

# CS 486/686: Introduction to Artificial Intelligence

Causality

# Plan for Today

- Example using HMMs
- Introduction to Causality

# HMM Example

- Imagine we are trying to figure out what the weather was sometime in the past. However, we have lost the temperature data. But we do have a copy of Kate's diary where she recorded how many ice cream cones she ate each day! What can we infer about the weather?



# HMM Example

- States = {H, C}
- Observations {1 ice cream, 2 ice creams, 3 ice creams}
- Prior:  $P(H)=0.8$ ,  $P(C)=0.2$
- Dynamics:  $P(H|H)=0.6$ ,  $P(H|C)=0.5$
- Observation Model:

	H	C
1 Ice Cream	0.2	0.5
2 Ice Creams	0.4	0.4
3 Ice Creams	0.4	0.1

# Introduction to Causality

- Causality is the study of how things influence each other (causes lead to effects)
- Causal dependence:  $X$  causes  $Y$  if and only if changes to  $X$  lead to changes in  $Y$ 
  - Example: Diseases cause symptoms, but symptoms do not cause diseases

# Causality and Correlation (not the same thing!)

- A joint distribution  $P(X,Y)$  captures correlations between  $X$  and  $Y$  but does not capture whether a causal relation exists between  $X$  and  $Y$ , nor the direction of the causal relation if one does exist
- A conditional distribution  $P(X|Y)$  does not necessarily indicate  $X$  causes  $Y$

$$P(X|Y) = \frac{P(Y|X)P(X)}{P(Y)}$$

# Spurious Correlations

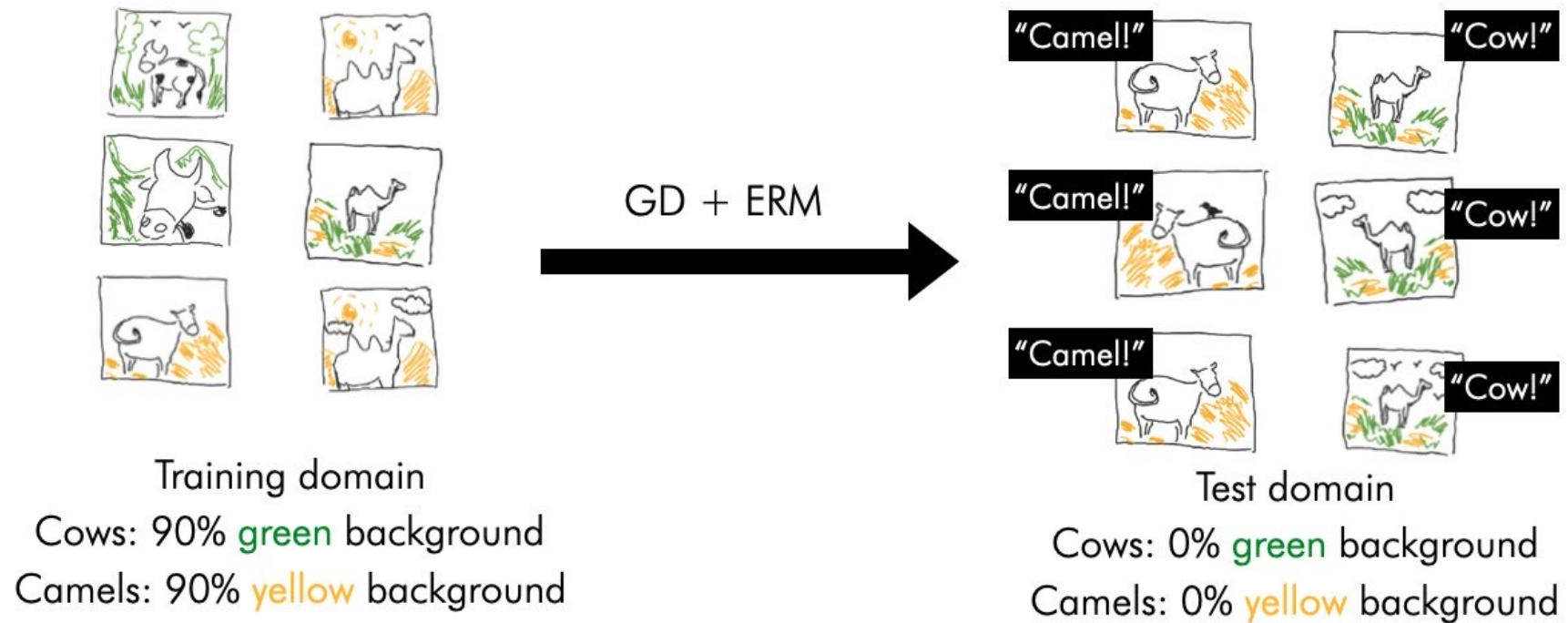
**Number of people who drowned by falling into a pool**  
correlates with  
**Films Nicolas Cage appeared in**



tylervigen.com

<https://www.tylervigen.com/spurious-correlations>

# Spurious Correlations

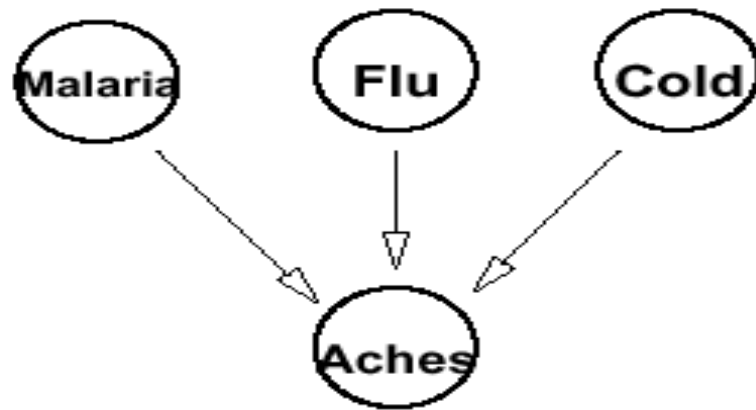


Standard example (Beery et al., '18 Arjovsky et al., '19)

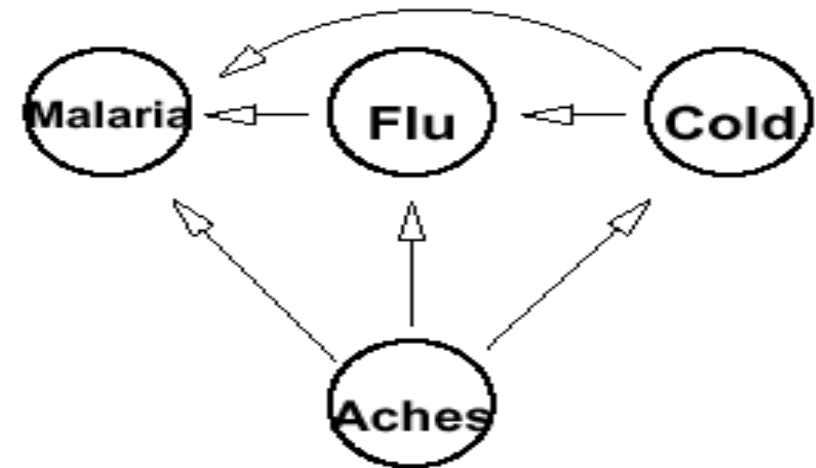


# Causal Bayes Net

- Bayes Net where all edges indicate direct causal effects.



Probabilistic Inference  
**Causal Inference**



Probabilistic Inference

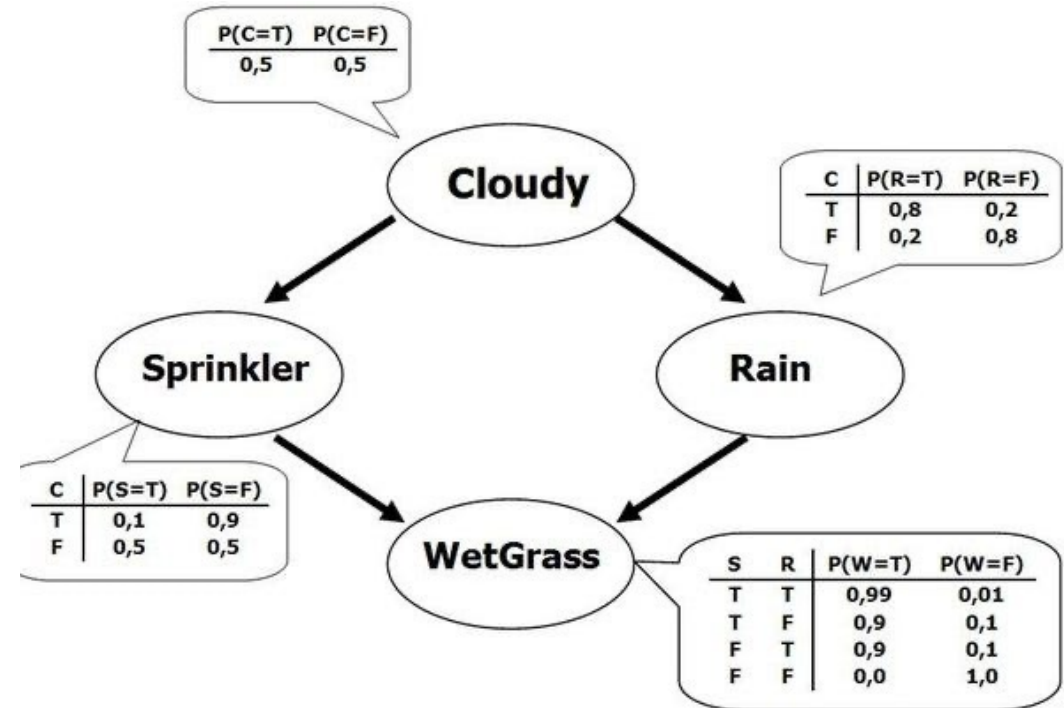
# Causal Inference

- Intervention: What is an effect of an action?
  - E.g. what is the effect of a treatment?

Causal networks support intervention queries but non-causal networks do not.

# Classic Causal Example

- Observation: What is the likelihood that the grass is wet when the sprinkler is observed to be on?
  - $P(WG|S=true)$
- Intervention: How does turning on the sprinkler affect the grass?
  - $P(WG|do(S=true))$



# Inference with Do Operator

Given a causal graph and query  $P(X | \text{do}(Y=y), Z)$

- Remove edges pointing to  $Y$  and  $\text{Parents}(Y)$
- Perform Variable Elimination on remaining graph
  - Restrict factors to evident  $Y=y, Z=z$
  - Eliminate variables
  - Multiply remaining factors and normalize

# What you should know

- Correlation does not imply causation!
- Causal Bayes Nets
  - Probabilistic inference vs causal inference and the do operator