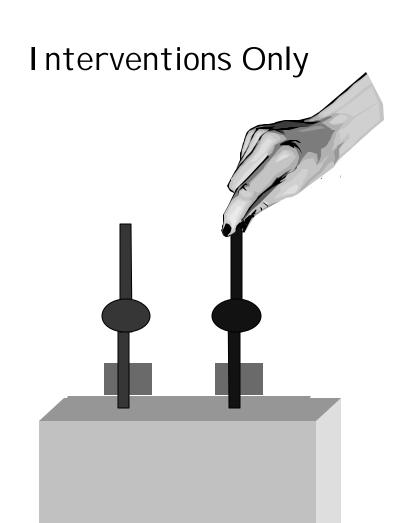
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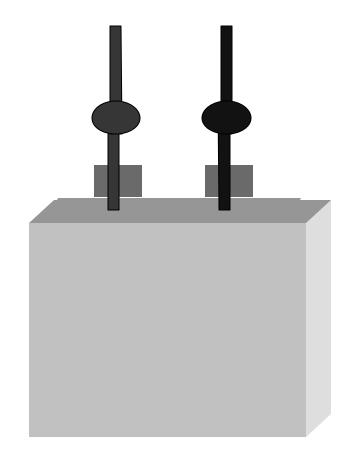




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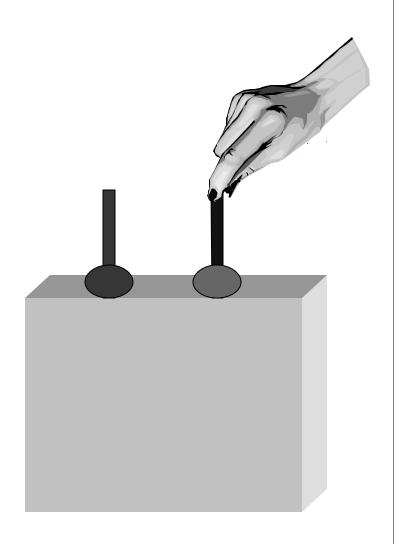
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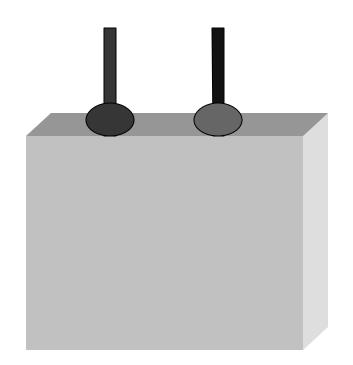


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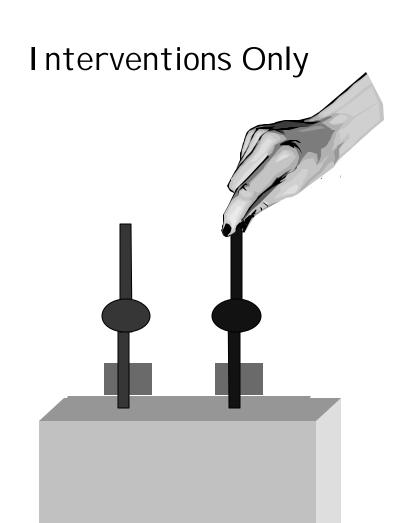
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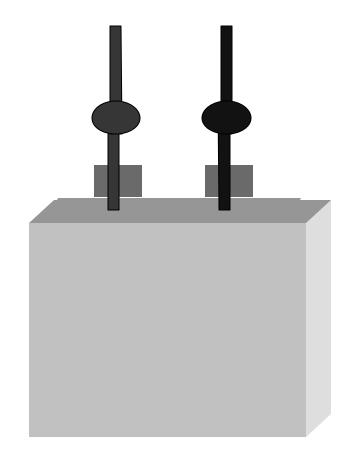




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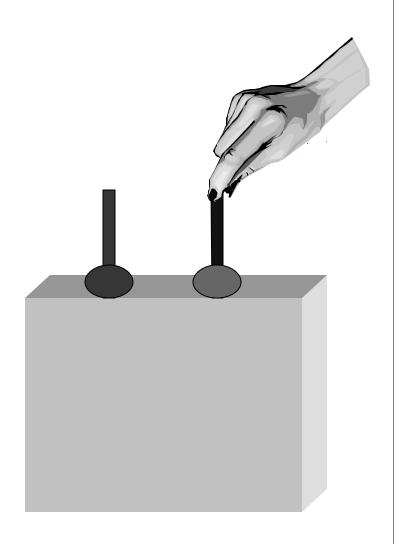
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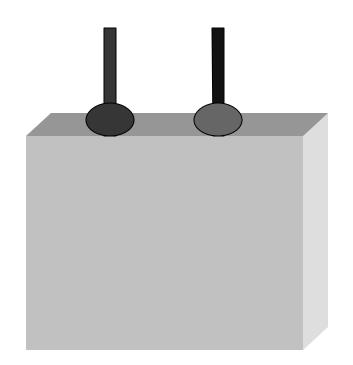


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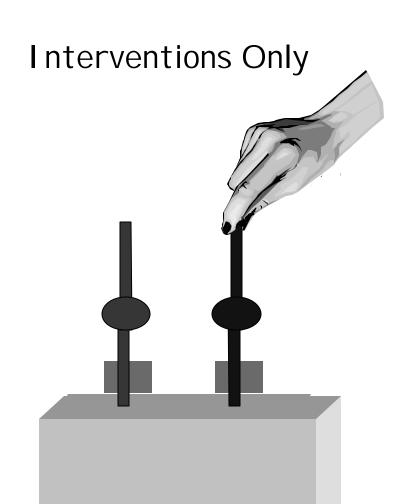
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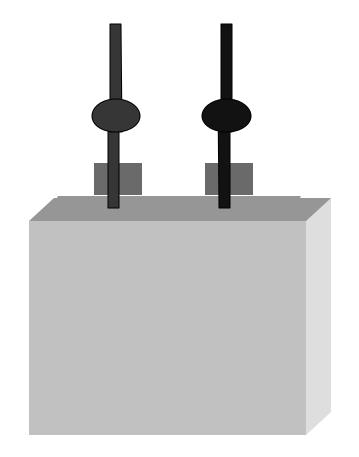




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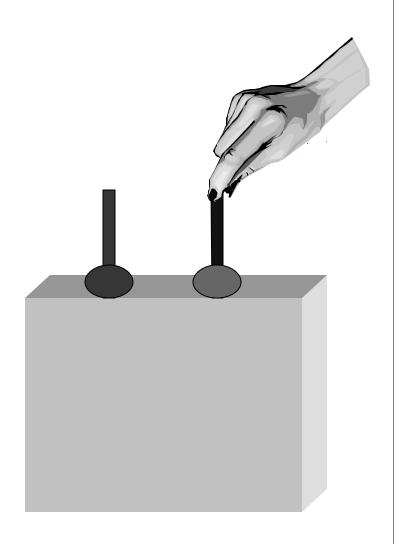
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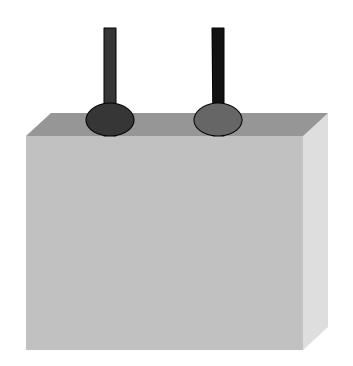


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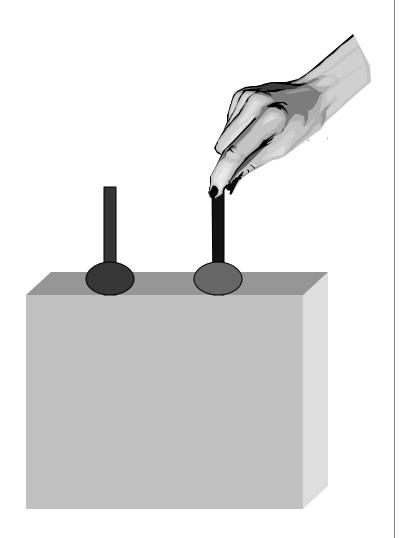


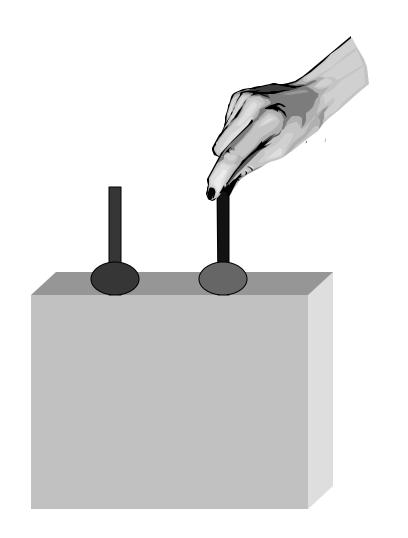


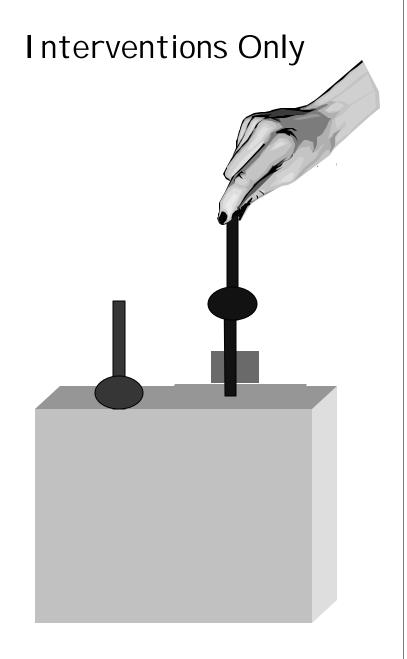
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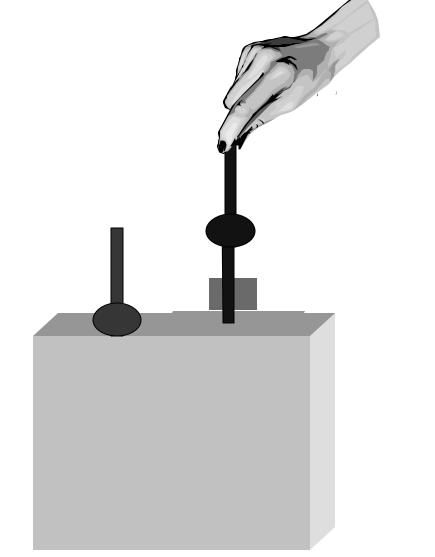
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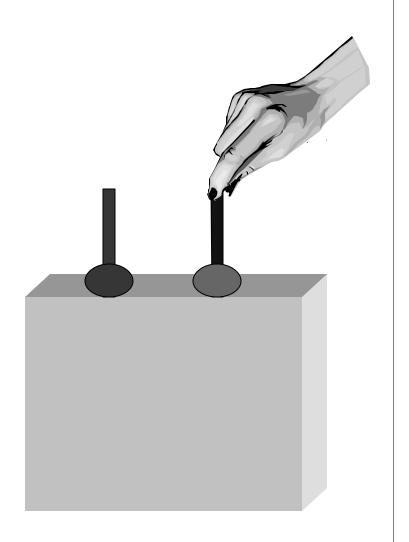


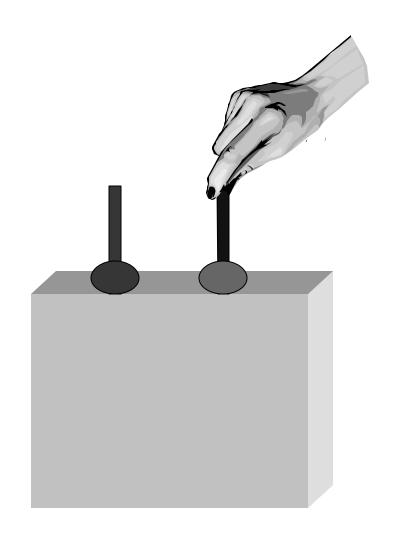




#### Interventions Only

#### Observations & Interventions



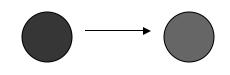


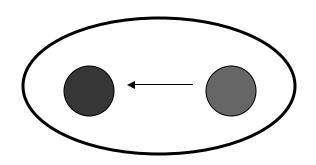
T. Kushnir, Few 2005

Gopnik et al, 2004

#### Interventions Only

Intervene on B: A & B – 5 times
Intervene on B: B alone – 1 time

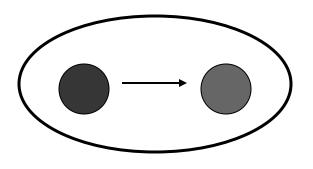


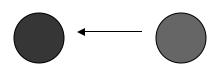


#### Observations & Interventions

A & B – 5 times

Intervene on B: B alone – 1 time





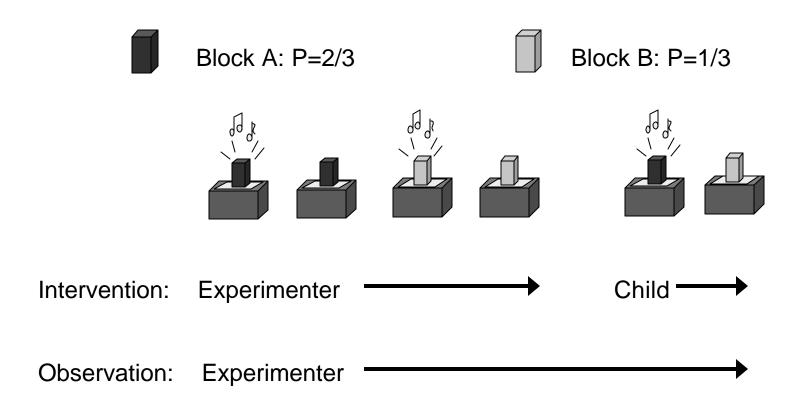
#### Part 1: Conclusion

- Both adults and children
  - Distinguish observations from interventions
  - Integrate evidence from both sources of data to learn causal structure.
  - Results (from yesterday) of complex causal structure learning in both children and adults are not possible without the ability to do both of the above.

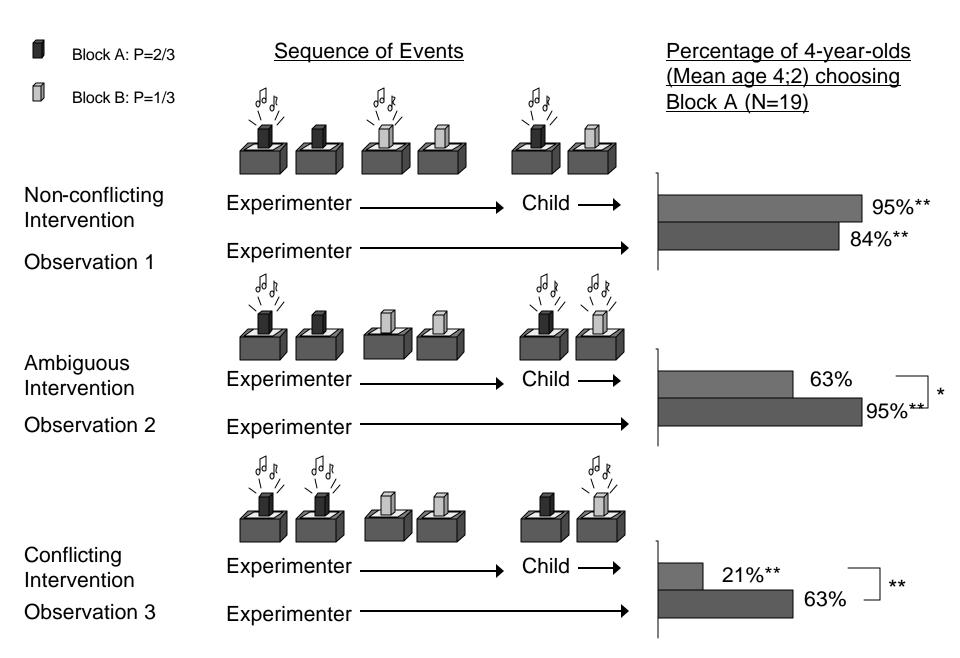
#### Part 2: Intervention Source

- Does the source of interventions one's own or someone else's - matter?
  - Study 1 (Children): How do causal strength judgments interact with intervention source?
  - Study 2 (Adults): How is structure learning affected by intervention source?
  - Study 3 (from yesterday): Children's own spontaneous interventions in play.

### Intervention Source & Probability Judgment



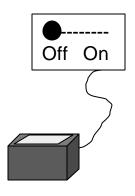
- 1. Intervention Question: "Pick the best one and make the toy go."
- 2. Hidden Cause Question: "Which one has more stuff inside?"

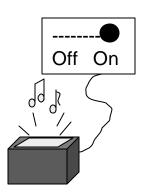


### Intervention Source & Probability Judgment

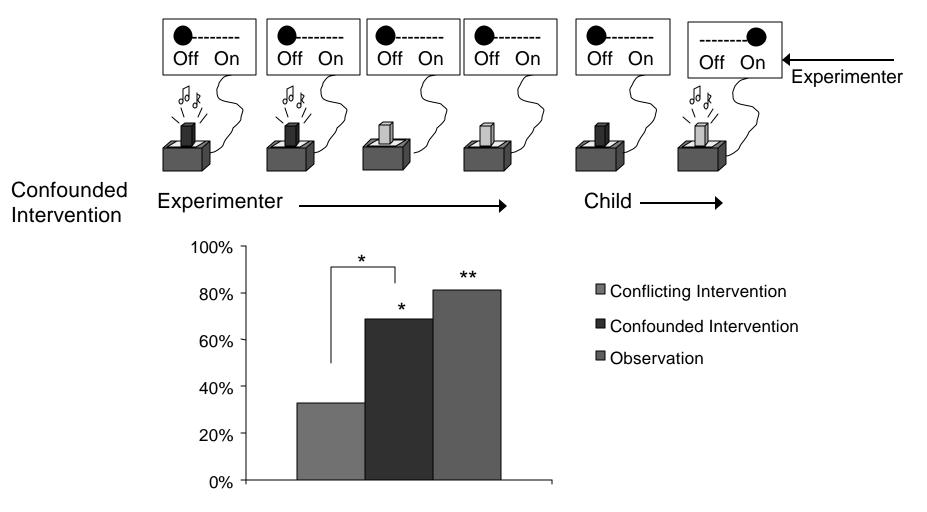
- When observing the experimenter intervene, probabilistic strength = more hidden stuff.
- When their own interventions conflicted with the overall probabilities, the object that worked for them = more hidden stuff.
- Is this because children believe their own interventions are more likely to be unconfounded?
- What if their own interventions are explicitly confounded?
   Will they still show a preference?

### Intervention Source and Confounding





### Intervention Source and Confounding



"Which one has more stuff inside?"

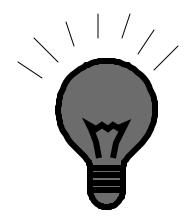
- The lab of "Dr. Science" (Sobel & Kushnir, in press).
  - The learning environment:
    - Colored lights and sensors
    - No spatio-temporal information
    - Generative & preventative interventions.
  - Three yoked conditions
    - Own Intervention
    - Observation of Intervention
    - Forced Intervention



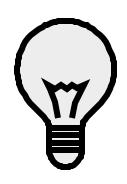




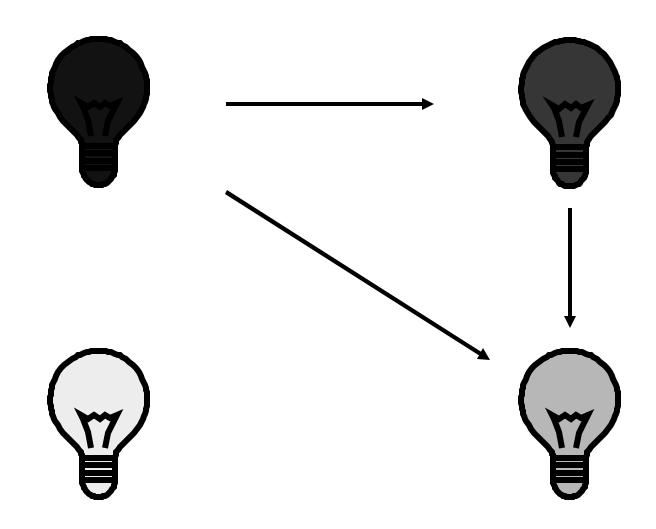


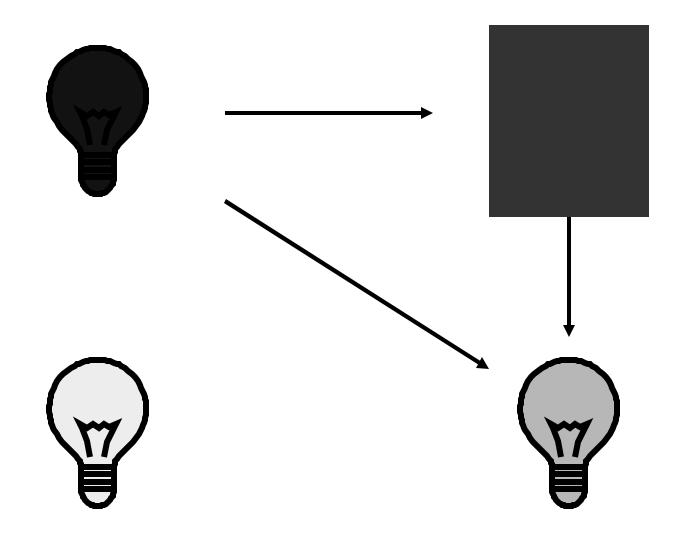


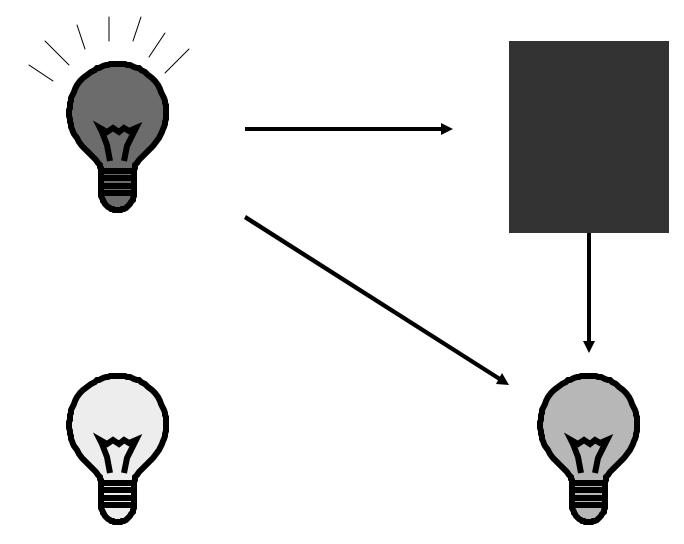


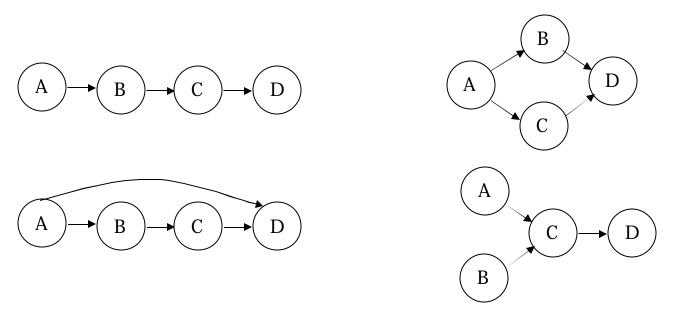












- Results: In the Own Intervention condition...
  - Accuracy significantly higher
  - Significantly more participants learned structure perfectly
  - Performance correlates with the number of critical interventions made.

#### Part 2: Conclusion

- There are several ways in which one's own interventions may be critical for causal learning.
  - Four-year-olds give preference to evidence from their own interventions over probabilities (because they believe those interventions are unconfounded).
  - Adults learn causal structure best when deciding how to intervene (because it enables them to pay attention to critical interventions).
  - Adults (from Dr. Science) and children (from yesterday) are capable of spontaneously generating appropriate interventions for structure learning through play.

### Part 3: Prior Knowledge

- By preschool, children have a substantial amount of knowledge about the world.
- Knowledge is organized into domain-specific theories.
  - Folk Physics (Spelke et al, 1992)
  - Folk Biology (Gelman & Wellman, 1991)
  - Folk Psychology (Wellman, 1990)
- What role does existing knowledge play in learning new causal relations?
  - Integrating prior knowledge: Base rates
  - Overriding prior knowledge: Domain specific assumptions

## Integrating Prior Knowledge

- Do preschool children pay attention to base rates?
- Backwards Blocking task
  - Objects that make the toy go are "Blickets"
  - Blickets are either rare (1/10) or common (9/10)

				Results		
Jo k	10 8	Jo k	Is a Blicket?	<u>Rare</u> 100%	Common 100%	
			ls a Blicket?	25%	81%	

Prior knowledge about base rates affects children's inferences about ambiguous causes

### Overriding Prior Knowledge

- Children's ability to make inferences about patterns of dependence and independence (i.e. "screening off") is domain independent. (Schulz & Gopnik, 2004)
- Are children's causal inferences restricted by domain-specific knowledge? Can they learn a new causal relation that overrides existing domain knowledge?

### Overriding Prior Knowledge

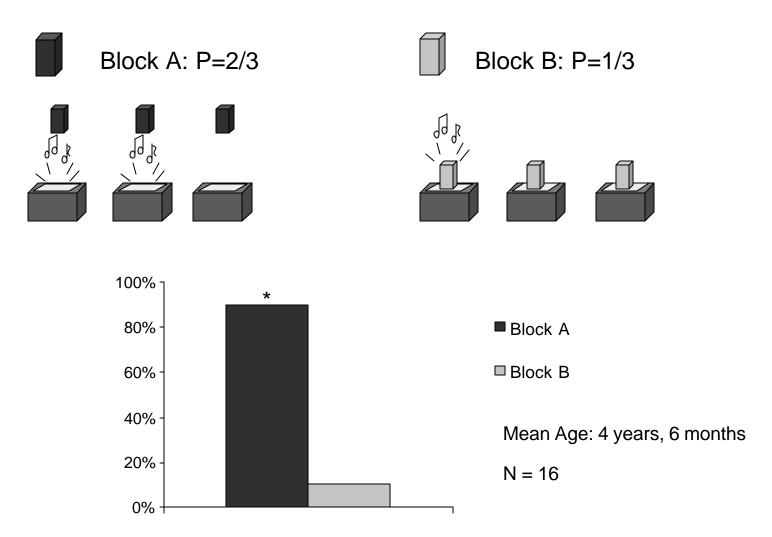
 Children's physical knowledge: causal relations constrained by time and space (billiard balls).

 Giving children mechanism information allows them to learn causal relations in the absence of spatial cues. (Bullock et al, Schultz)

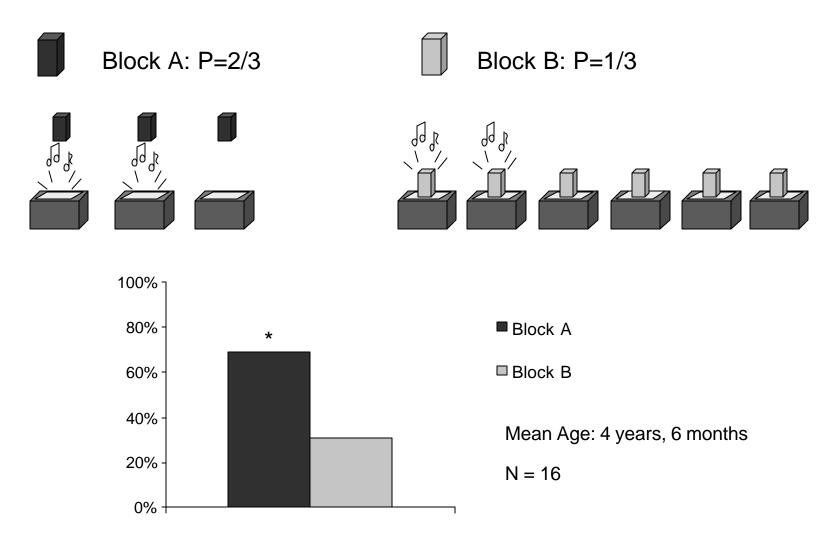
 Can they use patterns of probability, without mechanisms, to override spatial cues?



 Initial intervention (no training): ALL children make contact between block and toy – strong preference for spatial contiguity.



Causal Strength Question: "Which one works best?"

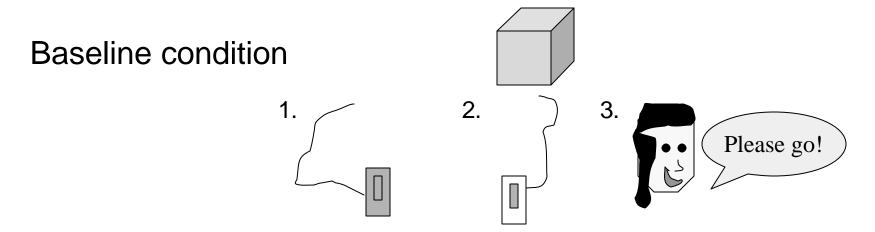


Causal Strength Question: "Which one works best?"



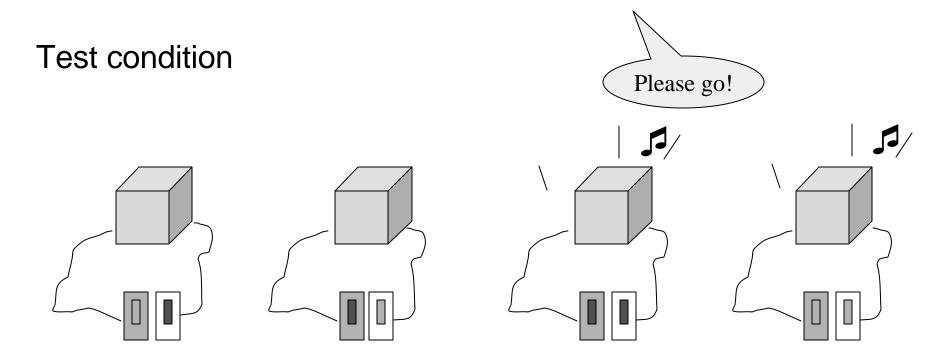
- Initial intervention (no training): ALL children make contact between block and toy – strong preference for spatial contiguity.
- Final intervention: A significant number of children revised their initial intervention and held the new object over the toy.

#### Overriding Prior Knowledge Across Domains



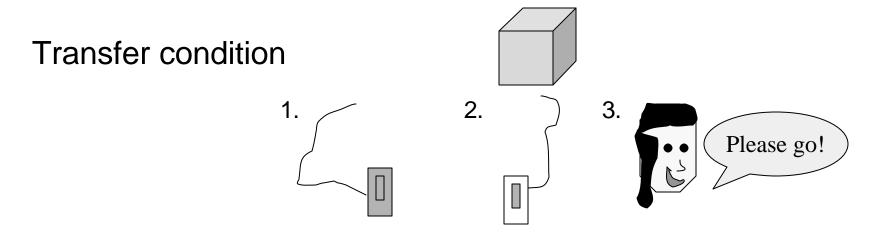
 Baseline intervention: ALL children choose the switches – strong preference for domainappropriate causes.

#### Overriding Prior Knowledge Across Domains



- Can you make the machine turn off?
- 75% said "please stop."
- Results identical to within-domain screening off.

### Overriding Prior Knowledge Across Domains



- Baseline intervention: ALL children choose the switches – strong preference for domainappropriate causes.
- Transfer intervention: Most chose switches, but significantly more chose to say "please go" than at baseline.

#### Part 3: Conclusion

- Prior knowledge plays a significant role in learning new causal relations.
  - Children pay attention to base rates when making inferences about ambiguous causes.
  - Children also have strong preferences for domainappropriate causal relations.
  - Children can override prior knowledge when faced with new evidence in the form of patterns of probabilities.

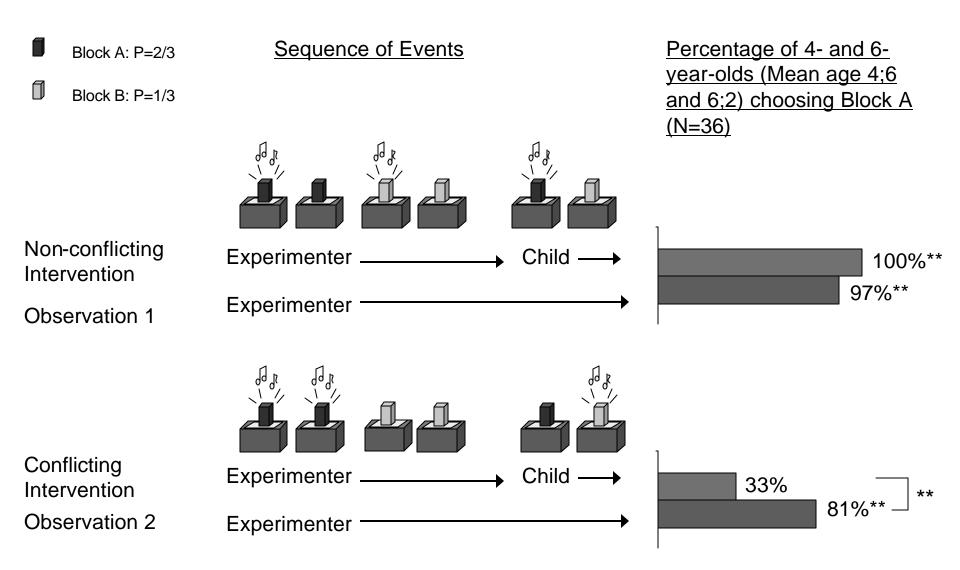
### Summary

- Part 1: Both adults and children distinguish between interventions and observations when learning causal structure.
- Part 2: The source of intervention information influences causal learning.
- Part 3: Causal learning involves both integrating and overriding prior knowledge.

#### **Future Directions**

- Probabilities
- Determinism
- Active Learning
- Experimental design and intervention
- Integration with spatio/temporal information
- Explanation
- Bootstrapping and revision

# Thank you



Hidden Cause Question: "Which one has more stuff inside?"

# Accuracy on the Causal Structure Questions and Percentage of Learners who Responded Correctly on All Twelve questions in Experiment 1

	<u>Condition</u>				
	Self-Int	ervention	Obs. of Intervention		
	Accuracy	<u>Percentage</u>	<u>Accuracy</u>	<u>Percentage</u>	
Chain Model	11.18	59	9.88	35	
Diamond Model	11.06	24	9.47	24	
Chain Model w/A→D	11.06	53	9.35	24	
Common Effect & Chain	11.65	82	9.18	35	

# Accuracy on the Causal Structure Questions and Percentage of Learners who Responded Correctly on All Twelve questions in Experiment 2

	<u>Condition</u>					
	Self-Int	ervention	Forced Intervention			
	Accuracy	<u>Percentage</u>	<u>Accuracy</u>	<u>Percentage</u>		
Chain Model	10.88	46	9.88	21		
Diamond Model	10.96	25	9.88	17		
Chain Model w/A→D	11.08	50	9.96	29		
Common Effect & Chain	11.33	75	9.79	13		

#### Mean Number of Interventions and Percentage of Critical Interventions Generated or Observed by Learners in Experiment 1

<u>Model</u>	Mean Ir	nterventic	ns SD	Min	Max	%Critical
Chain		44.29	16.33	25	79	15.38
Common Effect &	& Chain	40.53	18.09	25	72	21.27
Chain + AD		42.82	22.28	25	101	22.27
Diamond		45.53	18.68	25	96	25.39

#### Mean Number of Interventions and Percentage of Critical Interventions Generated or Observed by Learners in Experiment 2

<u>Model</u>	Mean Ir	nterventic	ns SD	Min	Max	%Critical
Chain		49.04	19.40	25	89	15.92
Common Effect 8	k Chain	44.79	30.51	25	166	25.85
Chain + AD		48.71	25.96	25	120	23.94
Diamond		52.08	42.90	25	234	25.15