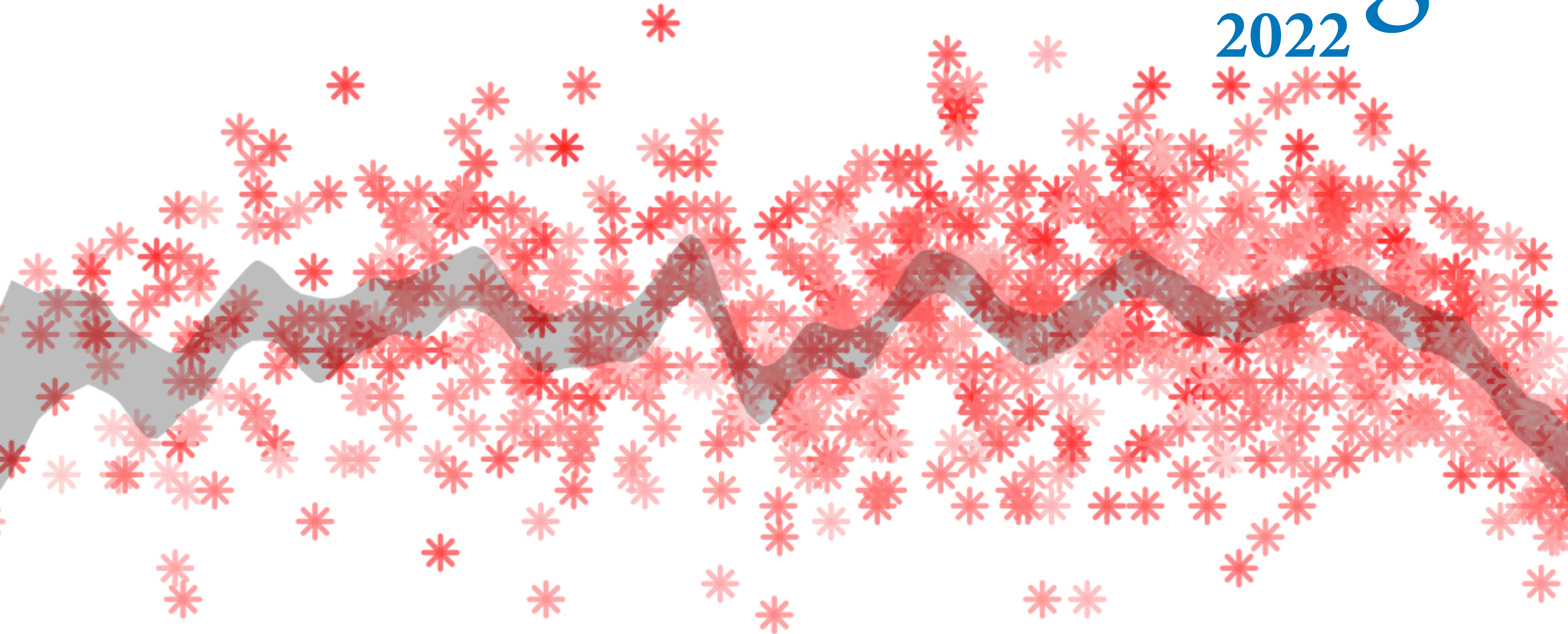


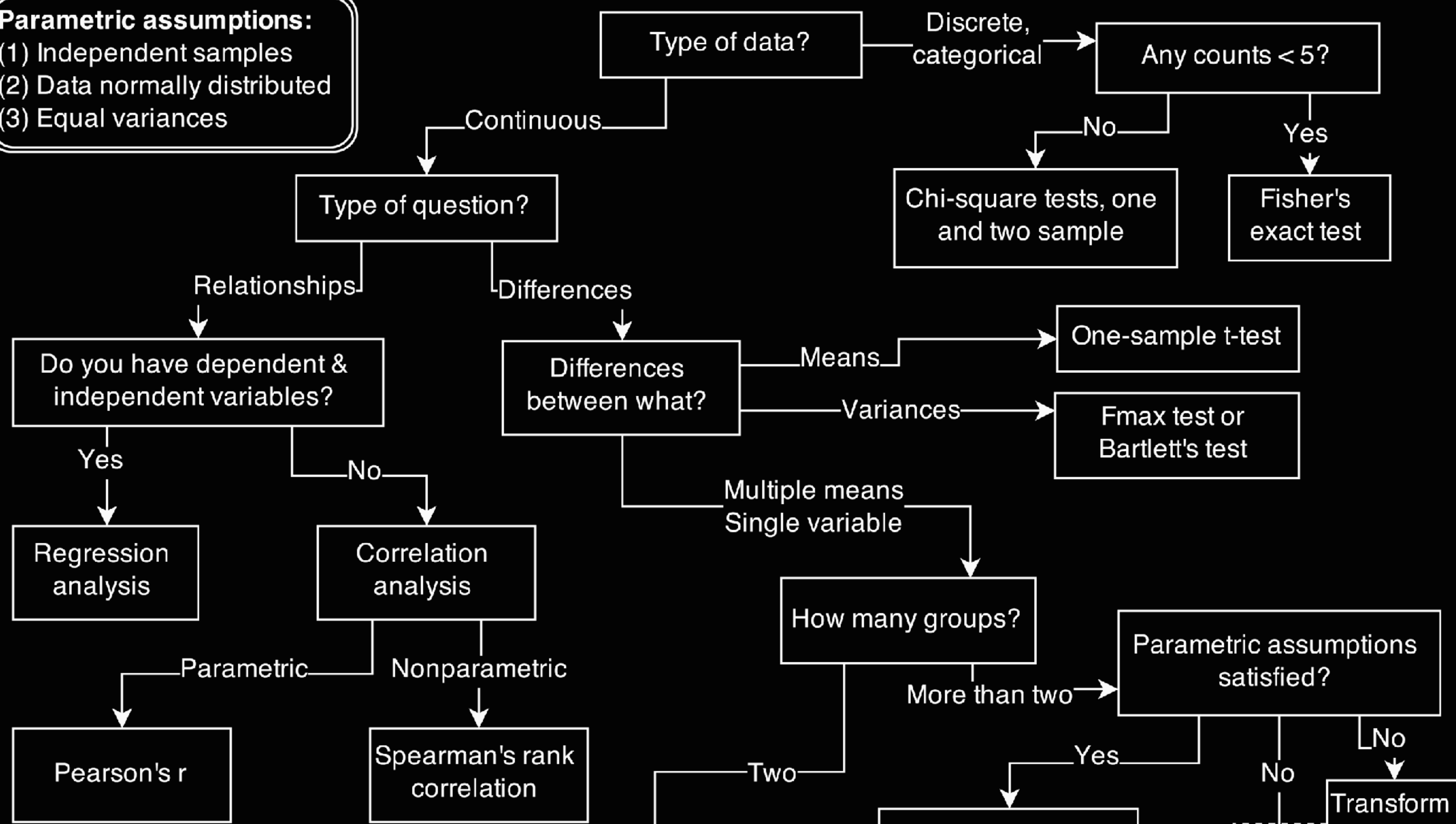
Statistical Rethinking

2022



(1) The Golem of Prague

Parametric assumptions:
 (1) Independent samples
 (2) Data normally distributed
 (3) Equal variances

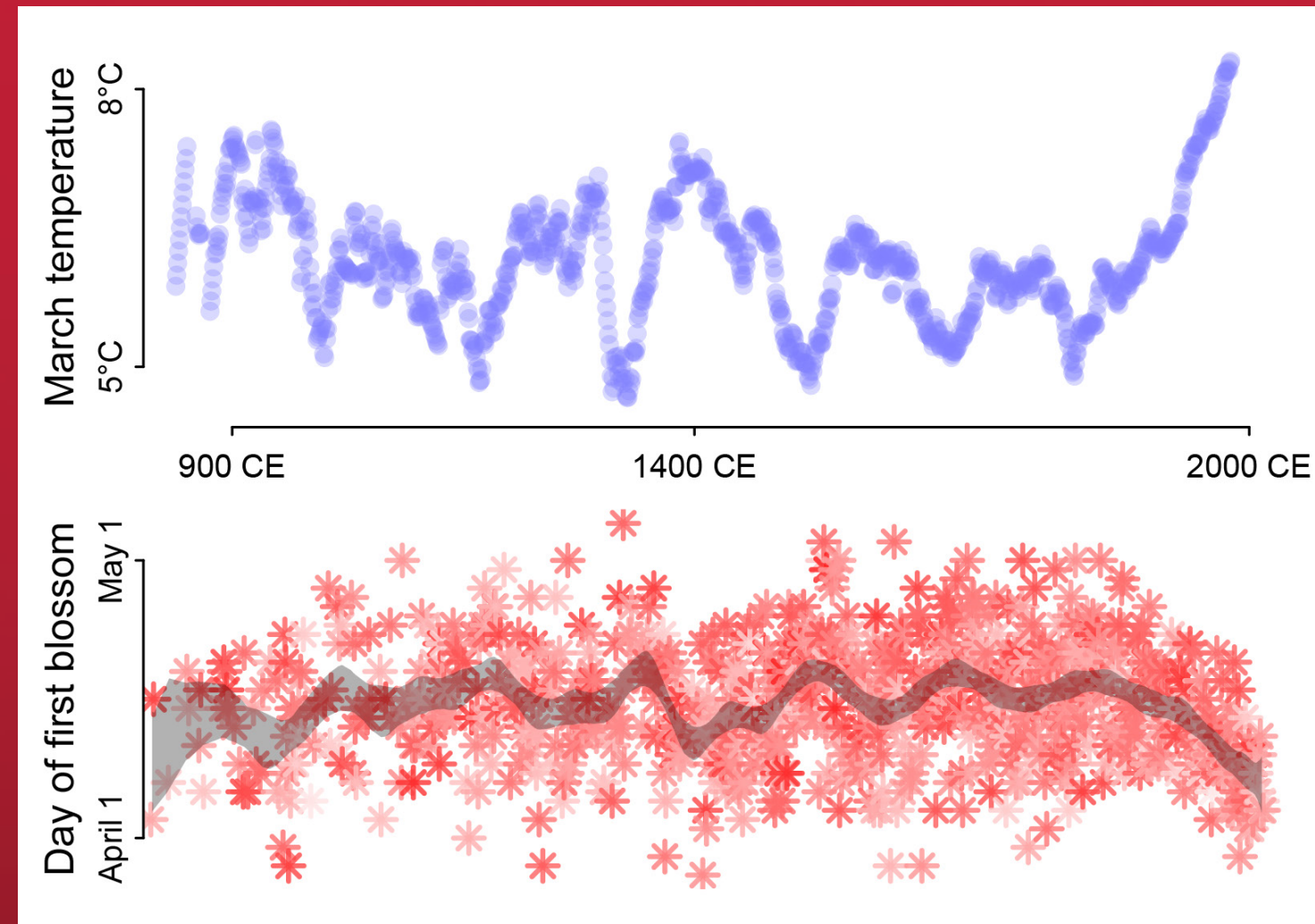


Second
Edition

Texts in Statistical Science

Statistical Rethinking

A Bayesian Course
with Examples in R and Stan
SECOND EDITION



Richard McElreath



 CRC Press
Taylor & Francis Group
A CHAPMAN & HALL BOOK

Rethinking
the role of
statistical
analysis in
research
20 lectures

Statistical Rethinking

McElreath

GOLEMS

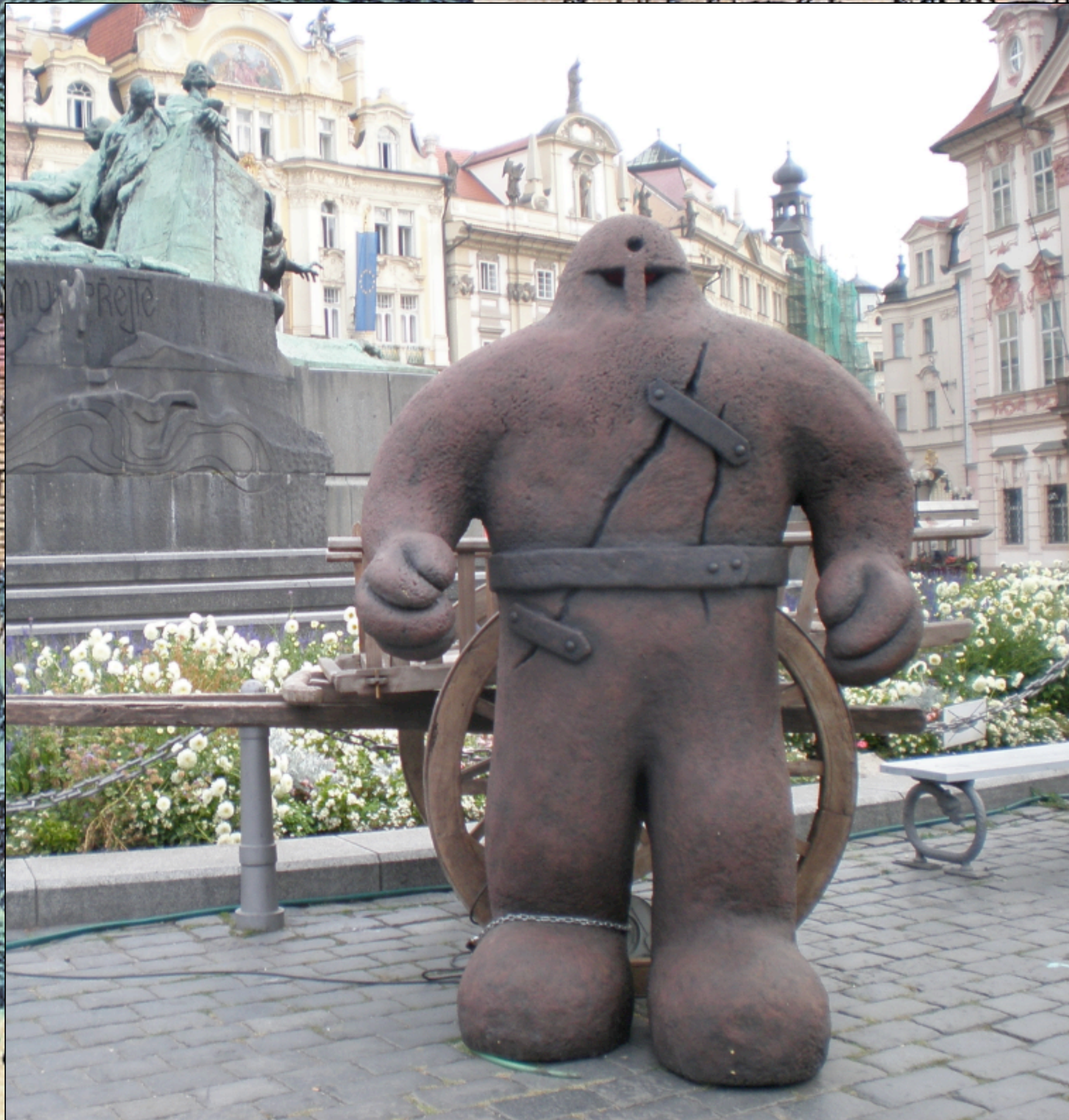
OWLS

DAGS

PRAGA

Prague 16th century







Art from: "Breath of Bones: A Tale of the Golem" (2014)

Golems

Clay robots

Powerful

No wisdom or foresight

Dangerous



“Breath of Bones: A Tale of the Golem” (2014)

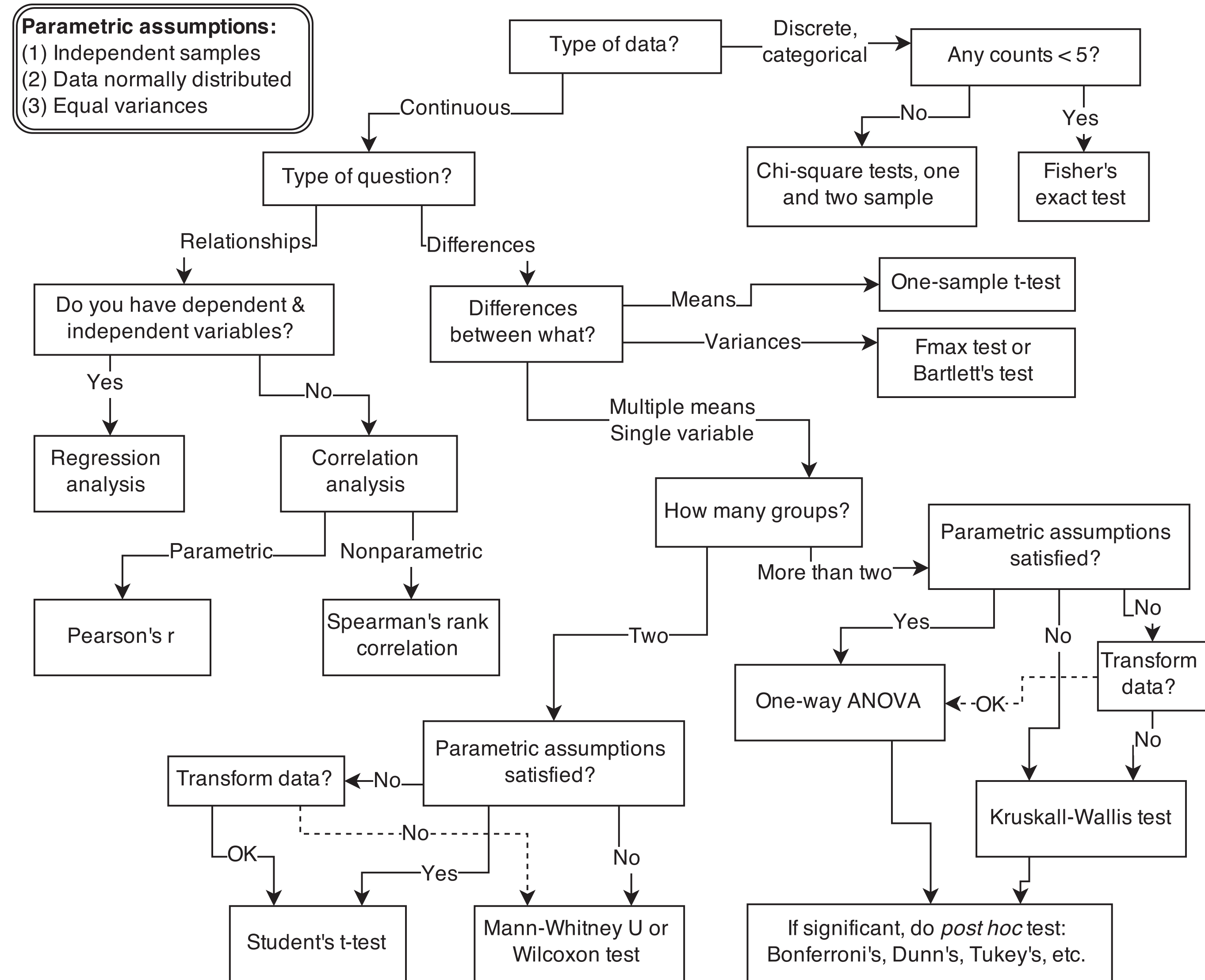
Statistical Models

Clay robots

Powerful

No wisdom or foresight

Dangerous



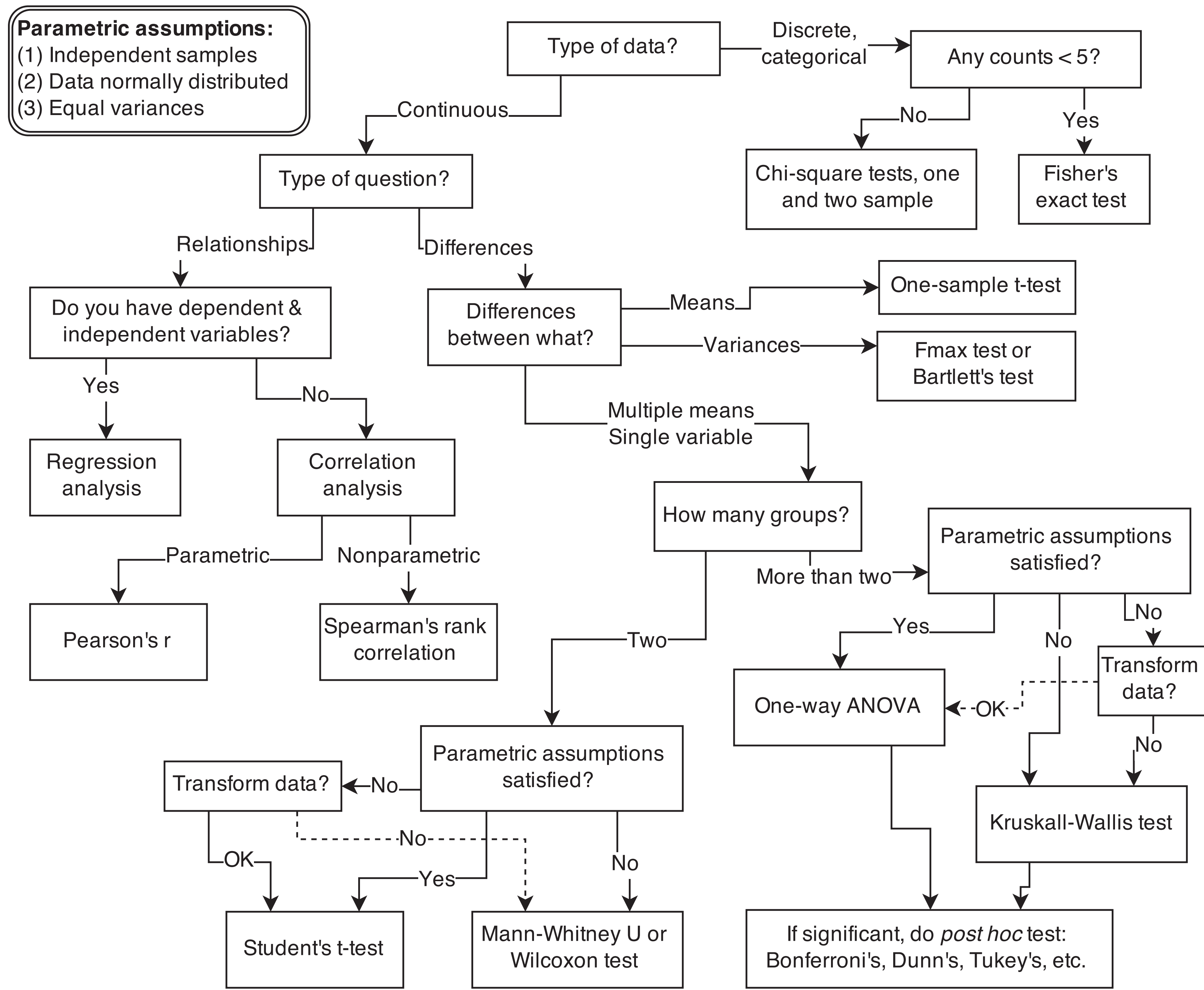


Figure 1.1

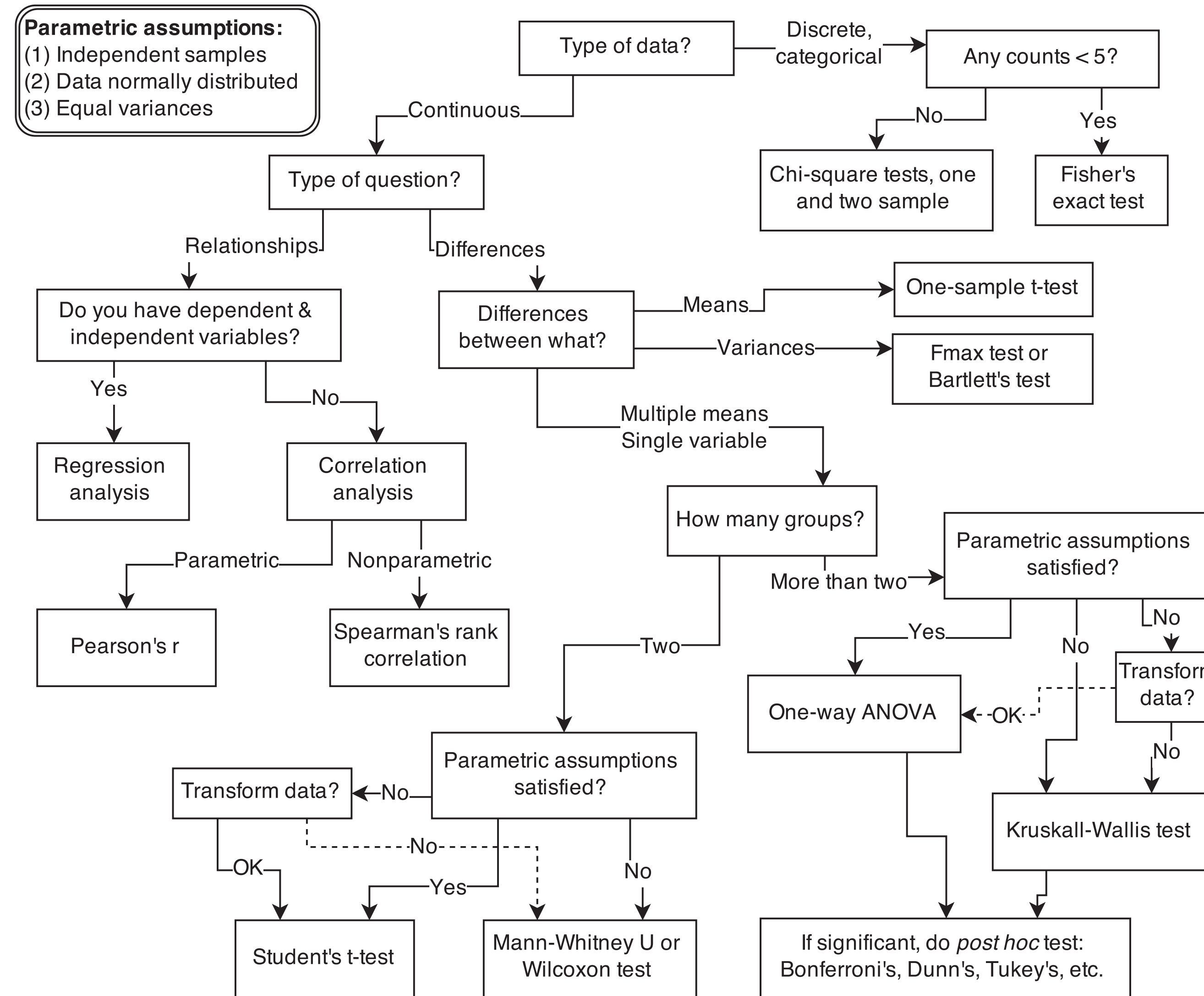
Statistical Models

Incredibly limiting

Focus on rejecting null hypotheses instead of research hypotheses

Relationship between hypothesis and test not clear

Industrial framework



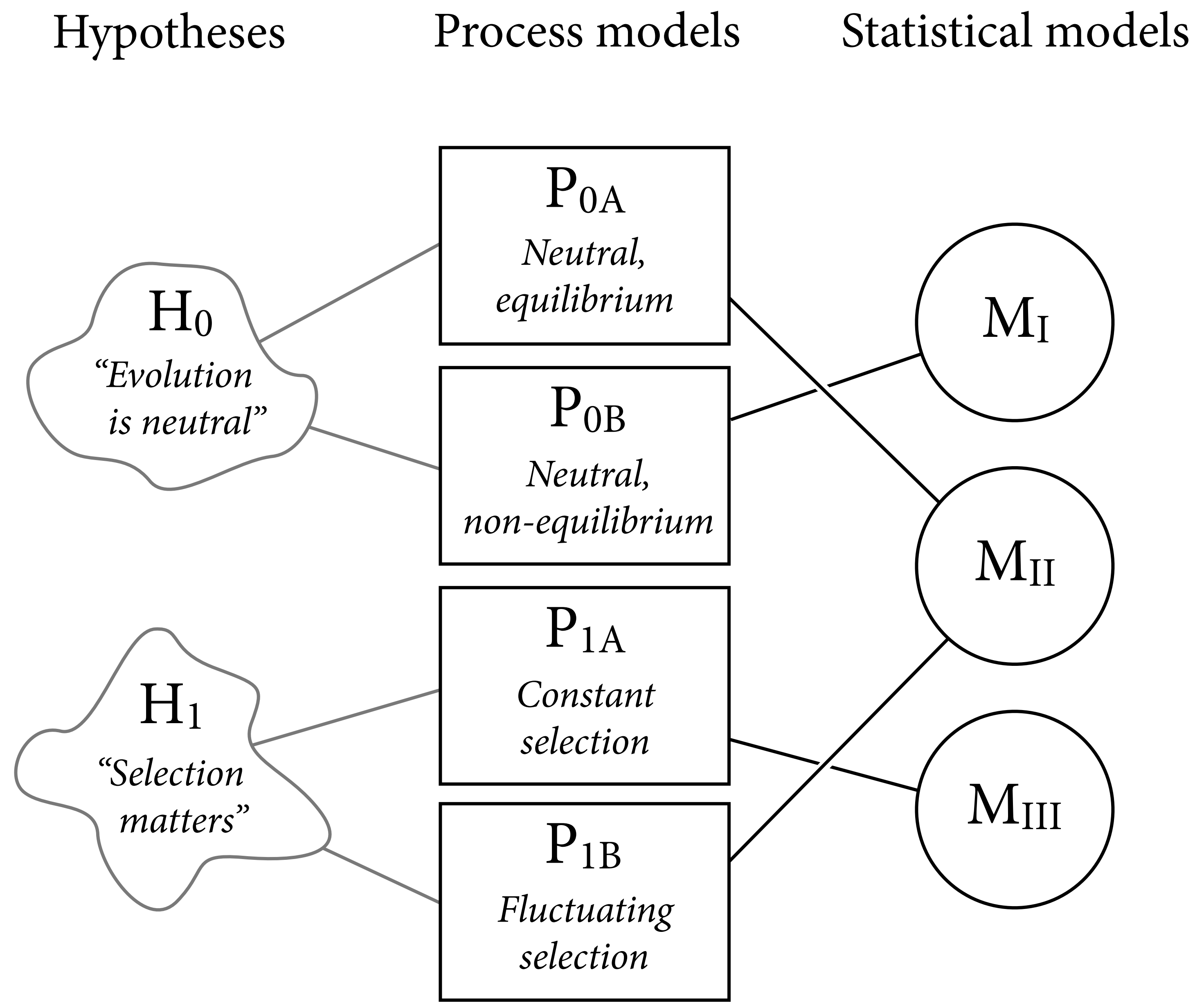


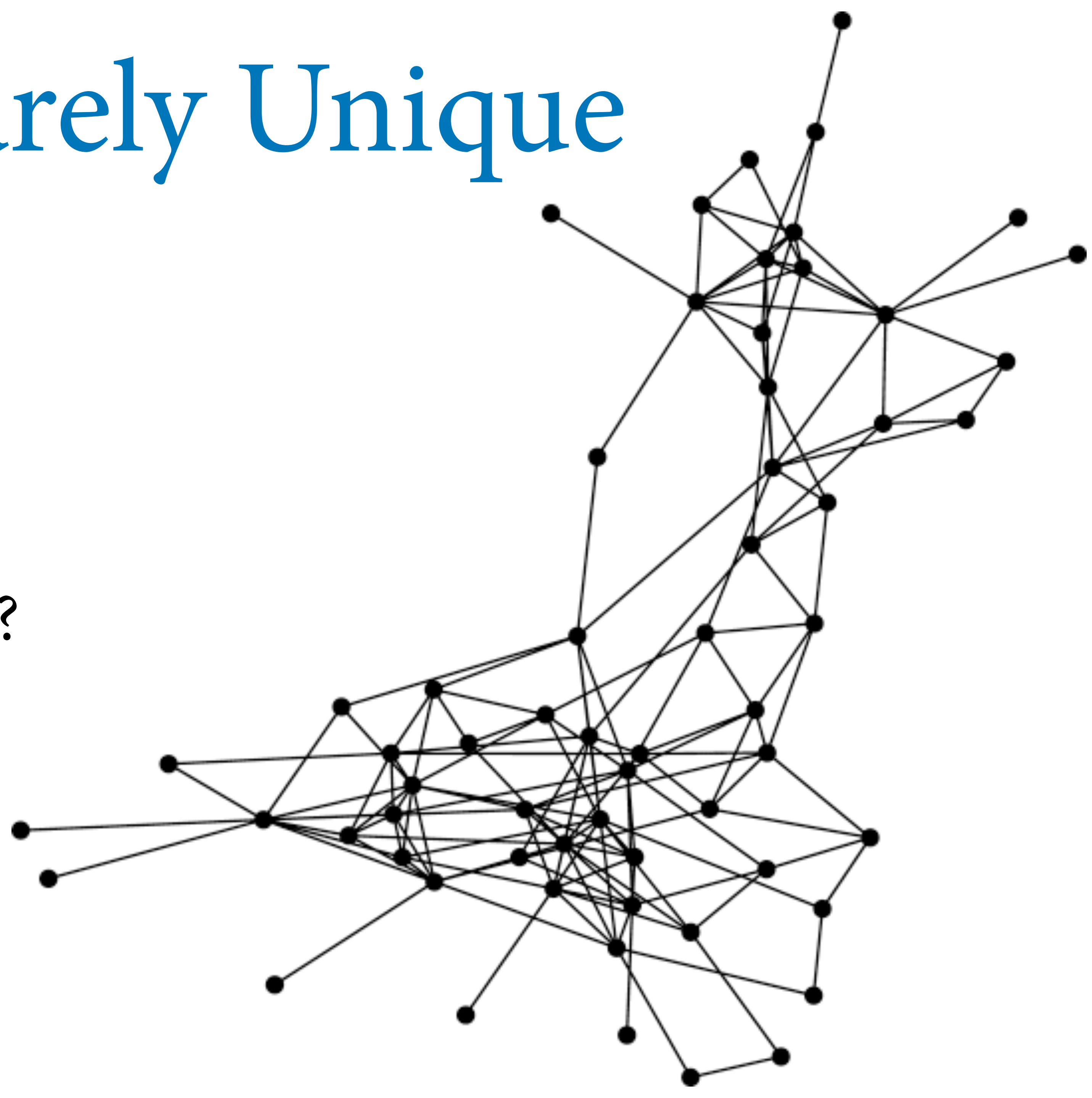
Figure 1.2

Null Models Rarely Unique

Null phylogeny?

Null ecological community?

Null social network?



Hypotheses and Models

Research requires more than than tiny null robots

Also requires:

Precise process model(s)

Statistical model (procedure, golem) justified by implications of process model(s) and question (estimand)

OWLS

HOW TO DRAW AN OWL



1. Draw some circles

HOW TO DRAW AN OWL

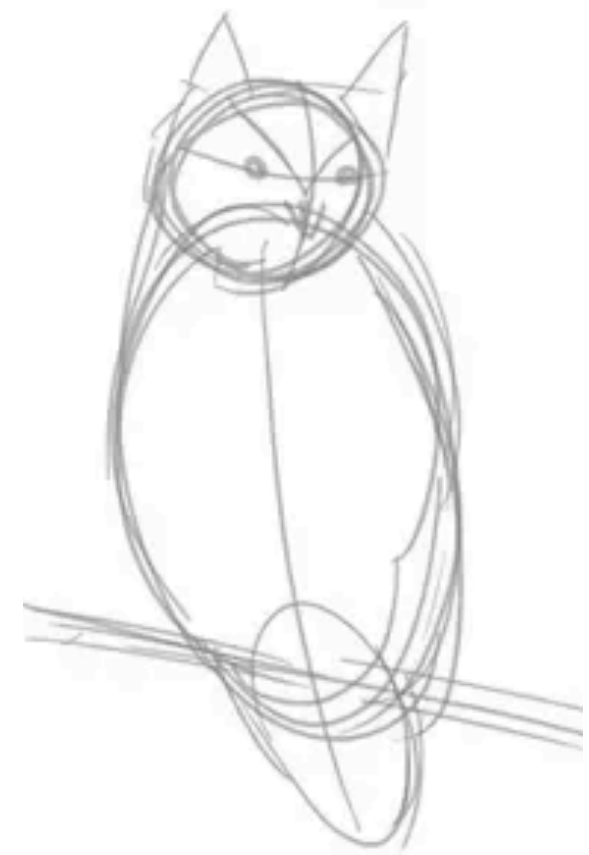


1. Draw some circles



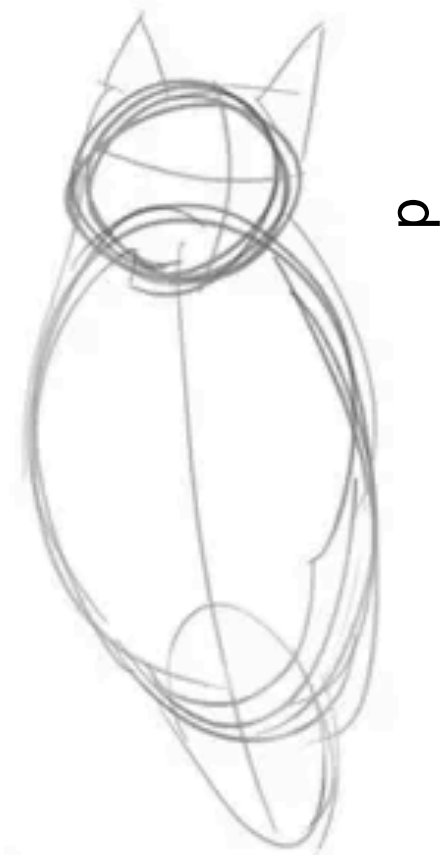
2. Draw the rest of the owl

HOW TO DRAW AN OWL

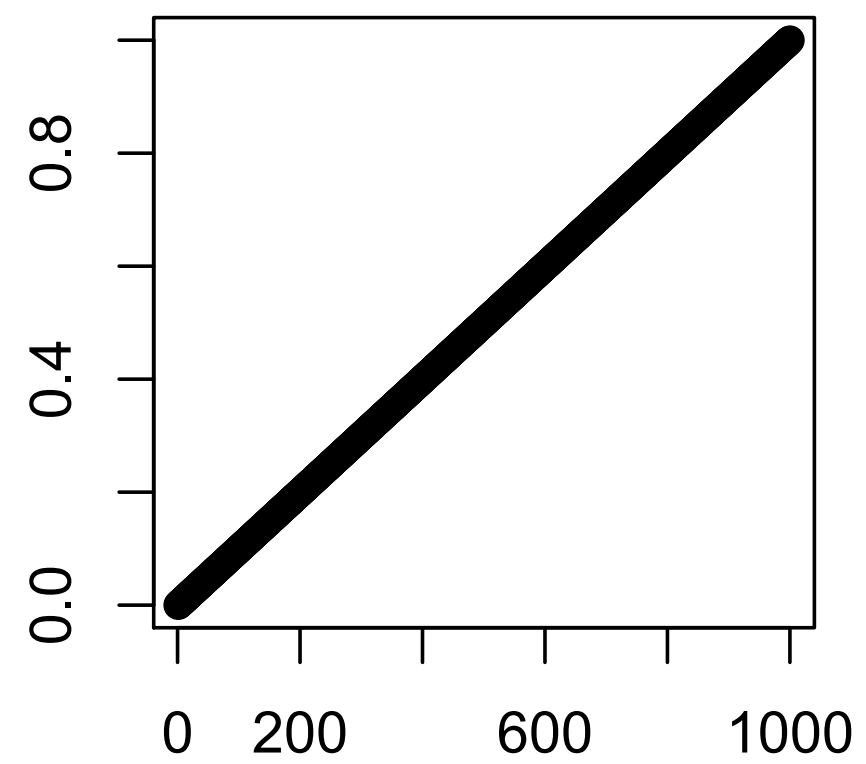


1. Draw some circles

2. Draw the rest of the owl



p

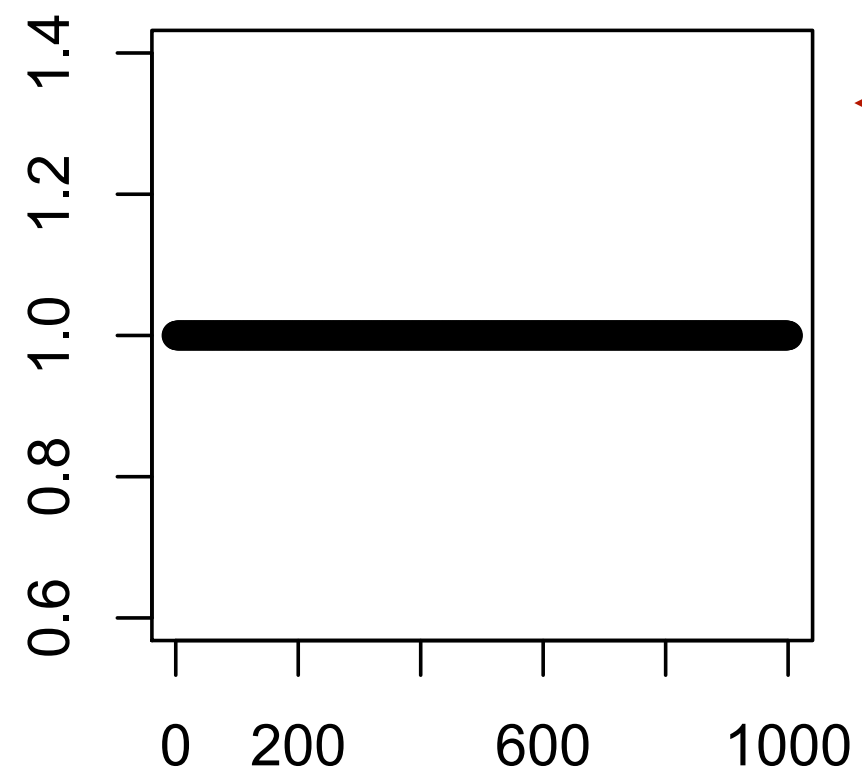


Index

```
p_grid <- seq( from=0 , to=1 , length.out=1000 )  
prob_p <- rep( 1 , 1000 )  
prob_data <- dbinom( 6 , size=9 , prob=p_grid )  
posterior <- prob_data * prob_p  
posterior <- posterior / sum(posterior)
```



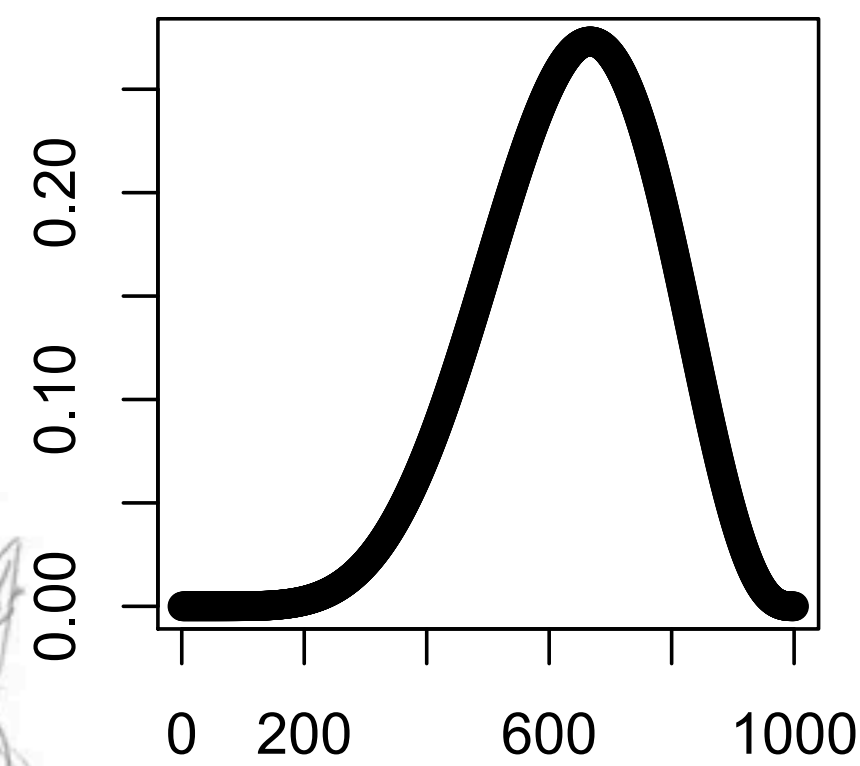
prior



Index

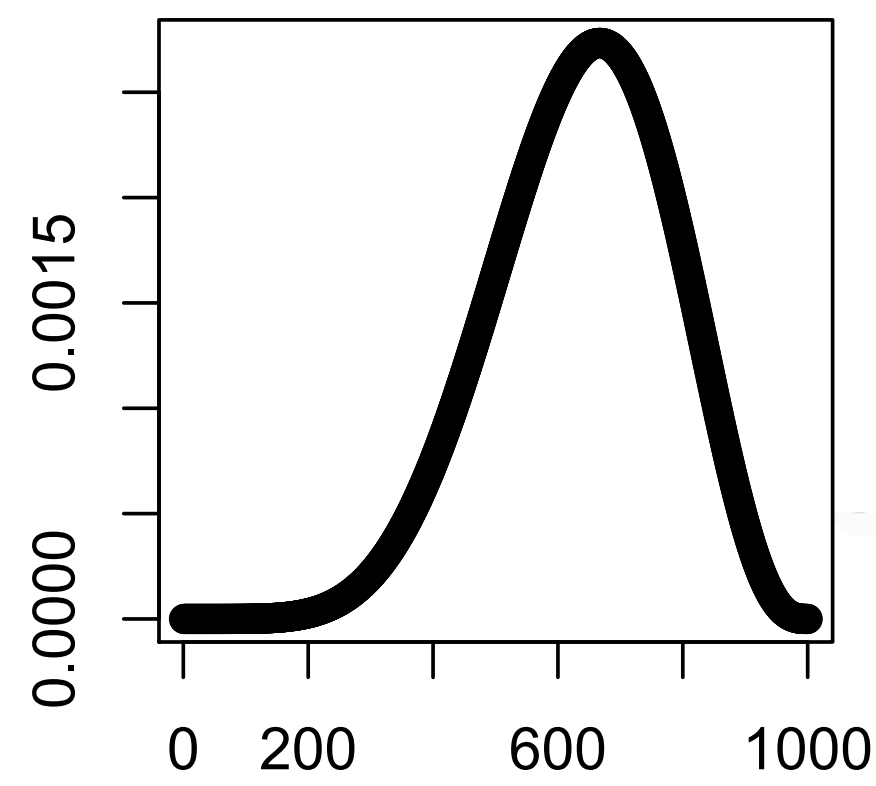


likelihood



Index

posterior



Index



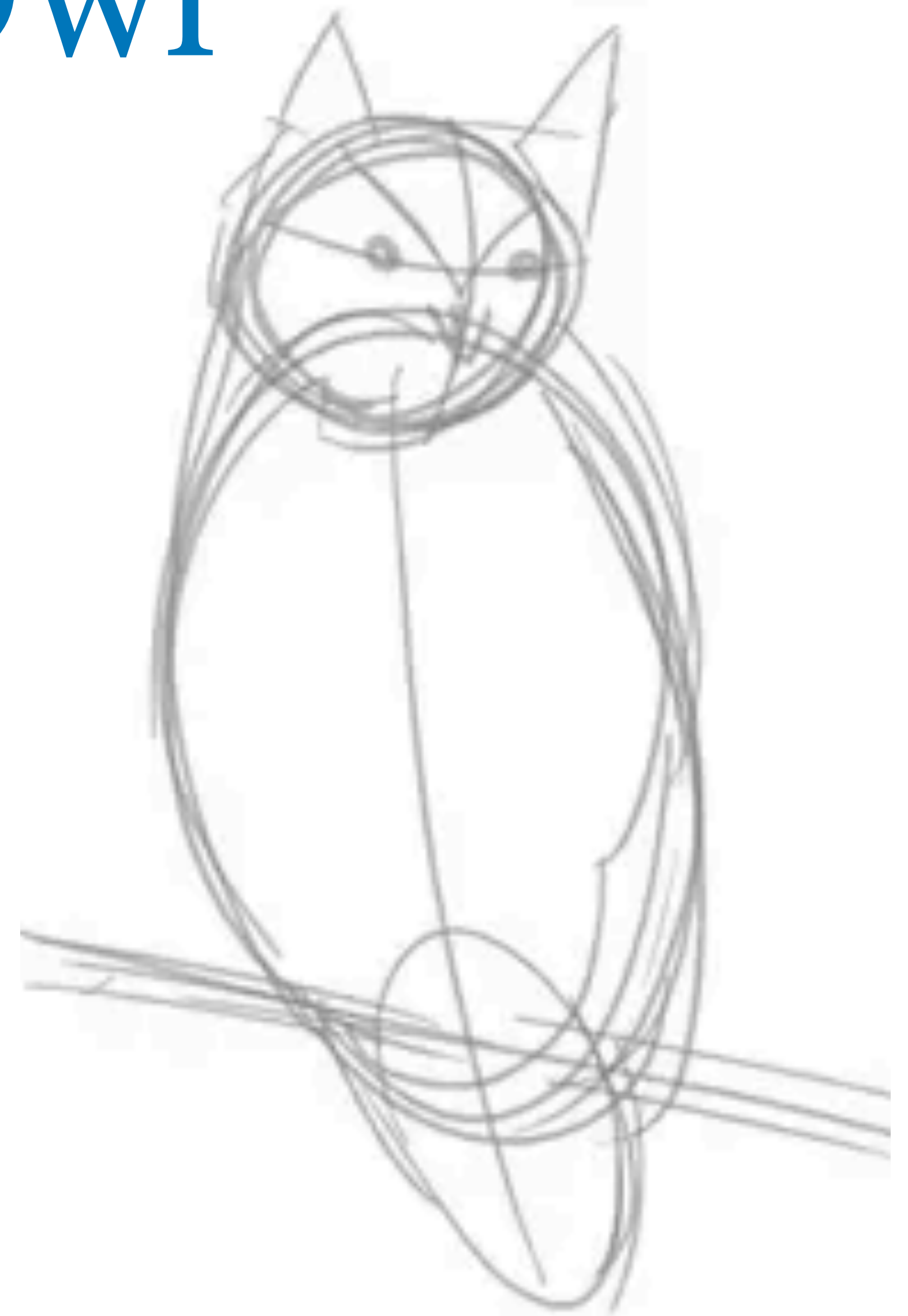
Drawing the Bayesian Owl

Three modes:

Understand what you are doing

Document your work, reduce error

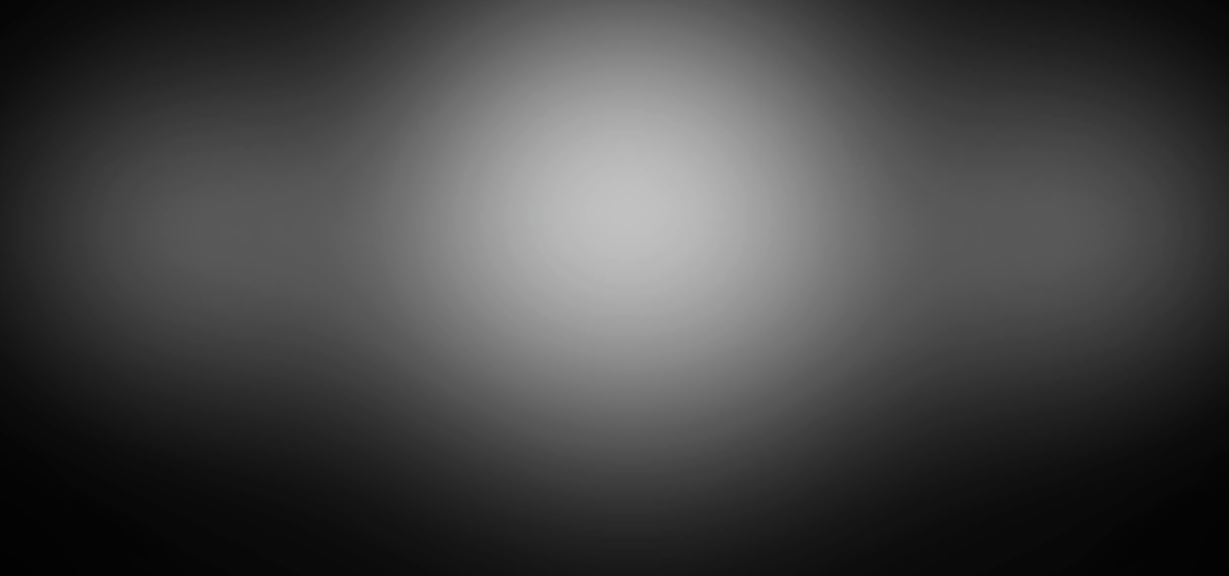
Respectable scientific workflow



Drawing the Bayesian Owl

1. Theoretical estimand
2. Scientific (causal) model(s)
3. Use 1 & 2 to build statistical model(s)
4. Simulate from 2 to validate 3 yields 1
5. Analyze real data





Saturn, Galileo 1610



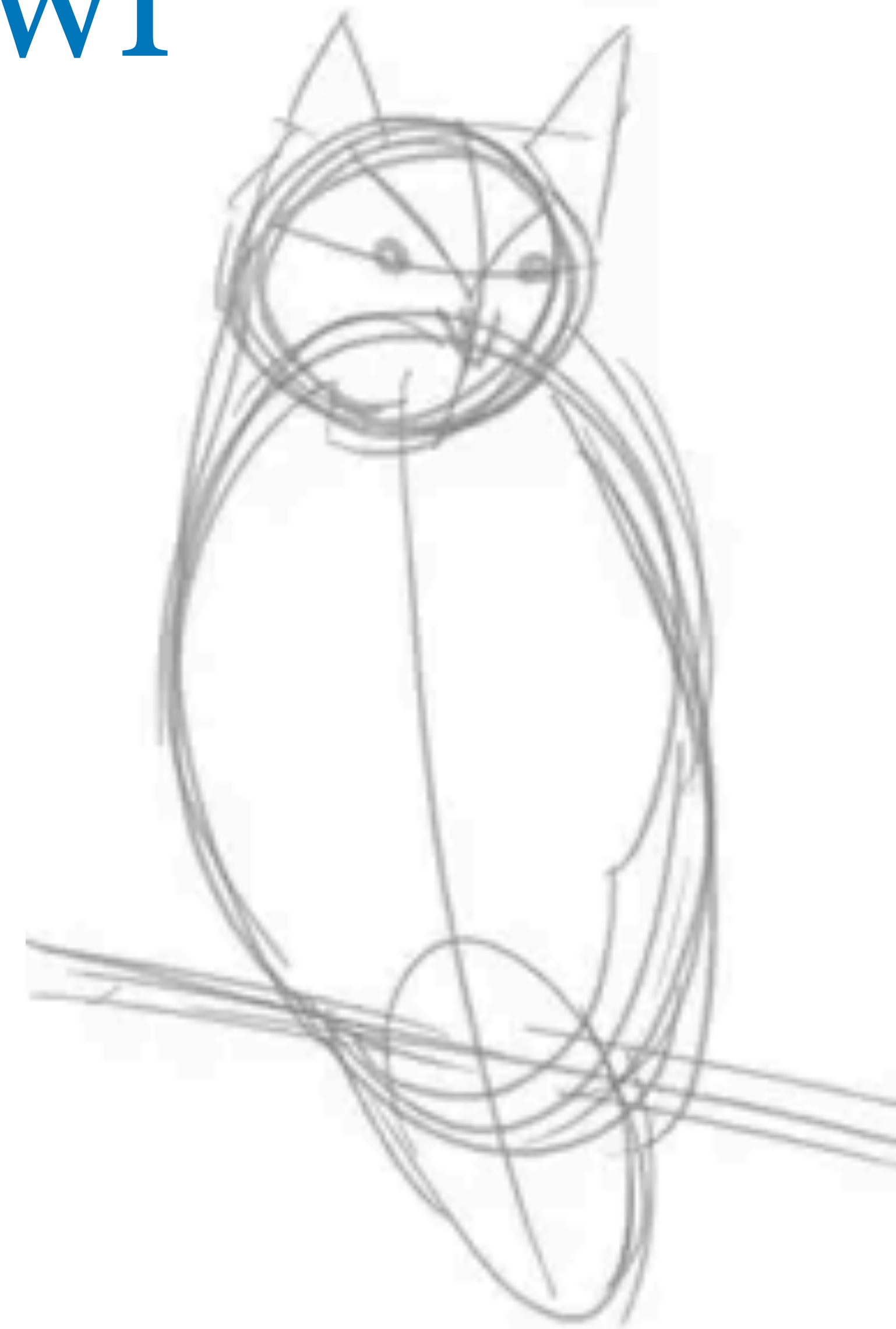
Drawing the Bayesian Owl

Bayesian approach is permissive, flexible

Express uncertainty at all levels

Direct solutions for measurement error,
missing data

Focus on scientific modeling



DAGS

BAYES

FREQUENTISM





**CAUSAL
INFERENCE**

Science Before Statistics

For **statistical models** to produce scientific insight, they require additional **scientific (causal) models**

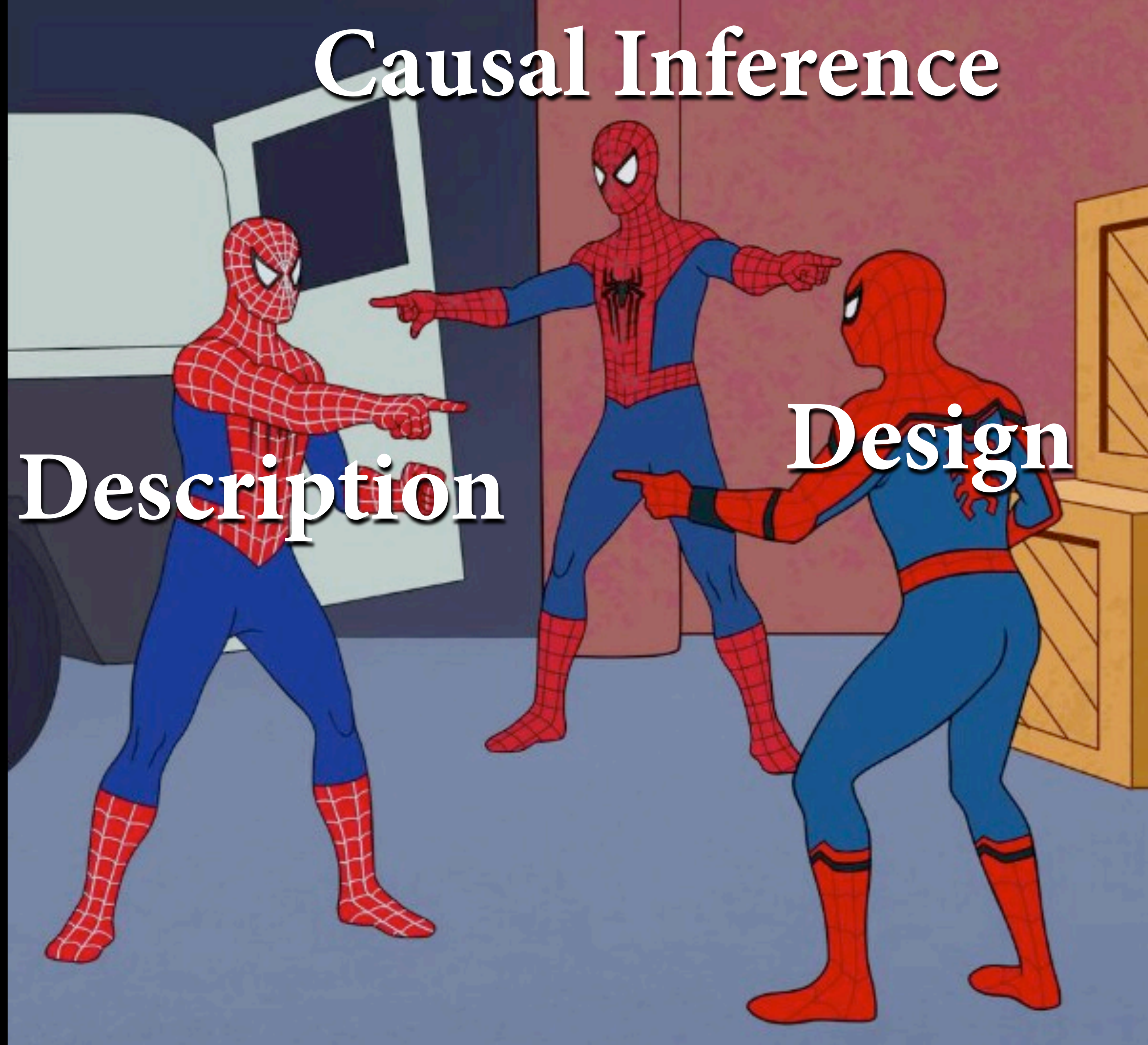
The **reasons** for a statistical analysis are not found in the data themselves, but rather in the **causes** of the data

The **causes** of the data cannot be extracted from the data alone. **No causes in; no causes out.**

Causal Inference

Description

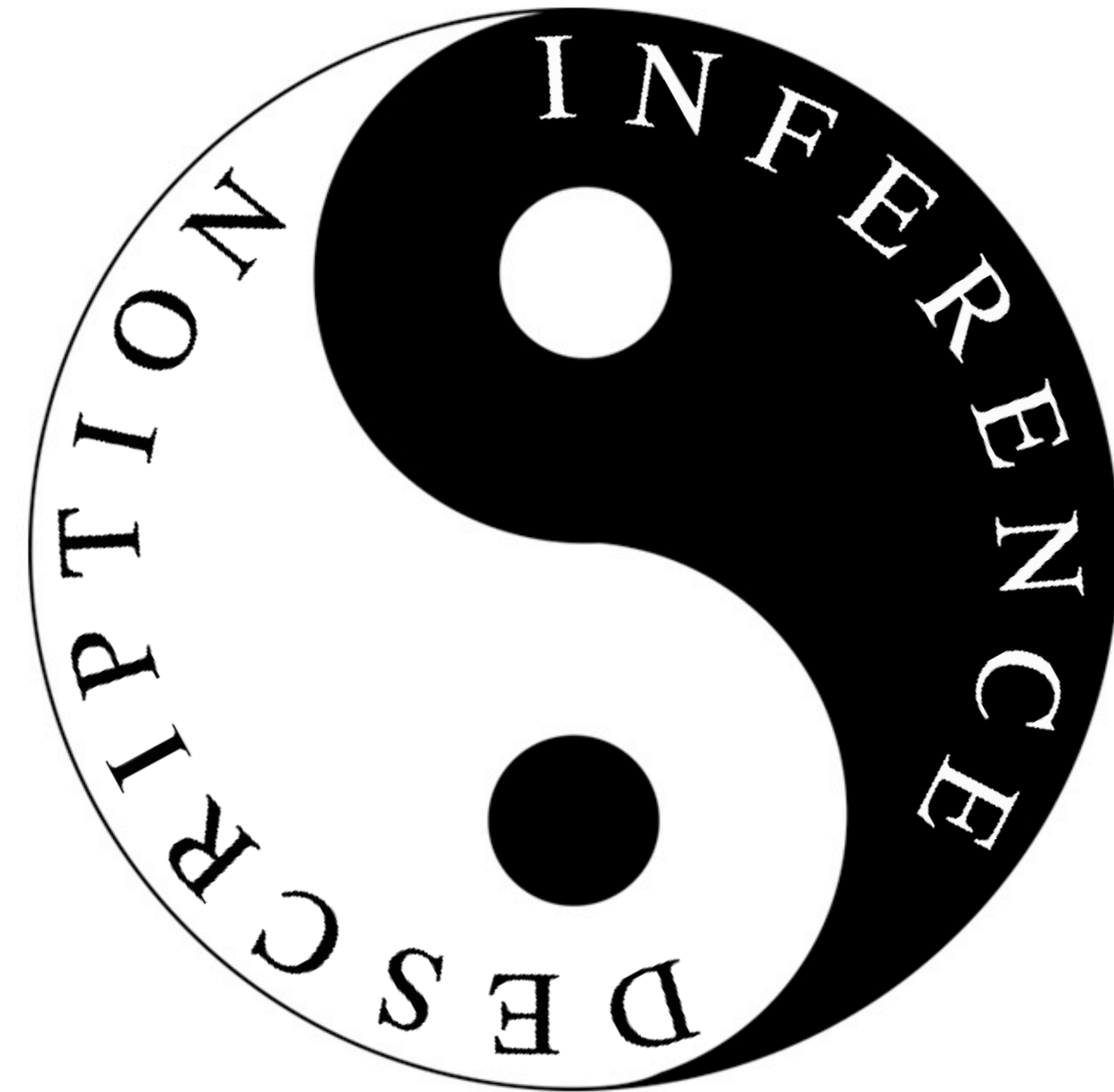
Design



Causes Are Not Optional

Even when goal is **descriptive**, need causal model

The **sample** differs from the **population**; describing the population requires causal thinking



What is Causal Inference?

More than **association** between variables

Causal inference is **prediction** of intervention

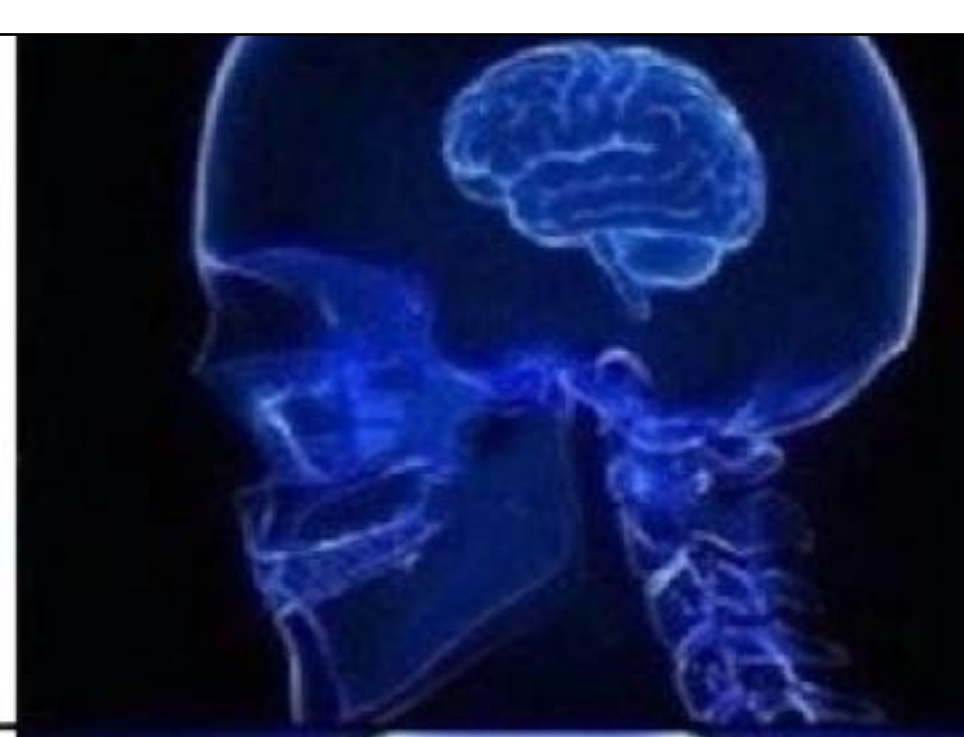
Causal inference is **imputation** of missing observations

**CORRELATION
IMPLIES
CAUSATION**

**CORRELATION
DOES NOT
IMPLY CAUSATION**

**CAUSATION
DOES NOT
IMPLY CORRELATION**

**REALITY IS
A SIMULATION**



Causal Prediction



Knowing a **cause** means being able to predict the **consequences** of an **intervention**.

What if I do this?

Causal Imputation

A black and white photograph of an astronaut on the moon. The astronaut is wearing a full spacesuit with a large life-support backpack and is walking towards the right. To the left of the astronaut, a bright red Chinese flag with five yellow stars is planted in the lunar soil. The background shows the dark, cratered surface of the moon under a black sky.

Knowing a **cause** means being able to construct unobserved **counterfactual outcomes**.

What if I had done something else?

DAGs

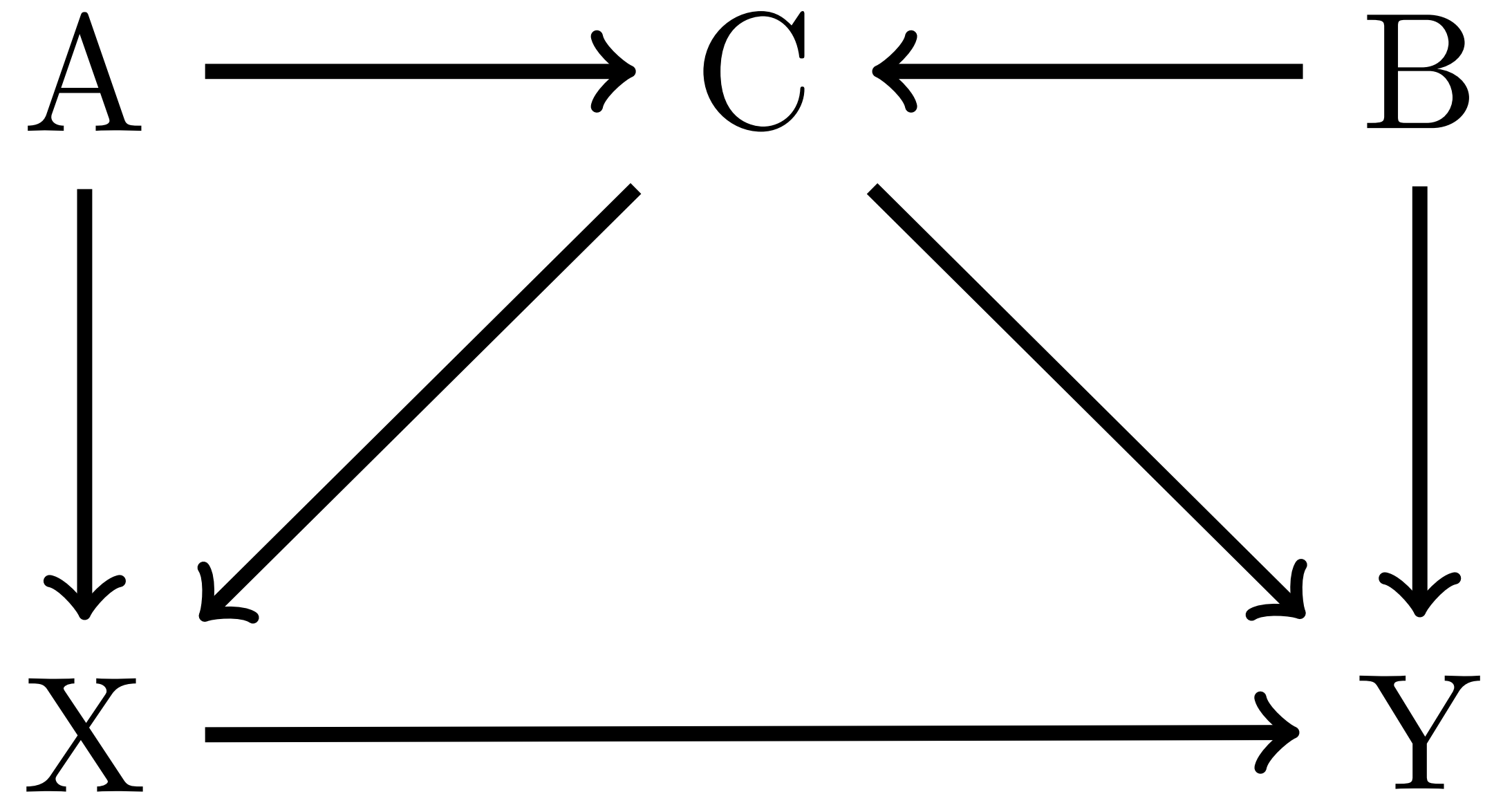
Directed Acyclic Graphs

Heuristic causal models

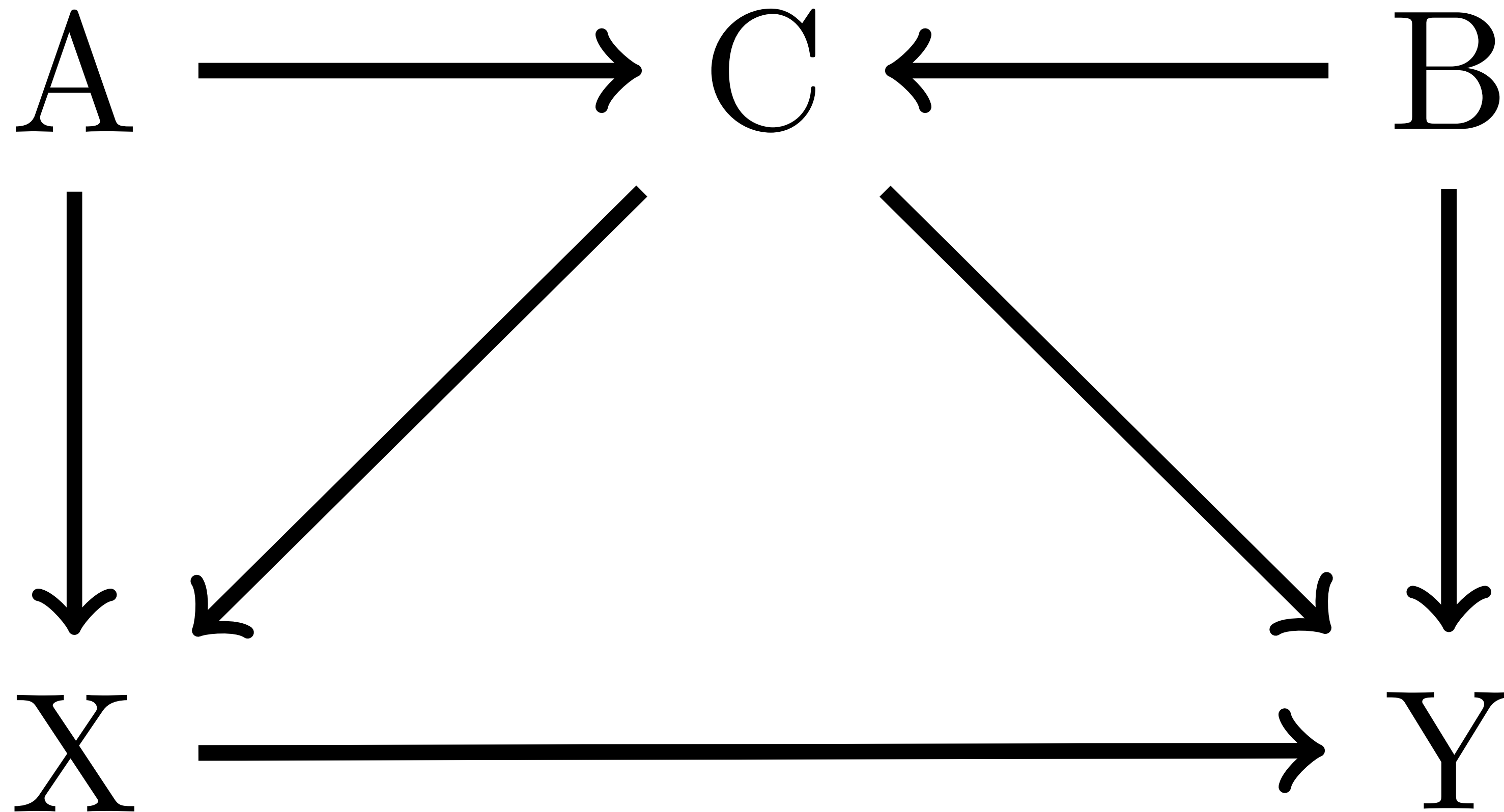
Clarify scientific thinking

Analyze to deduce appropriate statistical models

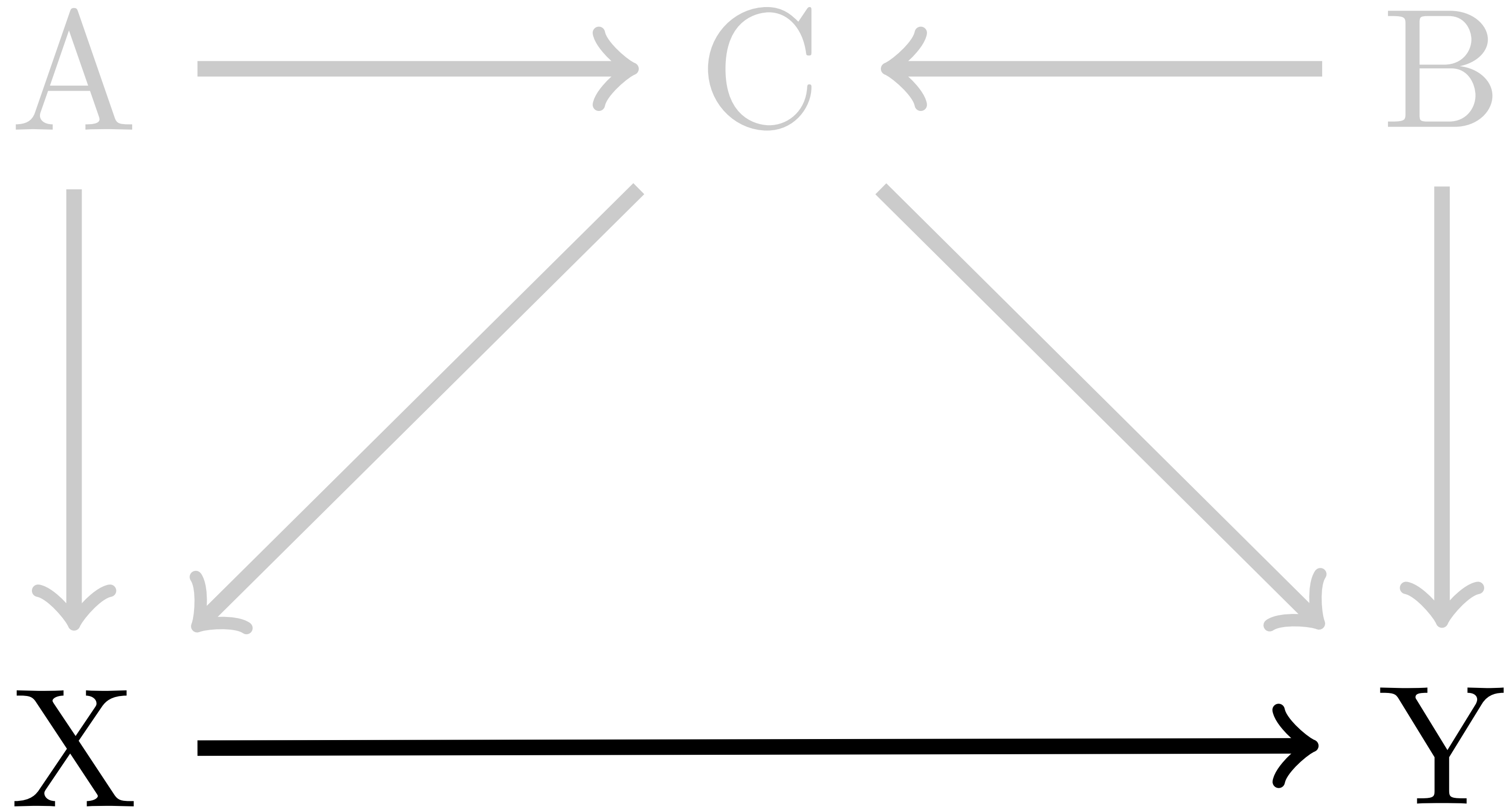
Much more as the course develops



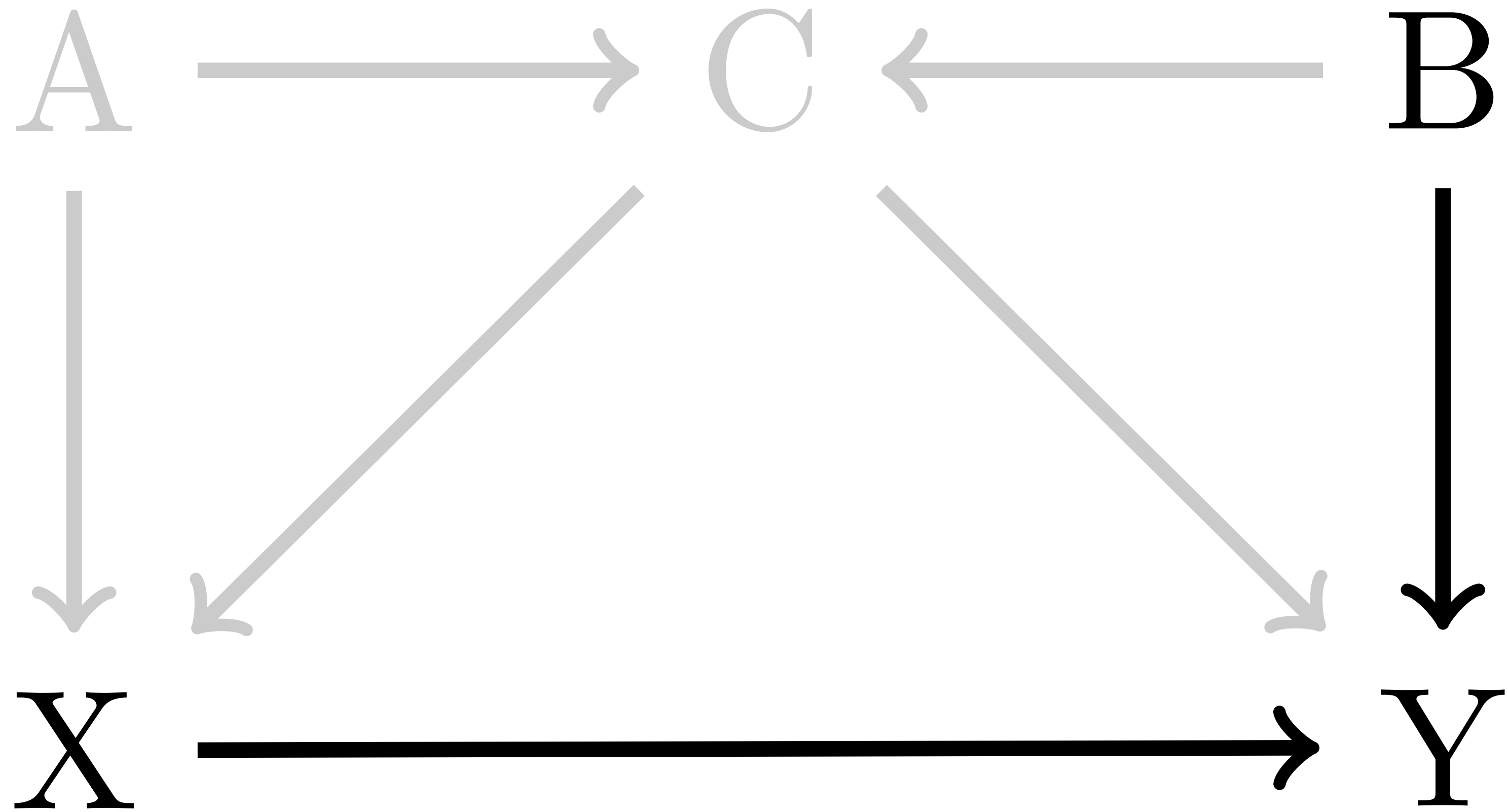
DAGs



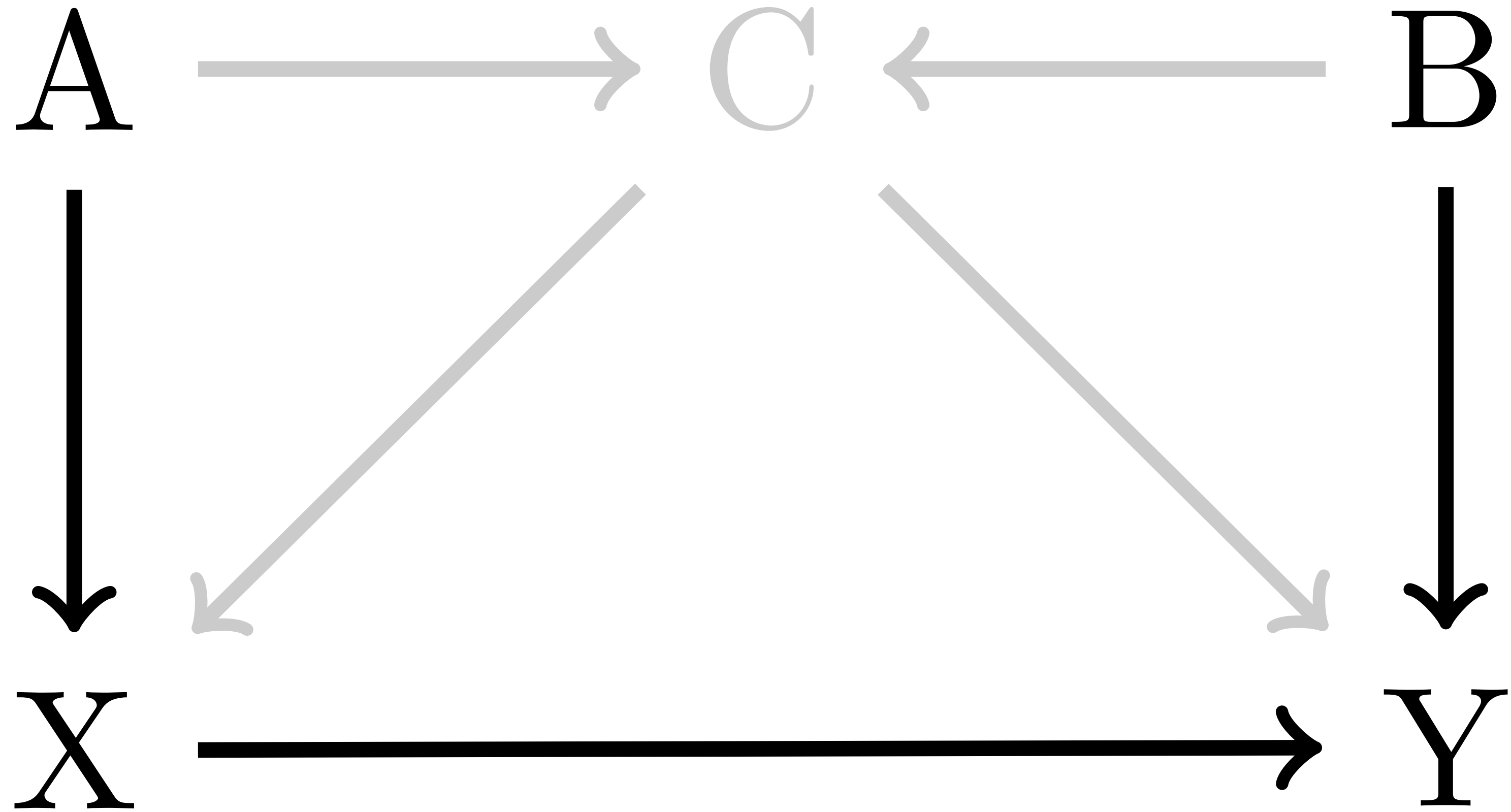
DAGs



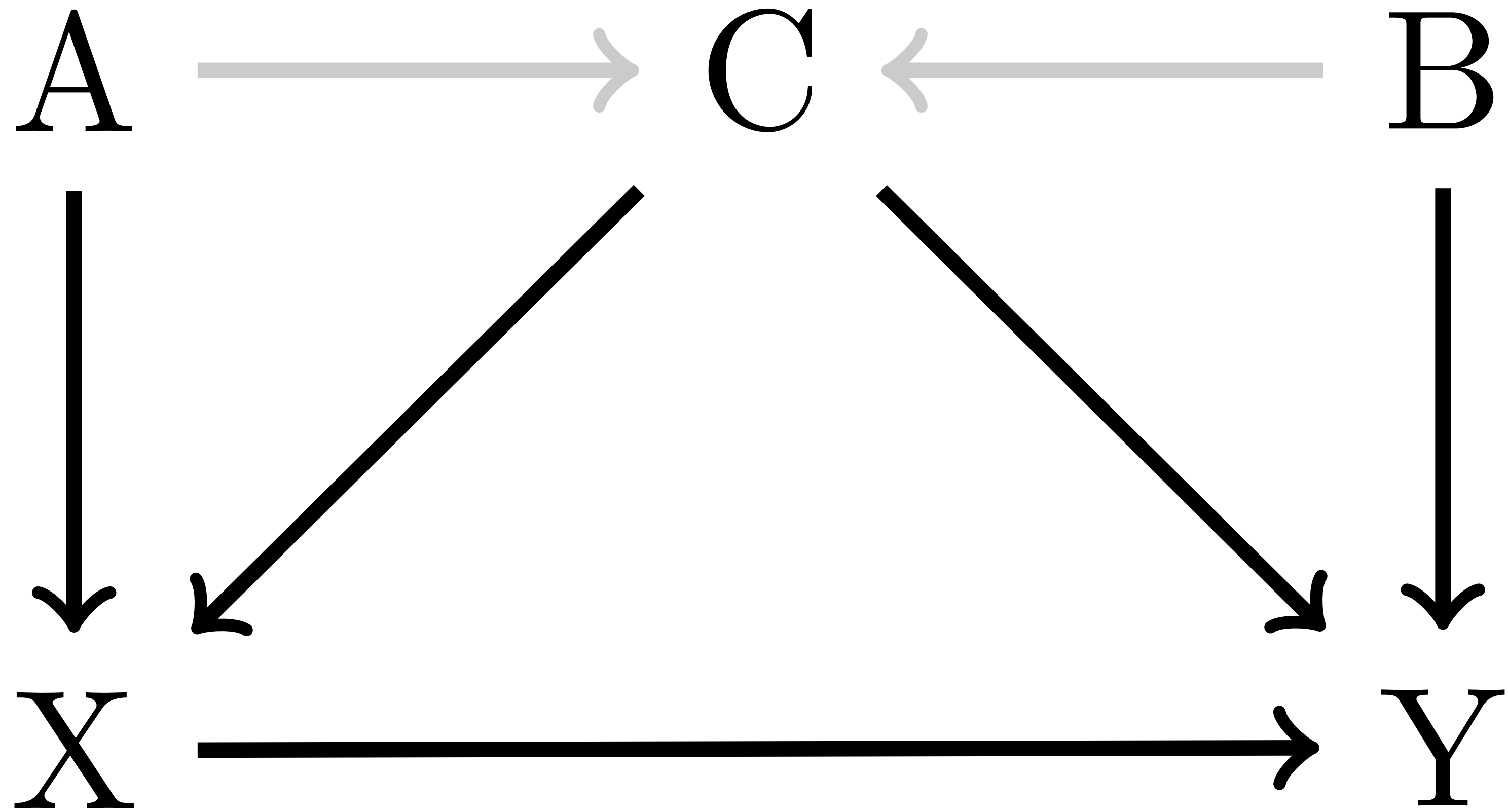
DAGs



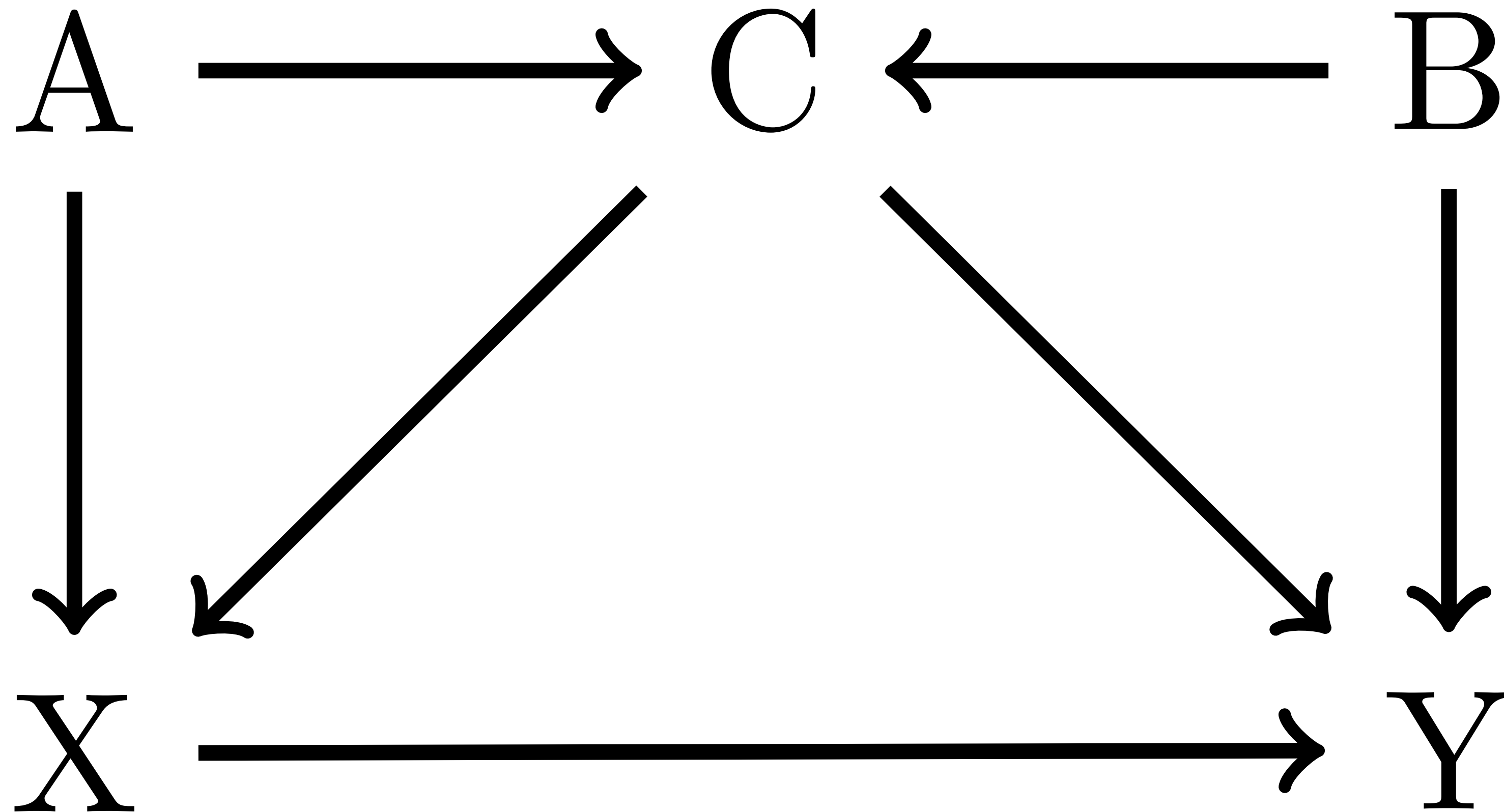
DAGs



DAGs



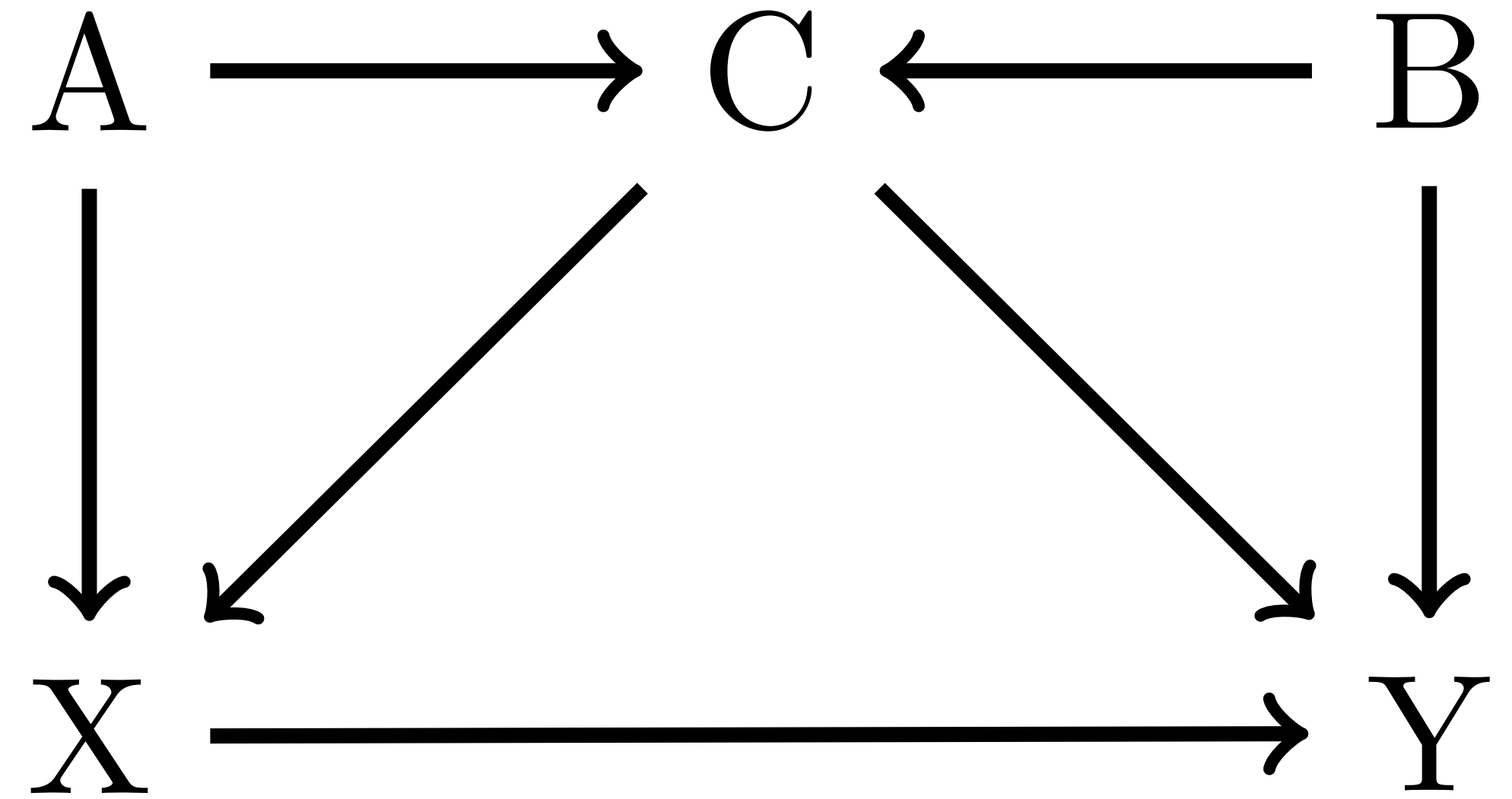
DAGs



DAGs

Different queries, different models

Which control variables?



Absolute not safe to add everything — **bad controls**

How to test the causal model?

With more scientific knowledge, can do more

Golems, Owls, DAGs

Golems: Brainless, powerful statistical models

Owls: Documented, objective procedures

DAGs: Transparent scientific assumptions to

justify scientific effort

expose it to useful critique

connect theories to golems

Course Schedule

Week 1	Bayesian inference	Chapters 1, 2, 3
Week 2	Linear models & Causal Inference	Chapter 4
Week 3	Causes, Confounds & Colliders	Chapters 5 & 6
Week 4	Overfitting / Interactions	Chapters 7 & 8
Week 5	MCMC & Generalized Linear Models	Chapters 9, 10, 11
Week 6	Integers & Other Monsters	Chapters 11 & 12
Week 7	Multilevel models I	Chapter 13
Week 8	Multilevel models II	Chapter 14
Week 9	Measurement & Missingness	Chapter 15
Week 10	Generalized Linear Madness	Chapter 16

https://github.com/rmcelreath/statrethinking_2022