# Cricket series example

# The Captain always calls heads

1) 10-match series Won 7 tosses	Would you believe it is a fair coin?	
2) 100-match series Won 70 tosses	Would you believe it is a fair coin?	
3) 1000-match series Won 700 tosses	Would you believe it is a fair coin?	

The important question we need to address is as follows:

What is the framework that can provide a quantifying metric to this intuition of ours?

# **Cricket series example**

1) What is our default assumption?

The coin is fair

2) When shall we reject this assumption?

We shall reject only when we have enough data that makes us conclude otherwise

# Judge in court

Assume that you are a judge in a court

A person is brought in front of you as a murder suspect

1) What is our default assumption?

The person is innocent

2) When shall we reject this assumption?

We shall reject only when we have enough data that makes us conclude that he is guilty

# **Machine Learning Deployment**

A machine learning model (legacy) is in production for a few years, and is doing fairly well You and your team have built a new model, and want to claim that it is better

1) What is the default assumption of the product owner?

The new model is not better than the legacy model

2) When shall we reject this assumption?

When enough data is given that the new model outperforms the legacy model significantly

# Third umpire

Suppose you are the third umpire

The on-field umpire has called for your help, and given a soft signal

1) What is our default assumption?

The on-field umpire is correct

2) When shall we reject this assumption?

We shall reject only when we have enough data that makes us conclude that the on-field umpire's decision can be changed

# Fingerprint scanner

We unlock our phones using fingerprint scanner A finger is now placed on the scanner

1) What should the default assumption be?

The fingerprint does not belong

2) When should the default assumption be rejected?

The default assumption should be rejected only when the data (fingerprint) is very conclusive that it belongs to the owner



# Radar example

A radar has to detect a plane

1) What should the default assumption be?

There is no plane

2) When should the default assumption be rejected?

The default assumption should be rejected only when the data is very conclusive that there is a plane





The coin is fair

The new model is not better than the legacy model

The person is innocent

The on-field umpire is correct

The fingerprint does not belong

There is no plane

All these are examples of setting up the Null Hypothesis

# **Terminologies**

 $H_0$  Null Hypothesis

#### Judge in court

# $H_0$ The person is innocent

We shall reject only when we have enough data that makes us conclude that he is guilty

#### Data:

The person has a knife in his pocket Innocent people can carry knife

The knife has blood stains

Maybe he is a cook/chef

Blood matches that of the victim Ok, this is too much

His shirt has fingerprints of the victim Highly unlikely that an innocent man has all these data points



Verdict: Guilty! (Reject the null hypothesis)



Probability of seeing data as extreme as what was observed, under the assumption that he is innocent, is very low

P data H0 is true is very low This is called p-value

If p-value is very low, we reject H0

# **Terminologies**

 $H_0$  Null Hypothesis

p-value

# Deep dive: coin toss

Put a quantitative metric on our suspicion that coin is biased

#### **Deep dive: coin toss**

Put a quantitative metric on our suspicion that coin is biased

 $H_0$ : coin is fair.

**Probability of heads = 0.5** 

1) 10-match series

7 Heads

Would you believe it is a fair coin?

Let T = number of heads

Test Statistic

Is T a random variable?

Yes

What is its distribution?

Binomial

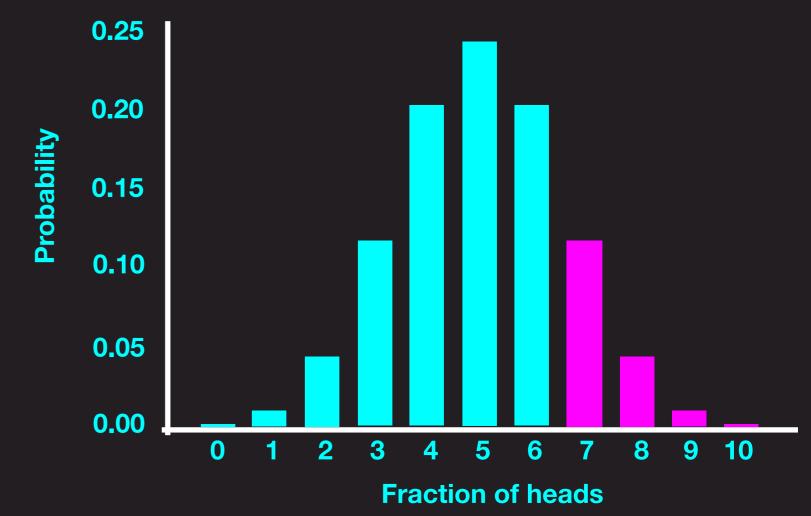
What is the observed value of T?

 $T_{\rm obs} = 7$ 

What is p-value?

p-value is the probability of observing data as extreme the observed test statistic under the assumption that the null hypothesis is true

The p- value is therefore given by  $P\left[T \geq 7 \middle| H_0 \text{ is true}\right]$ 



#### How do we compute P-value here?

binom.pmf(k=7, n=10, p=0.5) + binom.pmf(k=8, n=10, p=0.5) + binom.pmf(k=9, n=10, p=0.5) + binom.pmf(k=10, n=10, p=0.5)

1 - binom.cdf(k=6, n=10, p=0.5) = 0.172

**Deep dive: coin toss** 

Put a quantitative metric on our suspicion that coin is biased

 $H_0$ : coin is fair. Probability of heads = 0.5

**Test Statistic** Let T =number of heads

1) 10-match series 7 Heads

$$P\left[T \ge 7 \middle| H_0 \text{ is true}\right] = 1 - \text{binom.cdf(}k=6, n=10, p=0.5) = 0.172$$

2) 100-match series

70 Heads

$$P\left[T \ge 70 \middle| H_0 \text{ is true}\right] = 1 - \text{binom.cdf(}k=69, n=100, p=0.5) = 0.000039$$

3) 1000-match series 700 Heads

$$P\left[T \ge 700 \,\middle|\, H_0 \text{ is true}\right] = 1 - \text{binom.cdf(}k=699, n=1000, p=0.5) = 0$$

When do we reject the null hypothesis H0?

When the p-value is very low

Typically used threshold is 0.05 (This can change based on business needs)

This threshold is denoted by  $\alpha$  and is called Significance Level

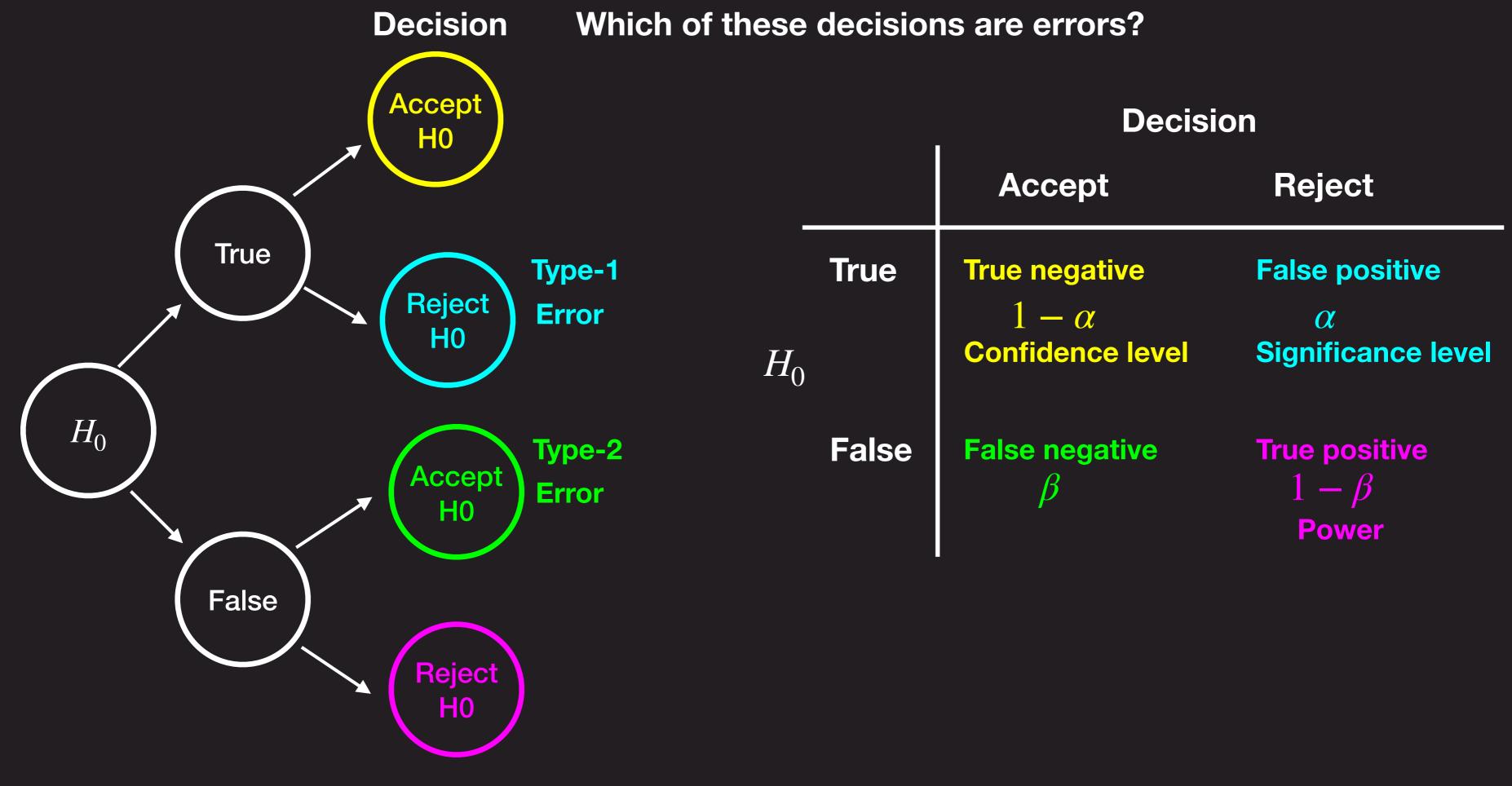
# **Terminologies**

 $H_0$  Null Hypothesis

p-value

**Test Statistic** 

Significance Level



Note: Statisticians do not say "Accept". They say "fail to reject"

#### **Null Vs Alternate**

H0 indicated "null" hypothesis

In the event of rejection of H0, we need the right "alternate hypothesis"

# **Court example**

 $H_0$  "person is innocent"

 $H_a$  "person is guilty"

Coin toss with 70% heads

 $H_0$  "coin is fair"

 $H_a$  "coin is biased towards heads"

# **ML** deployment

 $H_0$  "The new model is not better than the legacy model"

 $H_a$  "The new model is better than the legacy model"

# **Burger company**

A company selling burgers claims that its burger weighs 200 grams.

An unsatisfied customer, still feeling hungry after eating, wants to disprove the claim

### Which is the right setup?

 $H_0$ : weight = 200

 $H_a$ : weight  $\neq$  200



 $H_0$ : weight = 200

 $H_a$ : weight < 200

 $H_0$ : weight >= 200

 $H_a$ : weight < 200

 $H_0$ : weight = 200

 $H_a$ : weight > 200

 $H_0$ : weight <= 200

 $H_a$ : weight > 200

# **AI Chip startup**

An Al chip startup claims that it beats the GPU in computer vision tasks The training time for ResNet is 15 minutes on the GPU What is the right way for the startup to claim that it is better?

Which is the right setup?

 $H_0$ : training time = 15

 $H_a$ : training time < 15

 $H_0$ : training time >= 15

 $H_a$ : training time < 15

training time = 15

 $H_a$ : training time  $\neq$  15

 $H_0$ : training time = 15

 $H_a$ : training time > 15

 $H_0$ : training time <= 15

 $H_a$ : training time > 15

# **Google Pay**

You want to convince your parents that paying by Google Pay is faster than paying by cash and waiting for change

Your parents counter by saying that internet may be slow and Google Pay is slower

Let  $\mu$  be the proportion of times Google Pay was faster than cash

Which is the right setup?

$$H_0$$
:  $\mu = 0.5$ 

$$H_a$$
:  $\mu$  < 0.5

$$H_0$$
:  $\mu >= 0.5$ 

$$H_a$$
:  $\mu$  < 0.5

$$H_0$$
:  $\mu = 0.5$ 

$$H_a$$
:  $\mu \neq 0.5$ 

$$H_0$$
:  $\mu = 0.5$   $H_a$ :  $\mu > 0.5$ 

$$H_0$$
:  $\mu <= 0.5$ 

$$H_a$$
:  $\mu > 0.5$ 

# Height from your state

The average height of Indians is 65 inches

You want to verify whether this is true for people from your state

Let  $\mu$  be the average height of people from your state

Which is the right setup?

$$H_0$$
:  $\mu = 65$ 

$$H_a$$
:  $\mu$  < 65

$$H_0$$
:  $\mu >= 65$ 

$$H_a$$
:  $\mu$  < 65



$$H_0$$
:  $\mu$  = 65

$$H_a$$
:  $\mu \neq 65$ 

$$H_0$$
:  $\mu = 65$ 

$$H_a$$
:  $\mu > 65$ 

$$H_0$$
:  $\mu <= 65$ 

$$H_a$$
:  $\mu > 6$ !

# Left Vs right tailed

# **Burger company**

 $H_0$ : weight = 200

 $H_a$ : weight < 200

**Left-tailed** 

# **Al Chip startup**

 $H_0$ : training time = 15

**Left-tailed** 

 $H_a$ : training time < 15

# **Google Pay**

 $H_0$ :  $\mu = 0.5$ 

 $H_a$ :  $\mu > 0.5$ 

**Right-tailed** 

# Height from your state

 $H_0$ :  $\mu = 65$ 

**Two-tailed** 

 $H_a$ :  $\mu \neq 65$ 

- 1) Setup the Null and Alternate Hypothesis
- 2) Choose the right test statistic
- 3) Left-tailed Vs Right-tailed Vs Two-tailed
- 4) Compute p-value
- 5) If p-value is less than alpha, then reject the null hypothesis