

# Hypothesis Testing Framework

## Coin Toss :

10-match cricket series  
Captain always calls "heads"  
Won 7 tosses

Fair

22

Biased

3

100-matches  
"heads"  
Won 70 tosses

Fair

11

Biased

22

1000-matches  
"heads"  
Won 700 tosses

Fair

7

Biased

26

Framework to provide a quantifying metric  
to this intuition

Coin toss example:

① What was our default assumption about the coin?

"Coin is fair"

② When will we reject this assumption?

"If we have enough proof / evidence / data"

Judge in a Court → A suspect is brought in front of you

① What is our default assumption?

"He is innocent"

② When will you reject this assumption?

"When there is enough evidence"

Innocent until proven guilty  
default assumption      evidence to reject

Machine Learning → Deploy (Eg: Youtube / Amazon)  
Amazon recommends some similar items  
(existing ML model → "legacy code" model)

New model to improve performance

- ① What is the default assumption of the Product Dev?  
"New model is not better than legacy model"
- ② When shall we reject this assumption?  
"When we have enough evidence to show that new model is better"

Third umpire : ① on-field has called third umpire  
② "Soft signal"

① What is the default assumption of the 3<sup>rd</sup> umpire?

"On-field umpire's soft signal"

② When will he reject this assumption?

"When there is enough evidence to overrule the soft signal"

# ① Null Hypothesis &

↓ "Coin fair"  
↓ "Person is innocent"  
default assumption "Legacy is better"  
"On-field"

# Alternate Hypothesis

↓ "Coin is biased"  
↓ "guilty"  
↓ "new model is better"  
↓ "On-field is wrong"  
Burden of proof

$H_0 \rightarrow$  Null Hypothesis (default assumption)

$H_a \rightarrow$  Alternate

Recall Tidy example:

$H_0$ : person is innocent  
 $H_a$ : person is guilty  $\rightarrow$  Burden of proof

Data

Person has a knife in his pocket

Innocent people can carry a knife

Knife has blood stain

Cook / Chef / Butcher

Blood matches victims

This is too much!

CCTV  $\rightarrow$  scene of crime

Shirt has fingerprints of victims

Strong motive to kill

} Highly unlikely that innocent people have these data points

Probability of seeing such data under the assumption of innocence is very low

$P[\text{data} \mid \text{innocent}]$  is very low

$P[\text{data} \mid H_0]$  "p-value"

If p-value is small  $\rightarrow$  reject  $H_0$   
(pronounce guilty)



## Deep dive → Coin Toss

10-matches, "Heads", won 7 tosses

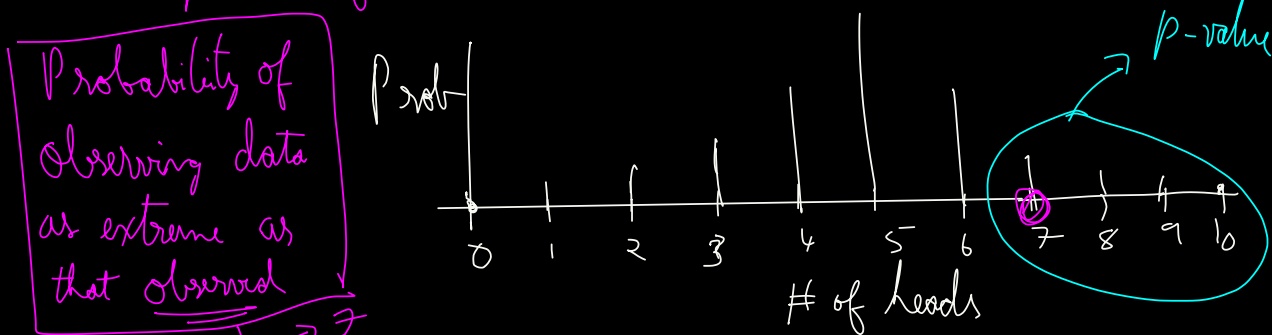
$H_0$ : coin is fair

$H_a$ : Biased towards heads

$T$ : Test statistic (function of data/observation)

$T = \text{no. of heads} \rightarrow \text{Random variable} \rightarrow \text{distribution}$   
Binomial

In p-value, we compute prob under the  
assumption of  $H_0$  pmf under assumption of  $H_0$



$$\begin{aligned} P[T \geq 7 \mid H_0] &= 1 - P[X \leq 6 \mid H_0] \\ &= 1 - \text{binom.cdf}(k=6, n=10, p=0.5) \\ &= 0.172 \end{aligned}$$

$$p\text{-value} = 0.172$$

If  $p\text{-value} < \text{"significance level"} (\alpha)$   
Reject  $H_0$

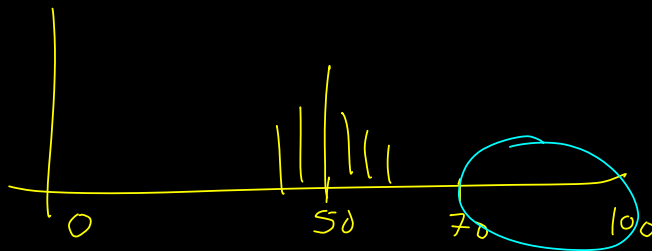


100 matches  $\rightarrow$  70 heads

$H_0$ : Coin is fair

$H_a$ : Coin is biased  $\rightarrow H$

$T$ : no of heads  $\rightarrow$  binomial



$$\begin{aligned} p\text{-value} &\rightarrow P[T \geq 70 \mid H_0] = 1 - P[T \leq 69 \mid H_0] \\ &= 1 - \text{binom.cdf}\left(\begin{matrix} k=69 \\ n=100 \\ p=0.5 \end{matrix}\right) \\ &= 3.9 \times 10^{-5} \end{aligned}$$

$$\alpha = 0.05$$

$p\text{-value} < \alpha$  ?  $\rightarrow$  Yes

Reject  $H_0 \rightarrow$  "Coin is biased"

1000 tosses  $\rightarrow$  700 heads

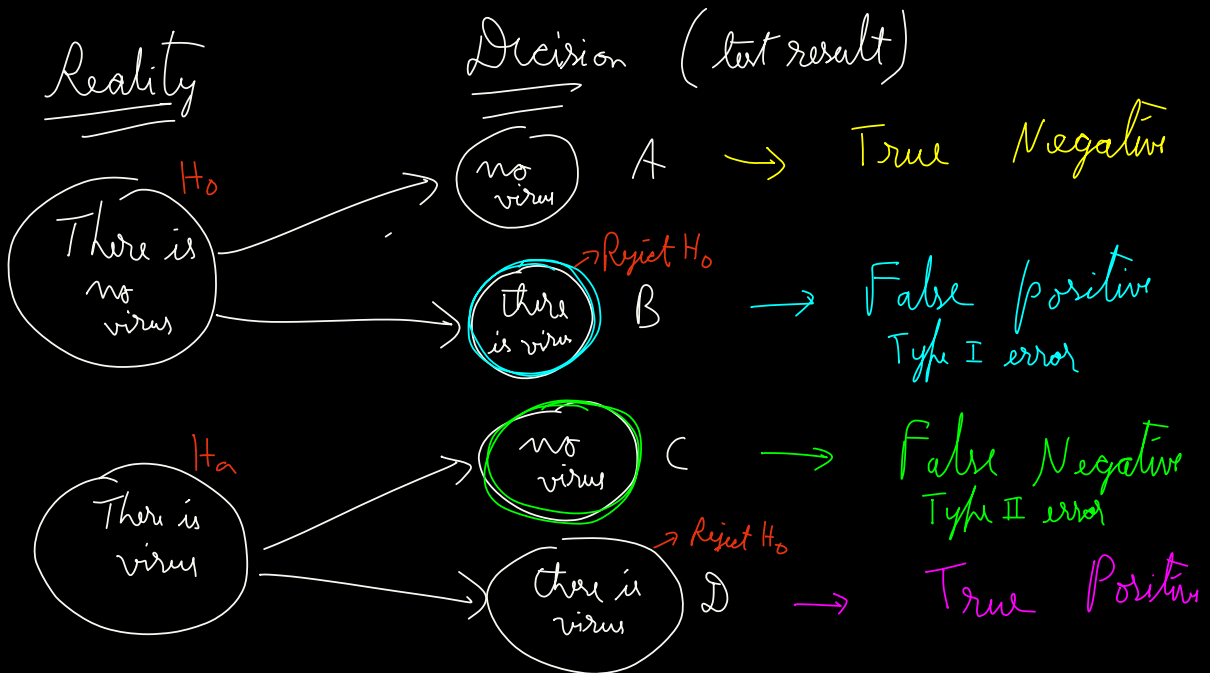
$$\begin{aligned} P[T \geq 700 \mid H_0] &= 1 - \text{binom.cdf}\left(\begin{matrix} k=699 \\ n=1000 \\ p=0.5 \end{matrix}\right) \\ &= 10^{-16} \end{aligned}$$

$p\text{-value} < 0.05 \rightarrow$  Reject  $H_0$

# Test for virus

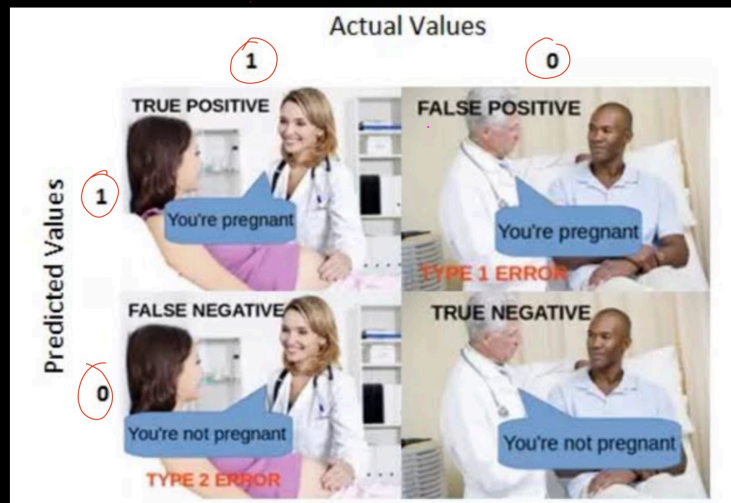
$H_0$ : no virus (negative)

$H_a$ : There is virus (positive)

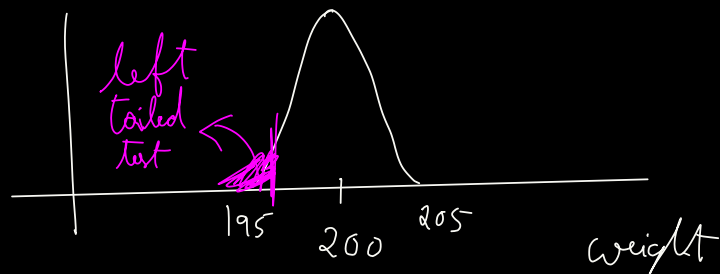


$H_0$ : not pregnant

$H_a$ : pregnant



Burger : Company claims that their  
burger weighs 200 gm on average



Unhappy customer, still hungry, disprove  
the company's claim  $\rightarrow$  weight  $< 200$

$$H_0: \mu = 200 \quad (\mu \rightarrow \text{avg weight})$$

$$H_a: \underline{\underline{\mu < 200}} \quad \text{left side of the mean}$$

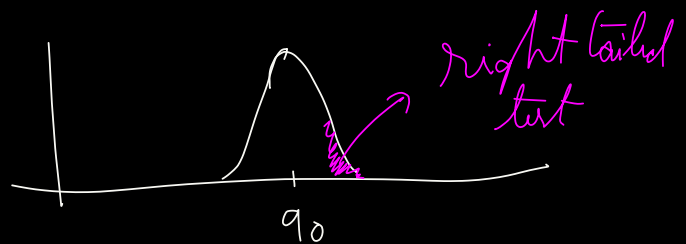
ML deployment : Legacy model  $\rightarrow$  90% accuracy on average

New model  $\rightarrow$  prove that new model is better.

$\mu \rightarrow$  avg accuracy of new model

$$H_0: \mu = 90 \text{ (or } \mu \leq 90)$$

$$H_a: \mu > 90$$



Height : India avg  $\rightarrow$  65 inch

Your state  $\rightarrow$  avg height " $\mu$ "

$$H_0 : \mu = 65$$

$$H_a : \mu \neq 65$$

