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# Bayesian Statistics

## Lecture 6

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# Learning objective

- Have a basic understanding of Bayesian statistics
- How to define a prior

# Bayesian statistics updates belief with data

Belief?

Update?

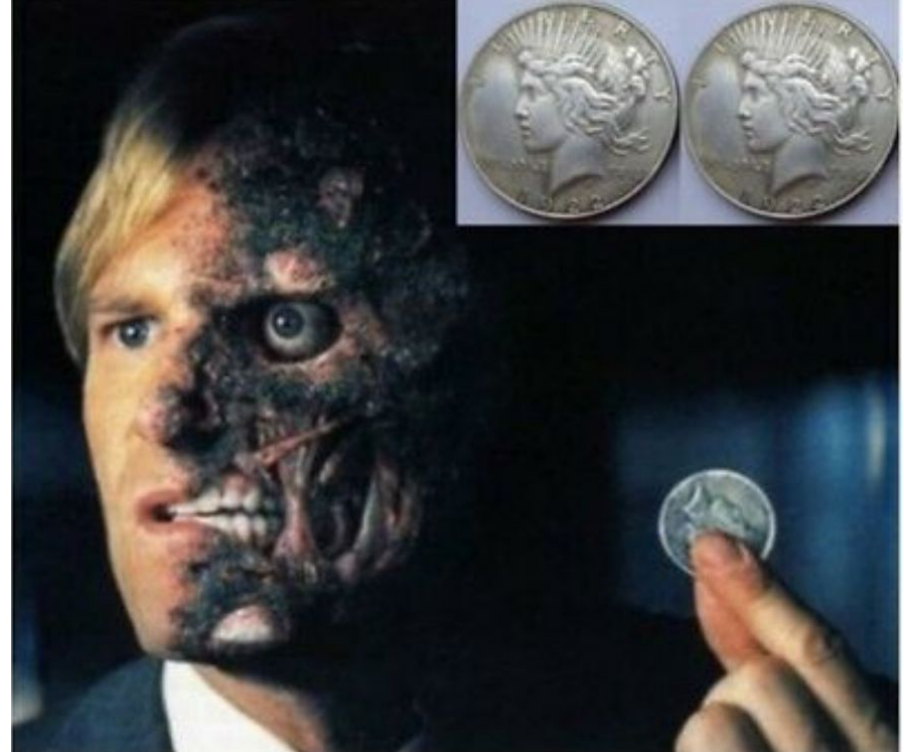
# Thought experiment 0

Imagine 2 coins:

- Biased coin - both sides are Heads
- Fair coin

Someone chooses one coin and tosses it.

What's the chance it's the fair coin?



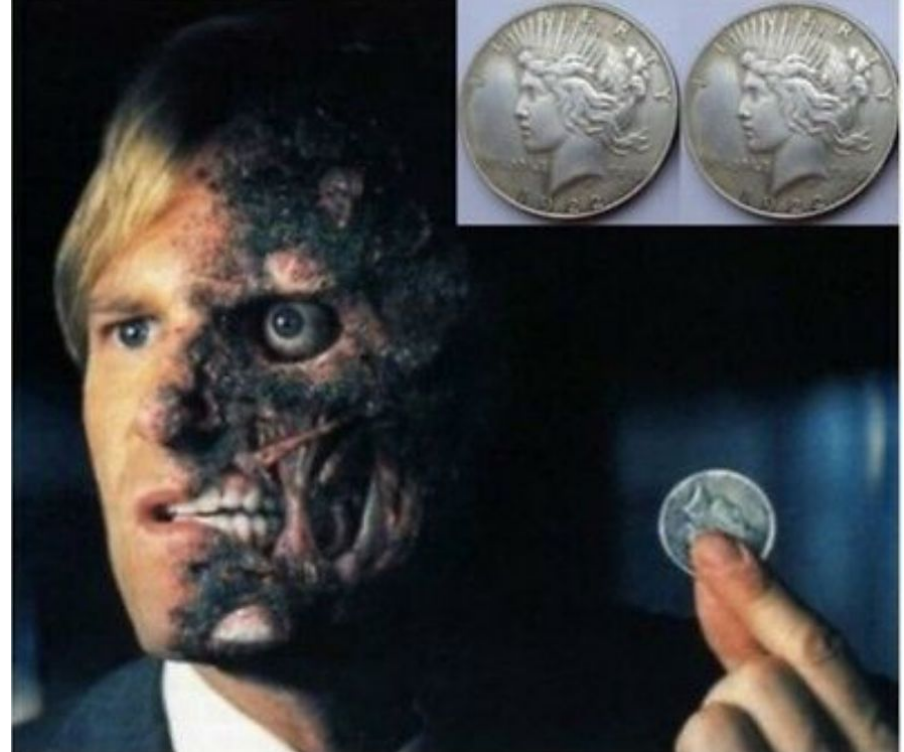
# Thought experiment 1

Imagine 2 coins:

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Someone randomly chooses one coin and tosses it.

What's the chance it's the fair coin?



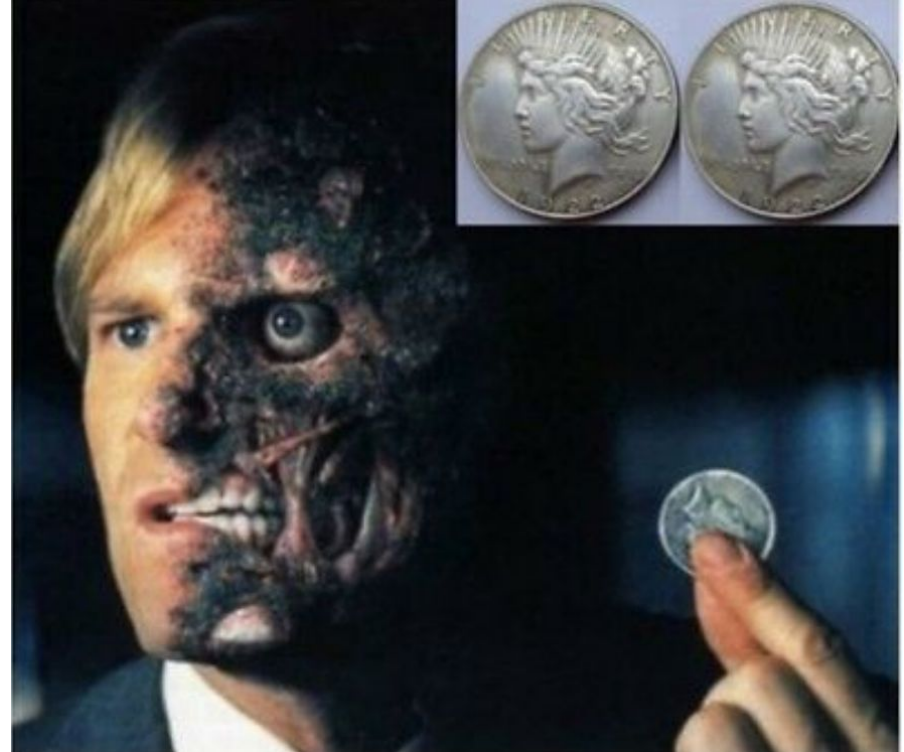
# Thought experiment 2

Imagine 2 coins:

- Biased coin - both sides are Heads
- Fair coin

Someone randomly chooses one coin and tosses it.

You see a **Tail**, what's the chance it's the fair coin?



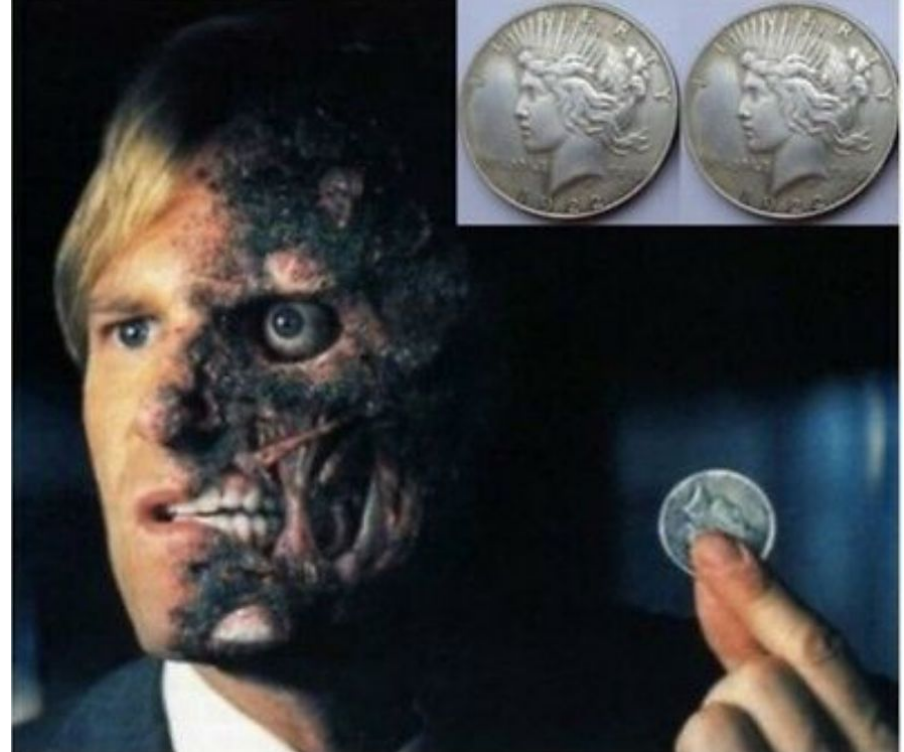
# Thought experiment 3

Imagine 2 coins:

- Biased coin - both sides are Heads
- Fair coin

Someone randomly chooses one coin and tosses it.

You see a **Head**, what's the chance it's the fair coin?



# Setting the problem up - what do we know



## Setting the problem up - what do we know

$$P(\text{toss=Heads}|\text{coin is fair}) = 1/2$$

$$P(\text{toss=Heads}|\text{coin is biased}) = 1$$

$$P(\text{coin is biased}) = P(\text{coin is fair}) = 1/2$$

## Setting the problem up - what are we solving?

$$P(\text{coin is fair} | \text{toss} = \text{Heads}) = ?$$

$$P(\text{toss} = \text{Heads} | \text{coin is fair}) = 1/2$$

$$P(\text{toss} = \text{Heads} | \text{coin is biased}) = 1$$

$$P(\text{coin is biased}) = P(\text{coin is fair}) = 1/2$$

# Recalling conditional probability

$$\begin{aligned}P(\text{coin is fair}|\text{toss=Heads}) &= \frac{P(\text{coin is fair AND toss=Heads})}{P(\text{toss=Heads})} \\&= \frac{P(\text{coin is fair})P(\text{toss=Heads}|\text{coin is fair})}{P(\text{toss=Heads})}\end{aligned}$$

$$P(\text{toss=Heads}) = 0.5 * 0.5 + 0.5 * 1 = 0.75$$

# General setup for Bayesian problem - choosing the likelihood and prior

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- $P(X \mid \text{Data})$

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# Prior is a special quantity in Bayesian Statistics

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**Prior:** before seeing any data, what is my belief about X?

- $P(X)$

# What's the big deal?

Imagine if the 2 different coins were 2 different hypotheses?



# Practice - setup the Bayesian problem

You tossed a coin 1 time, it's shows up TAILS, what is its chance to land a H?

- What is the problem?
- What is the likelihood?
- What is the prior?

See notebook on Canvas “bayesian-intro.ipynb”!

# How else could we estimate this?

You tossed a coin 1 time, it's shows up TAILS, what is its chance to land a H?

What chance would maximize the probability of seeing a TAILS from 1 toss?

# Pro/Con of Bayesian Methods

## Con

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- Can be computationally difficult
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## Pro:

- Very flexible and intuitive (e.g. what we perceive as chance)
- Can introduce prior knowledge easily
- Can update as you see more data
- Very comprehensive framework

# Notes on “Bayes” in machine learning

- Bayesian statistics is often used as a framework for motivating certain calculations
- Bayesian statistics can “regularize” some classic statistics