

## QUANTITATIVE ANALYSIS

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# DIFFERENCE OF MEANS (1)

# AGENDA

1. Follow-up
2. Revisiting Distributions
3. One Sample
4. Independent Samples
5. Dependent Samples
6. Effect Sizes

# 1 FOLLOW-UP

# 2 REVISITING DISTRIBUTIONS

# VARIANCE

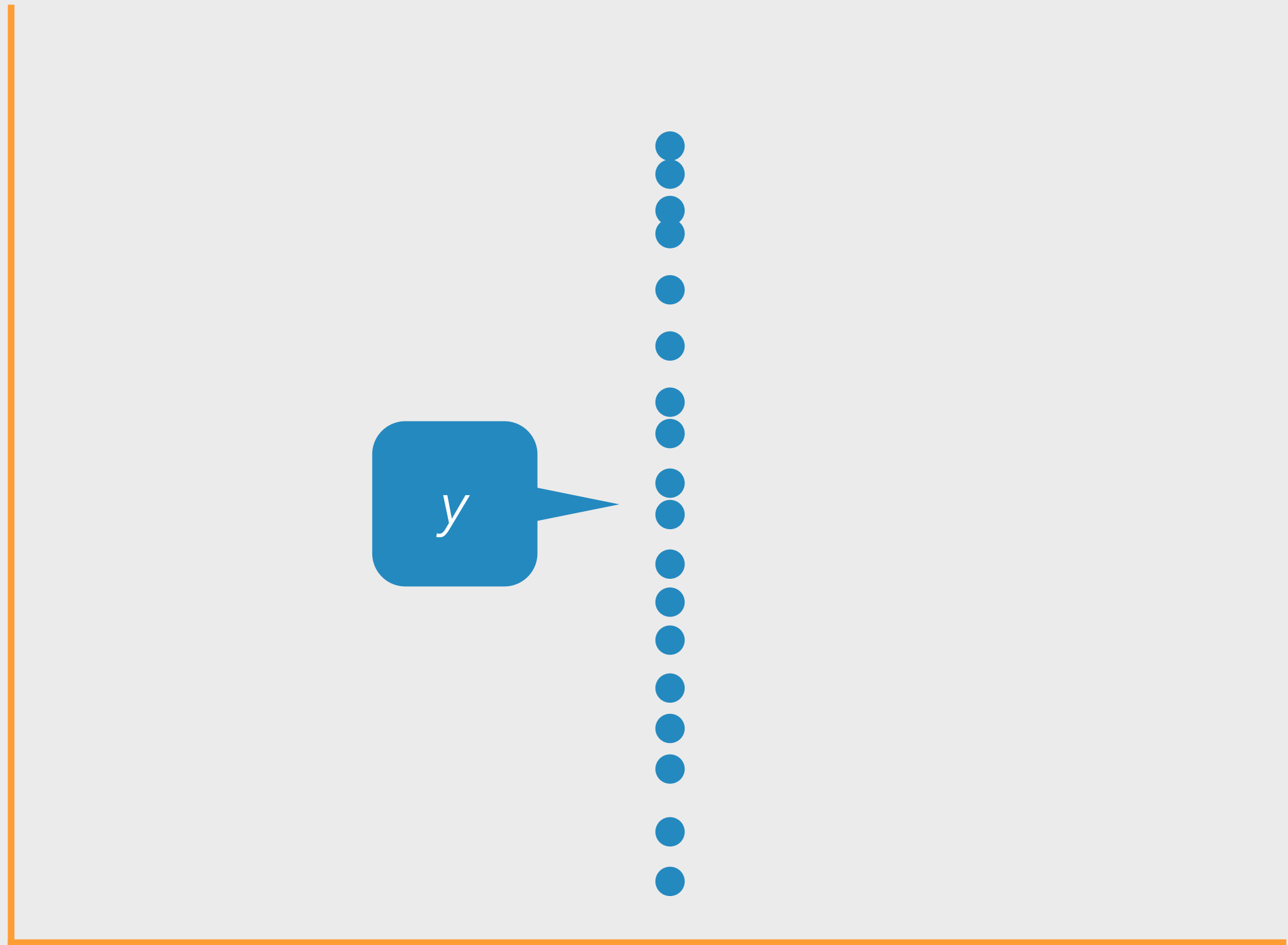
SECOND  
MOMENT

$$s^2 = \frac{\sum_{i=1}^n (x - \bar{x})^2}{n - 1}$$

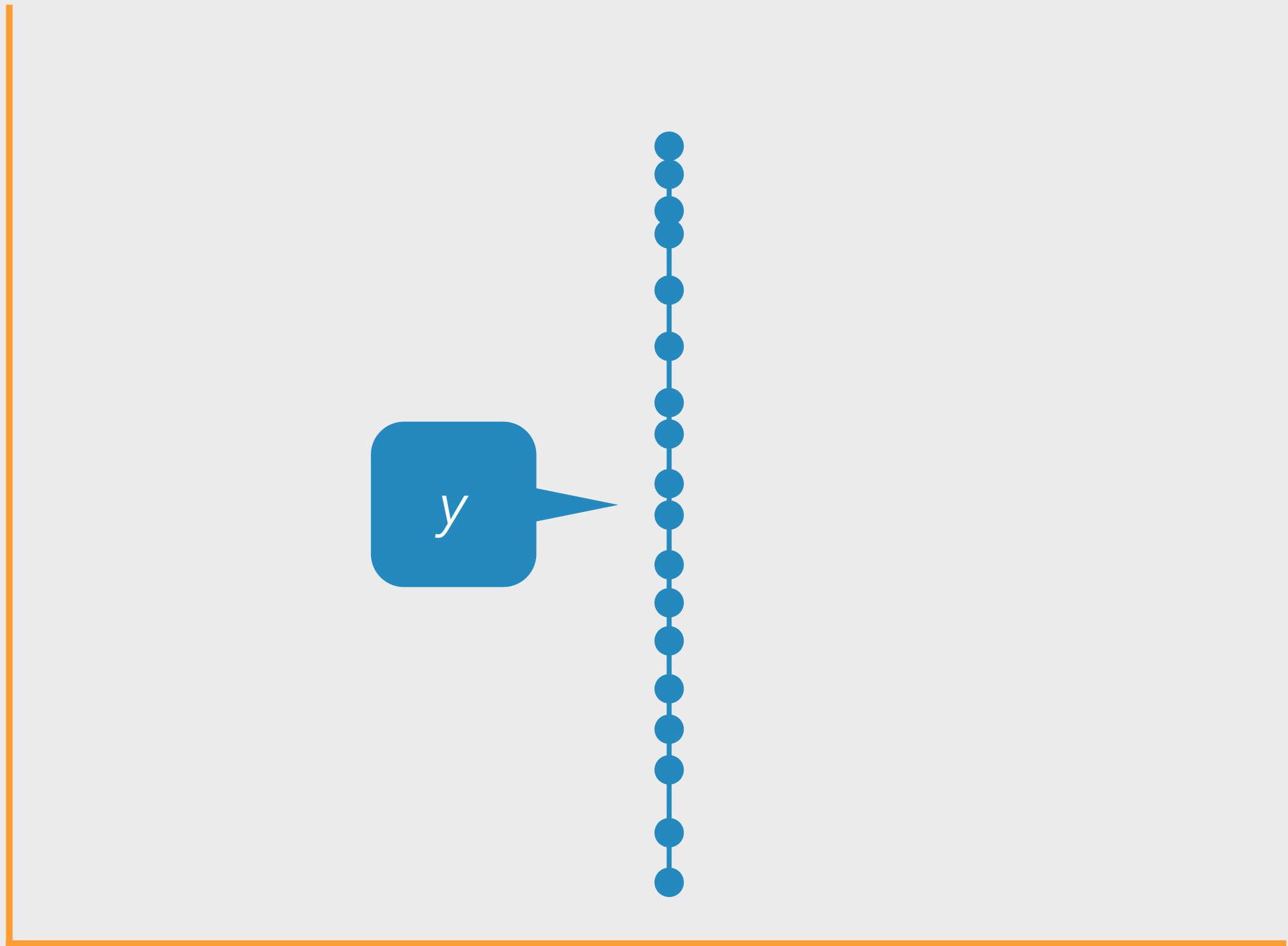
### DEFINITION

SUM OF ALL DEVIANCES,  
SQUARED AND DIVIDED BY  
ONE DEGREE OF FREEDOM;  
EXPECTATION OF HOW  
DISTRIBUTION DEVIATES  
FROM THE MEAN

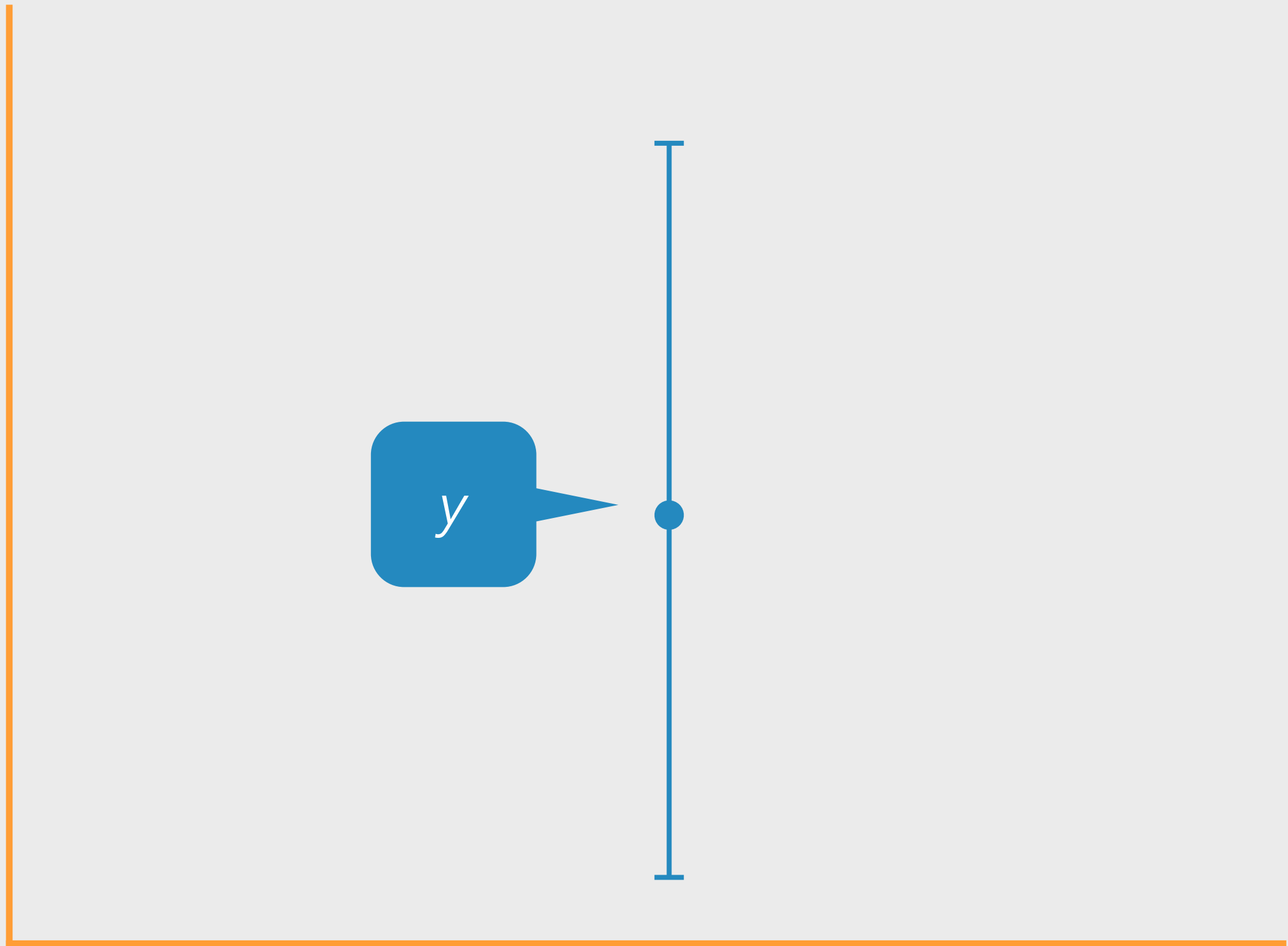
# VARIANCE



# VARIANCE

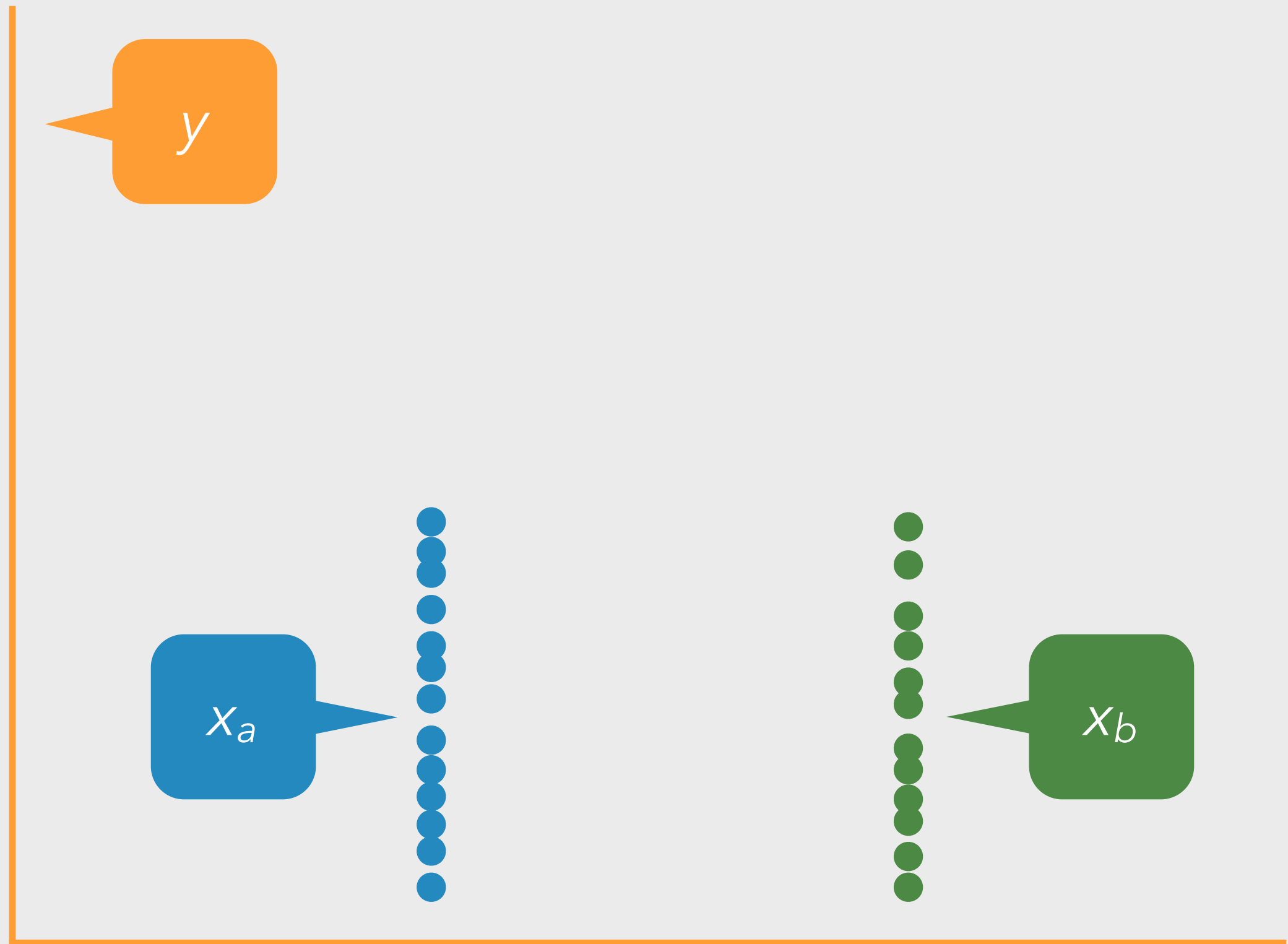


# VARIANCE

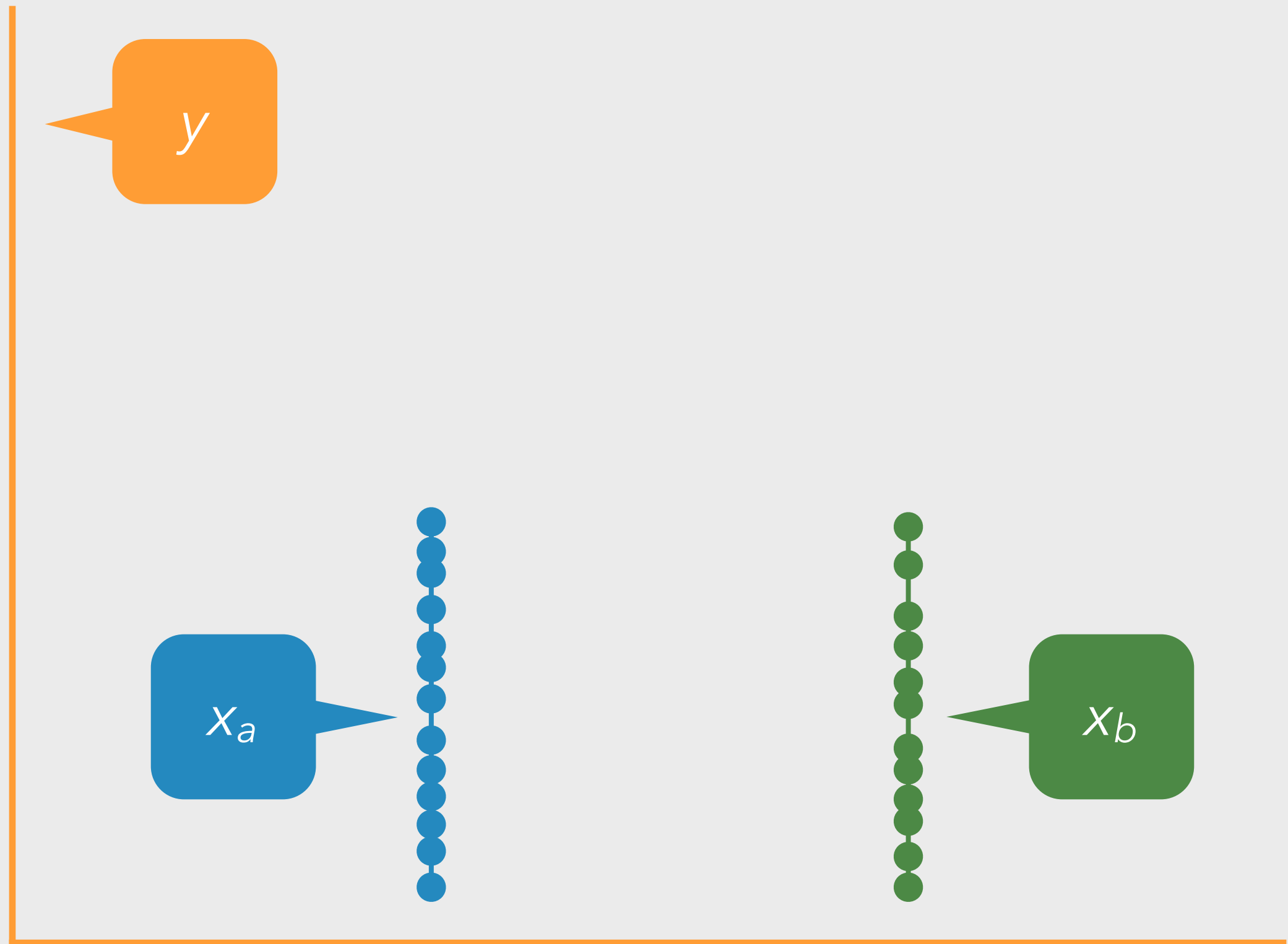




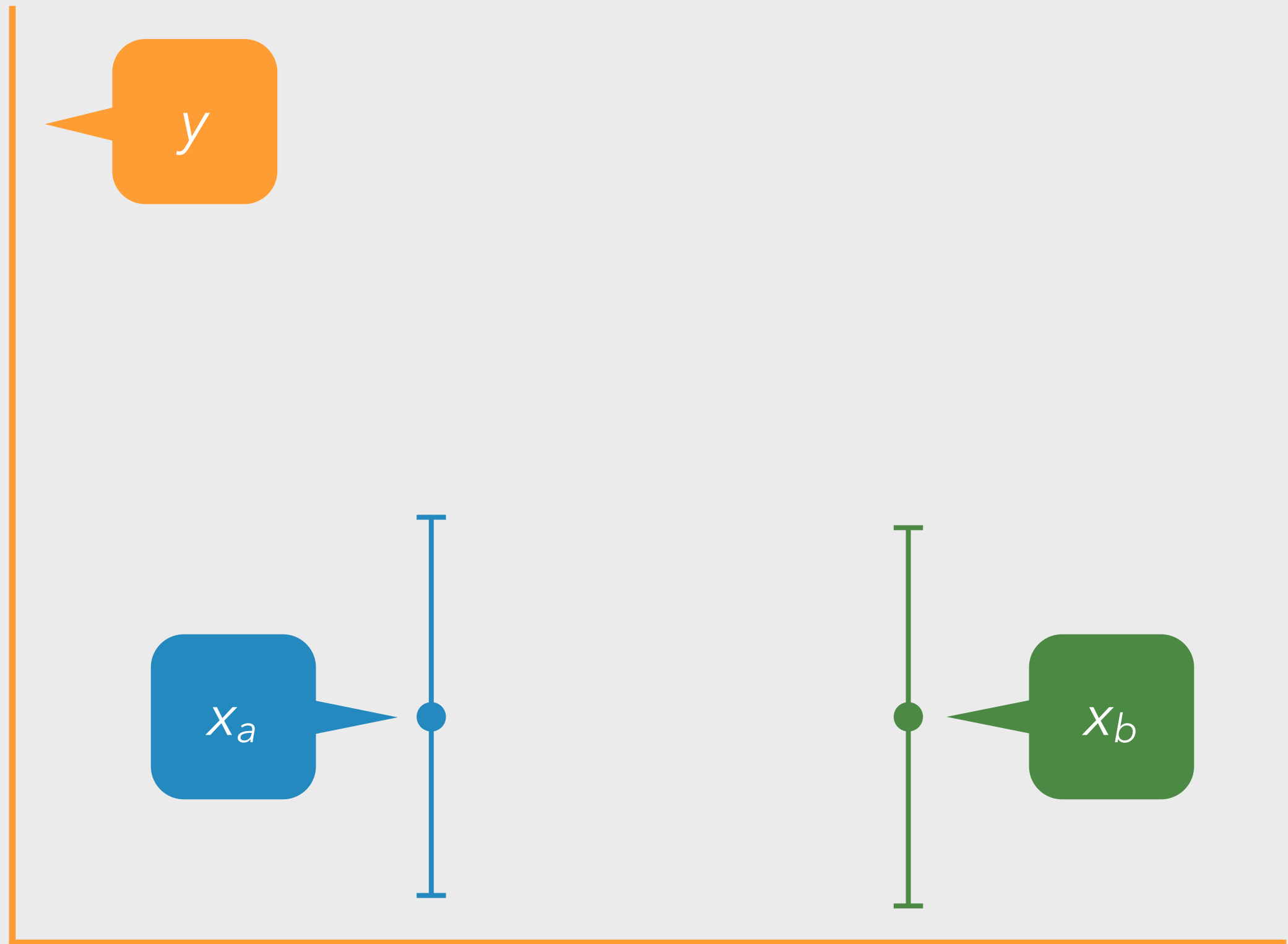
# VARIANCE



# VARIANCE

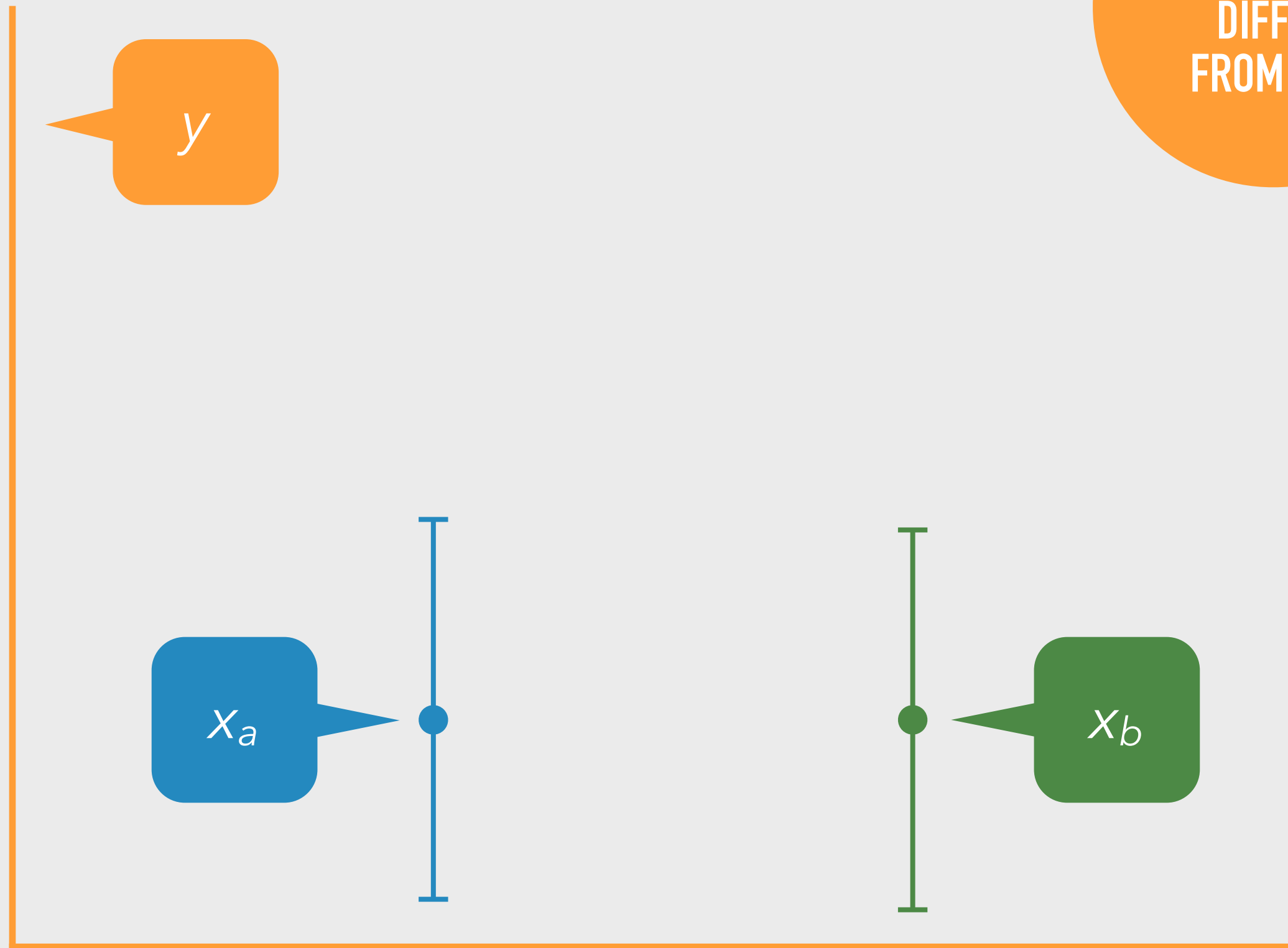


# VARIANCE

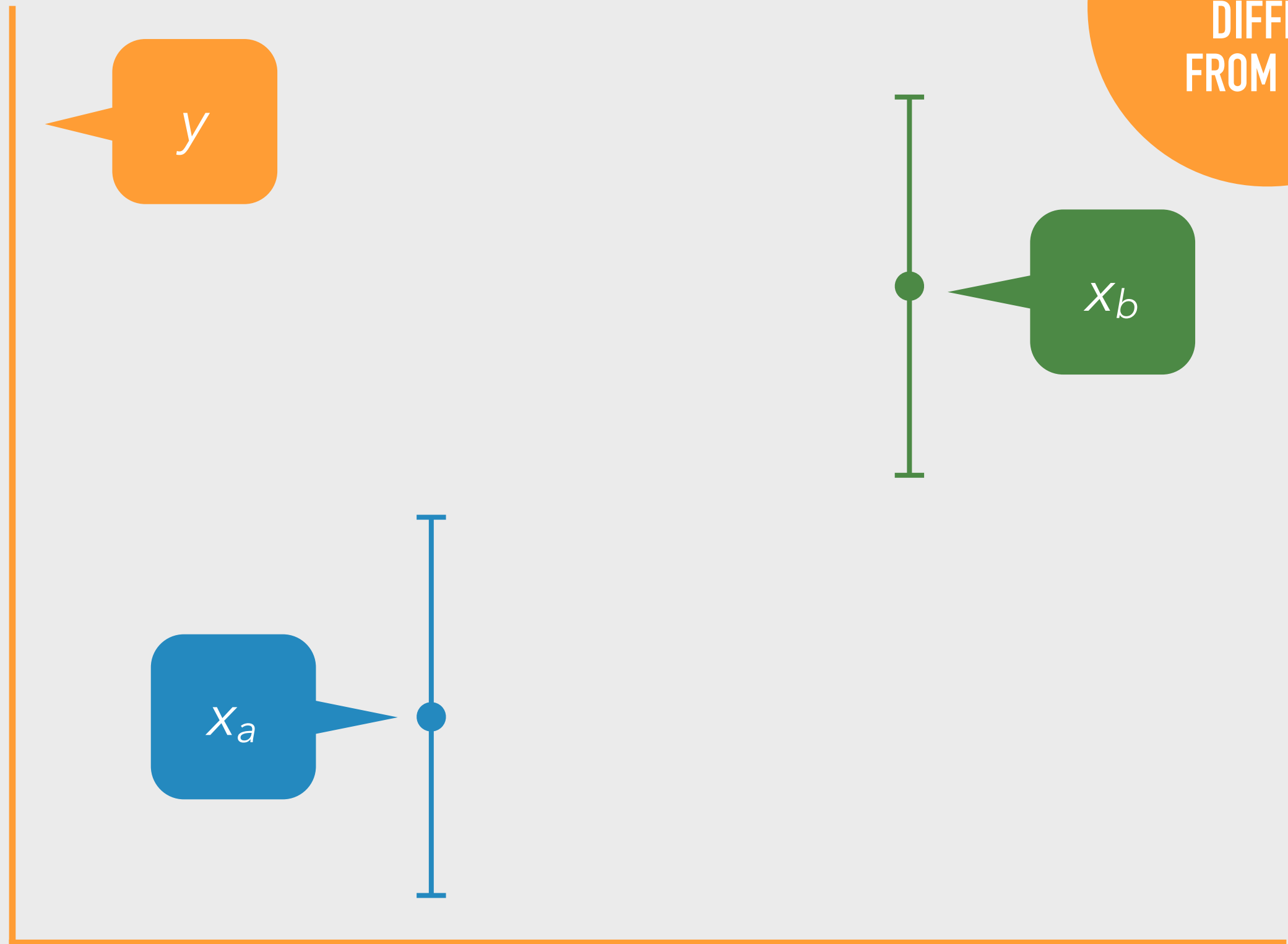


# VARIANCE

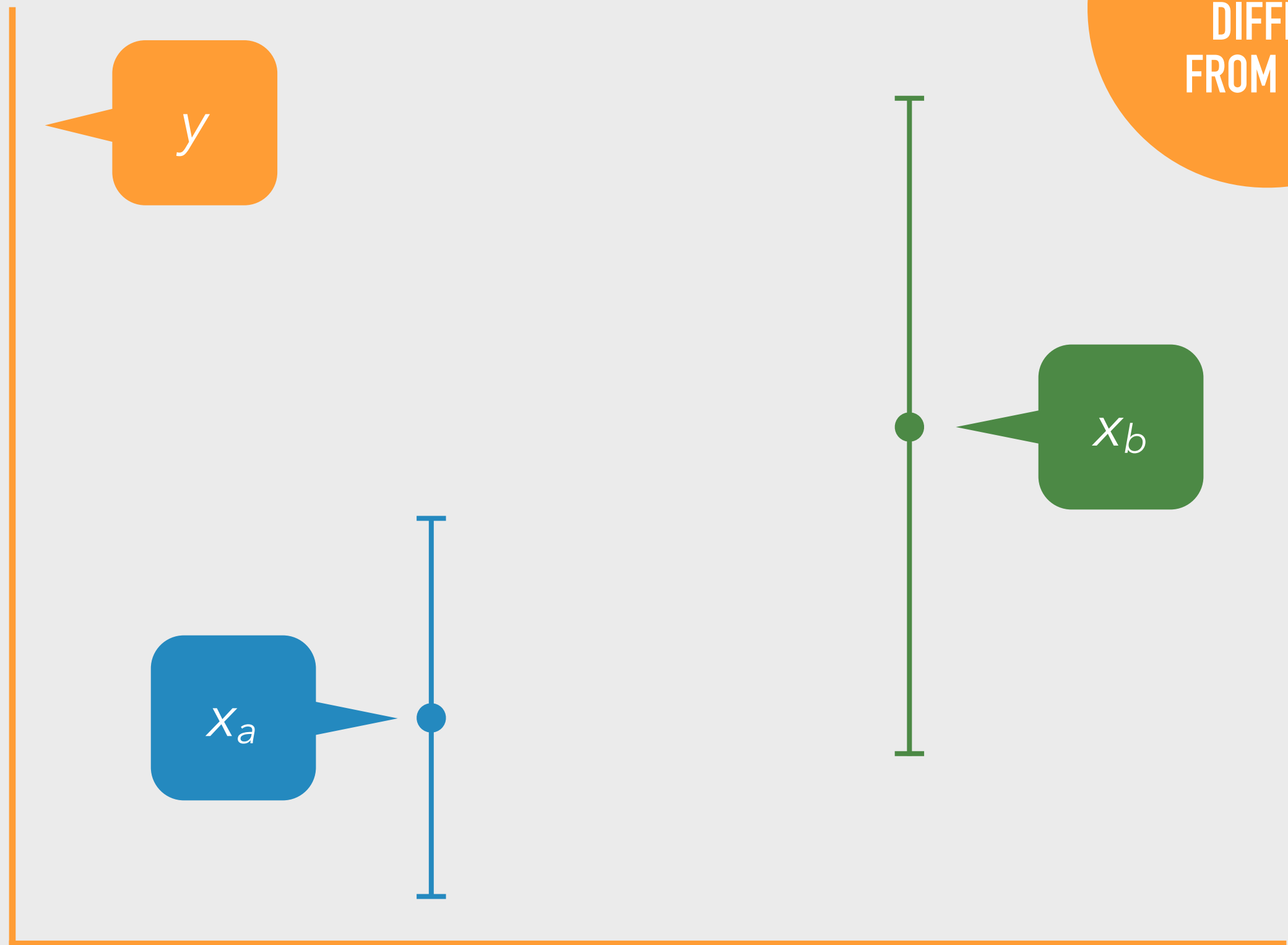
DOES THE  
VARIANCE OF  $X_A$   
DIFFER  
FROM  $X_B$ ?



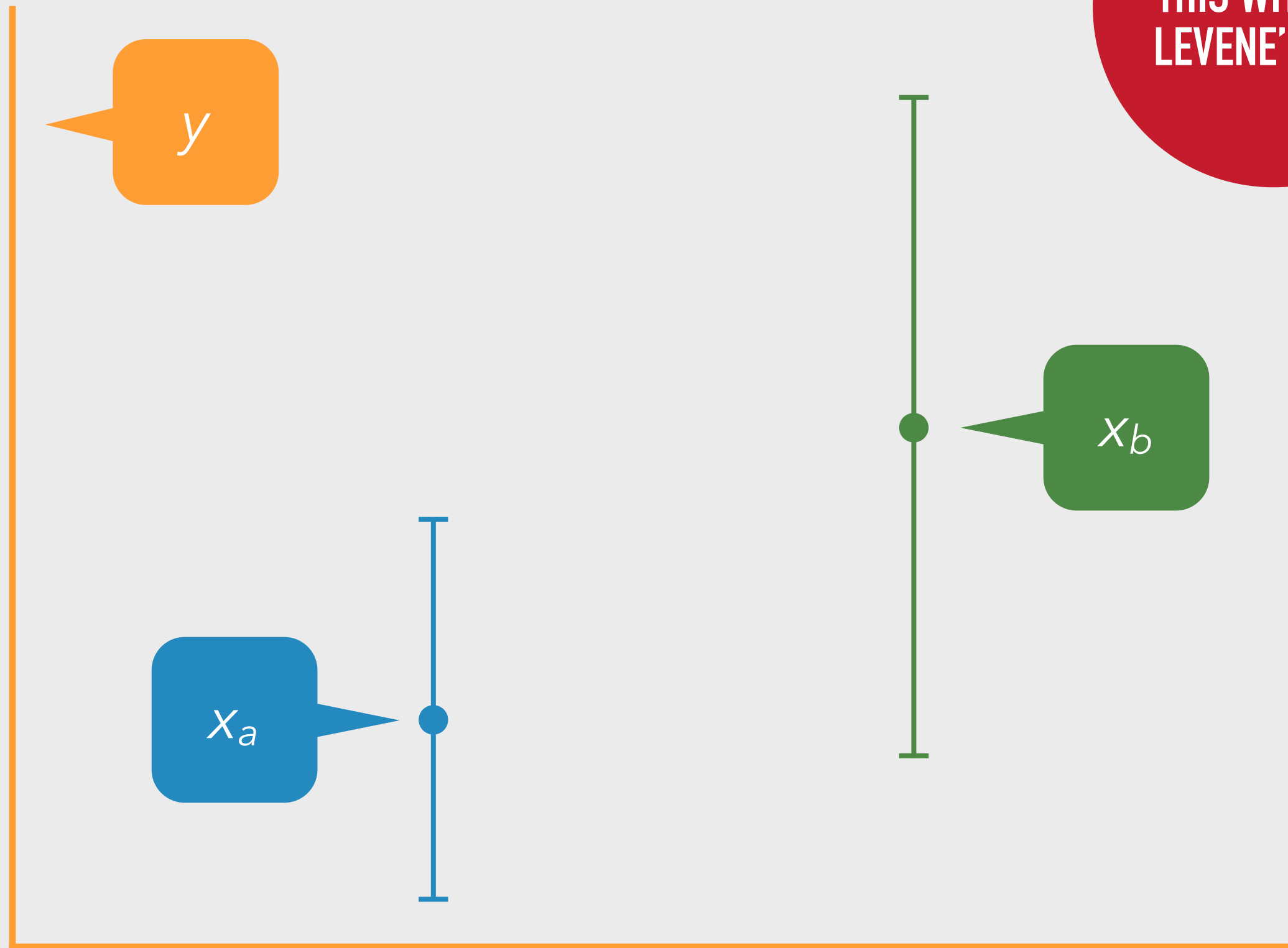
# VARIANCE



# VARIANCE



# VARIANCE



WE CAN TEST  
THIS WITH THE  
LEVENE'S TEST

3 ONE SAMPLE



# STUDENT'S T-TEST

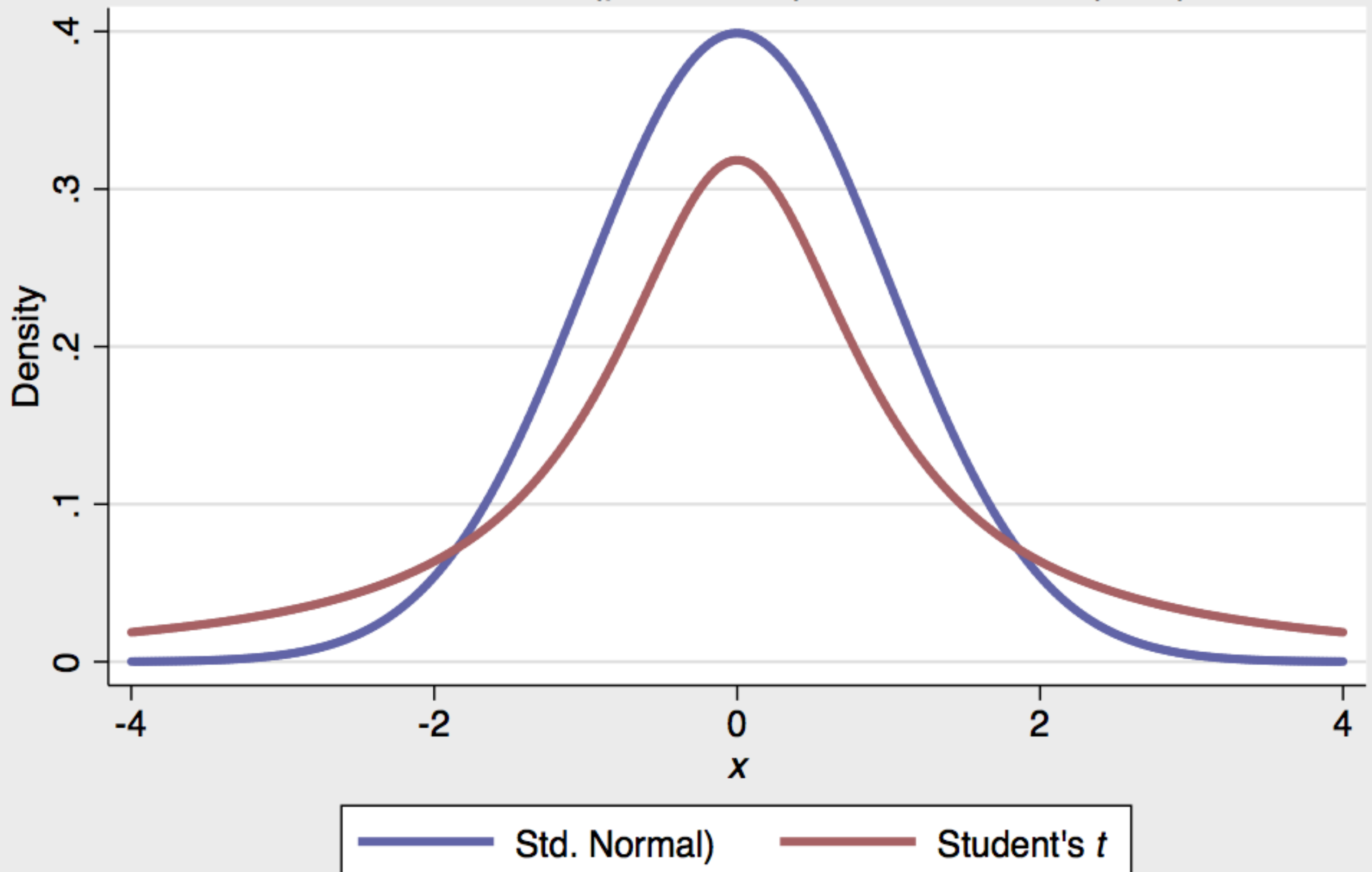
- ▶ Employee of the Guinness company who published his work under the pseudonym "Student".
- ▶ Student of Karl Pearson's while on research leaves from Guinness.
- ▶ Original t-tests were developed to conducting quality control testing on Guinness stout.

**WILLIAM SEALY GOSSET (1876–1937)**  
**"STUDENT"**



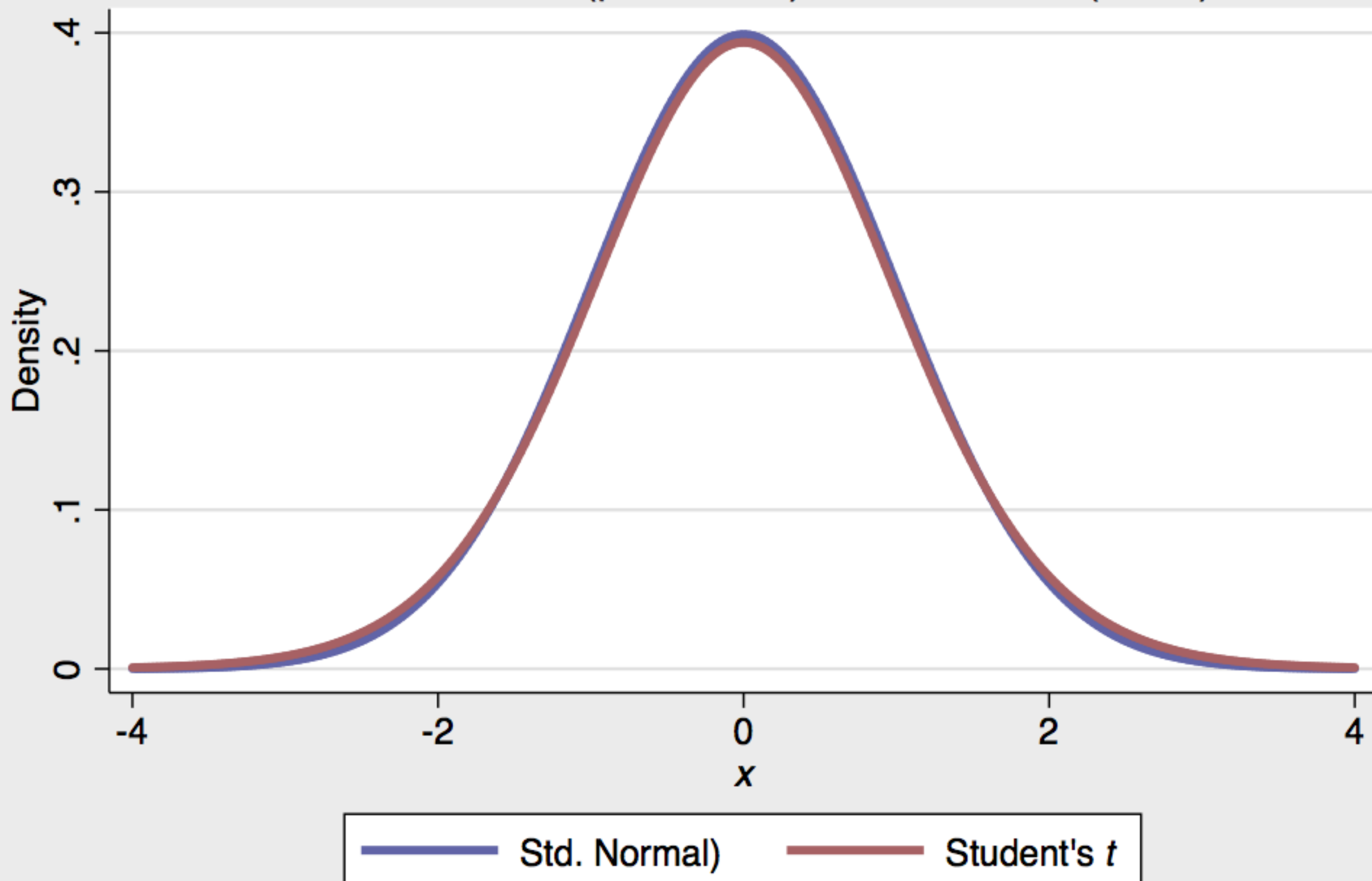
# Probability Density Functions Compared

Standard Normal ( $\mu=0$ ,  $\sigma=1.0$ ) and Student's  $t$  ( $df=1$ )



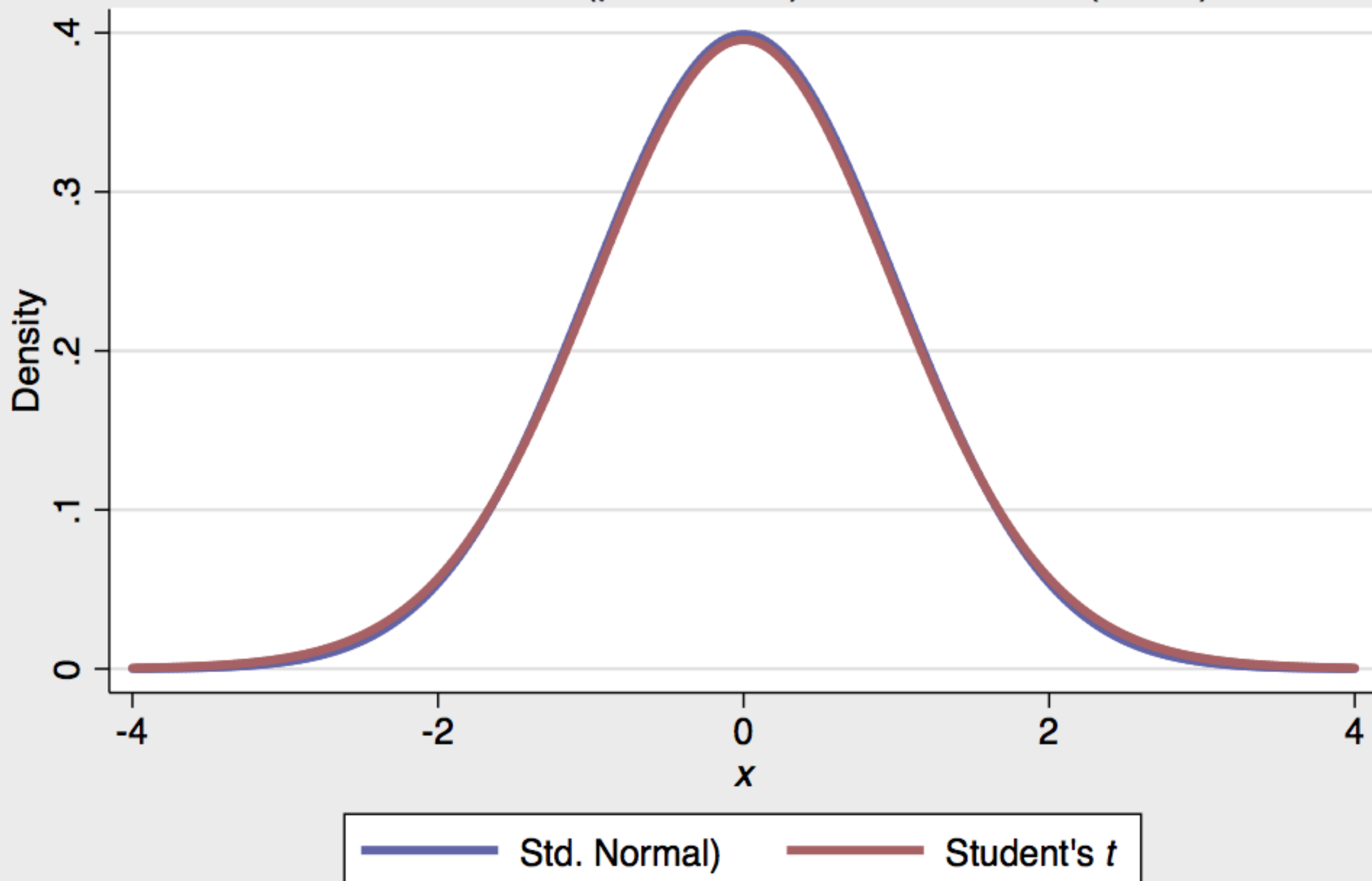
# Probability Density Functions Compared

Standard Normal ( $\mu=0$ ,  $\sigma=1.0$ ) and Student's  $t$  ( $df=20$ )



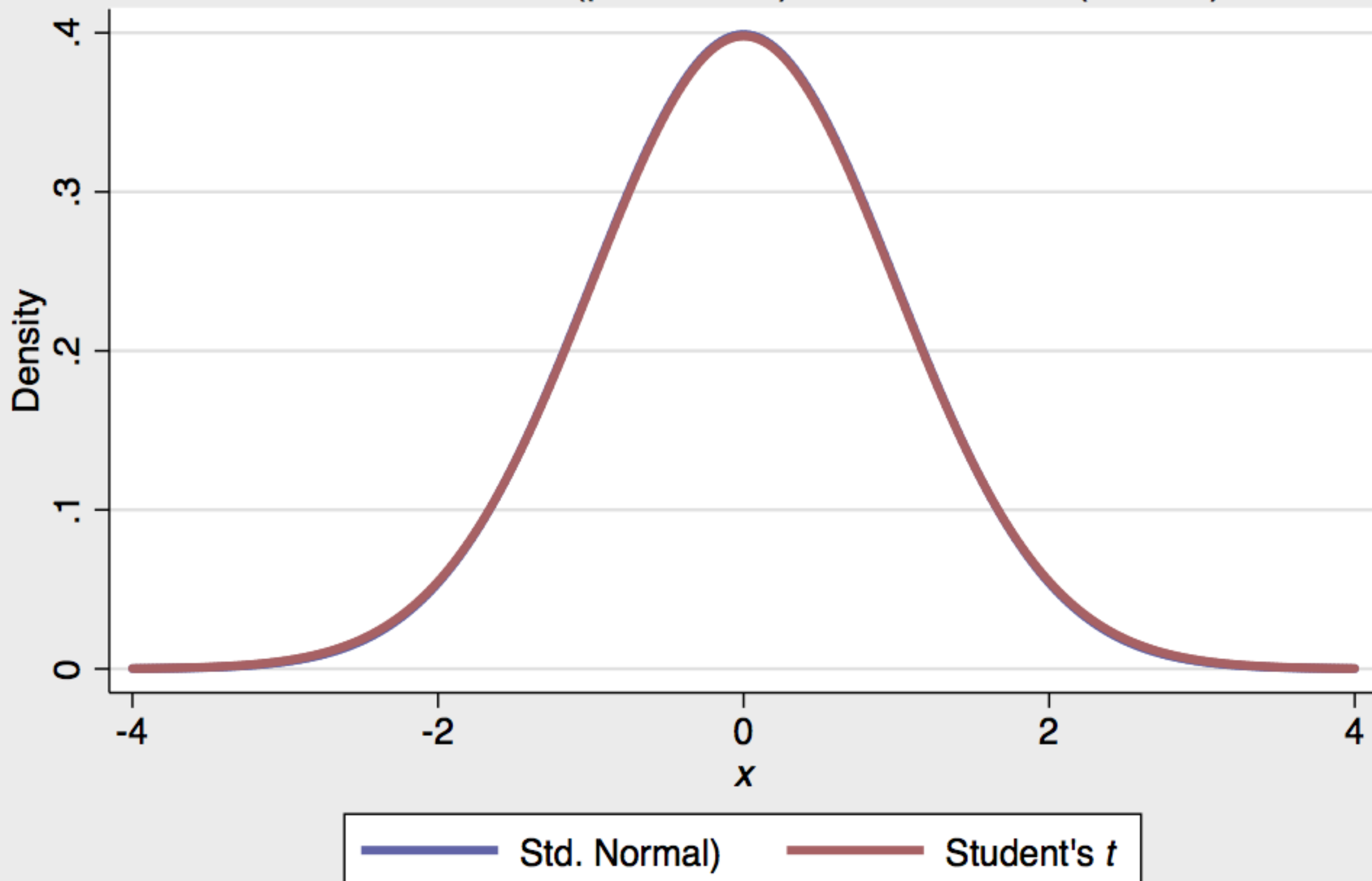
# Probability Density Functions Compared

Standard Normal ( $\mu=0$ ,  $\sigma=1.0$ ) and Student's  $t$  ( $df=30$ )

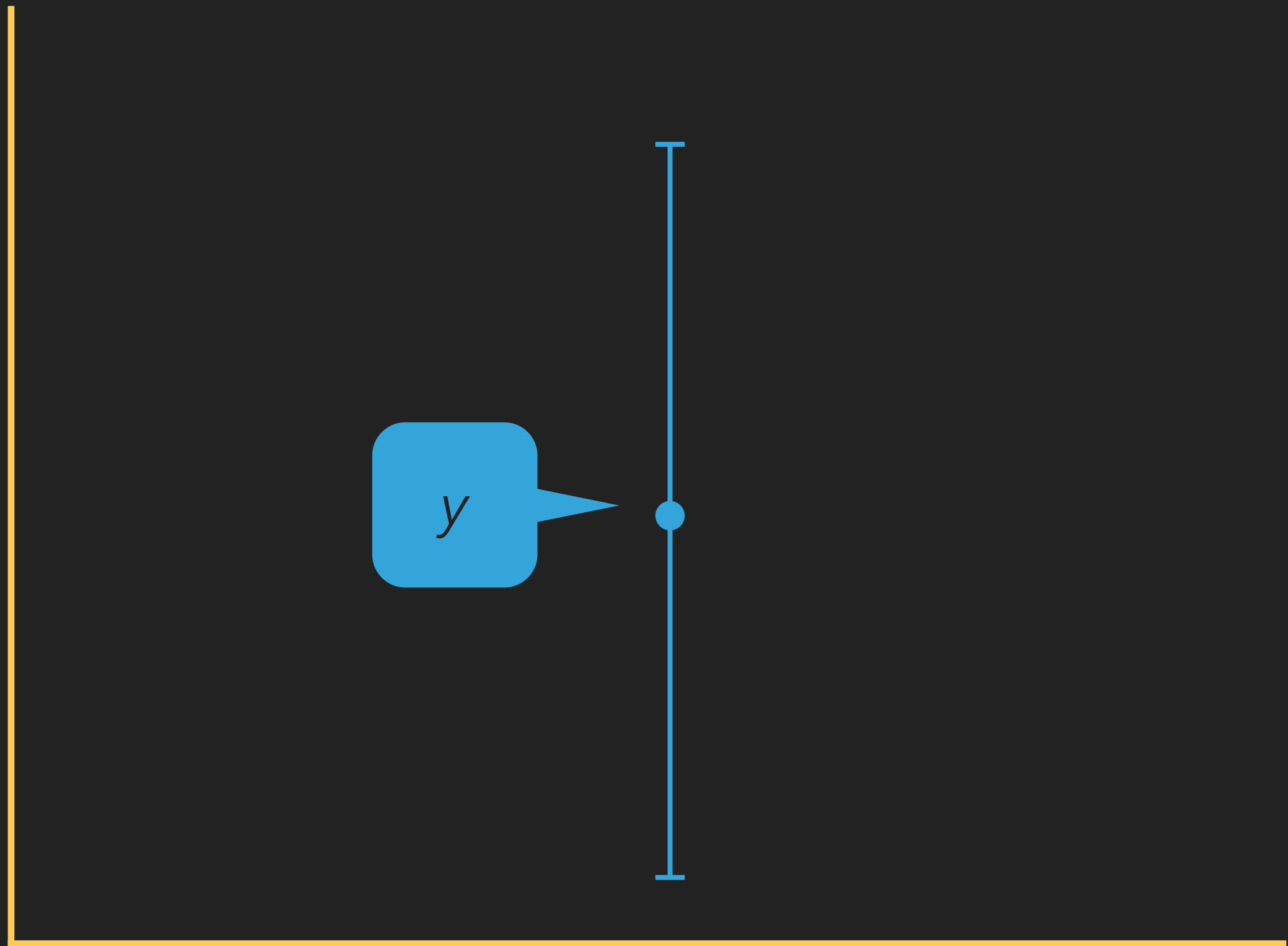


# Probability Density Functions Compared

Standard Normal ( $\mu=0$ ,  $\sigma=1.0$ ) and Student's  $t$  ( $df=100$ )

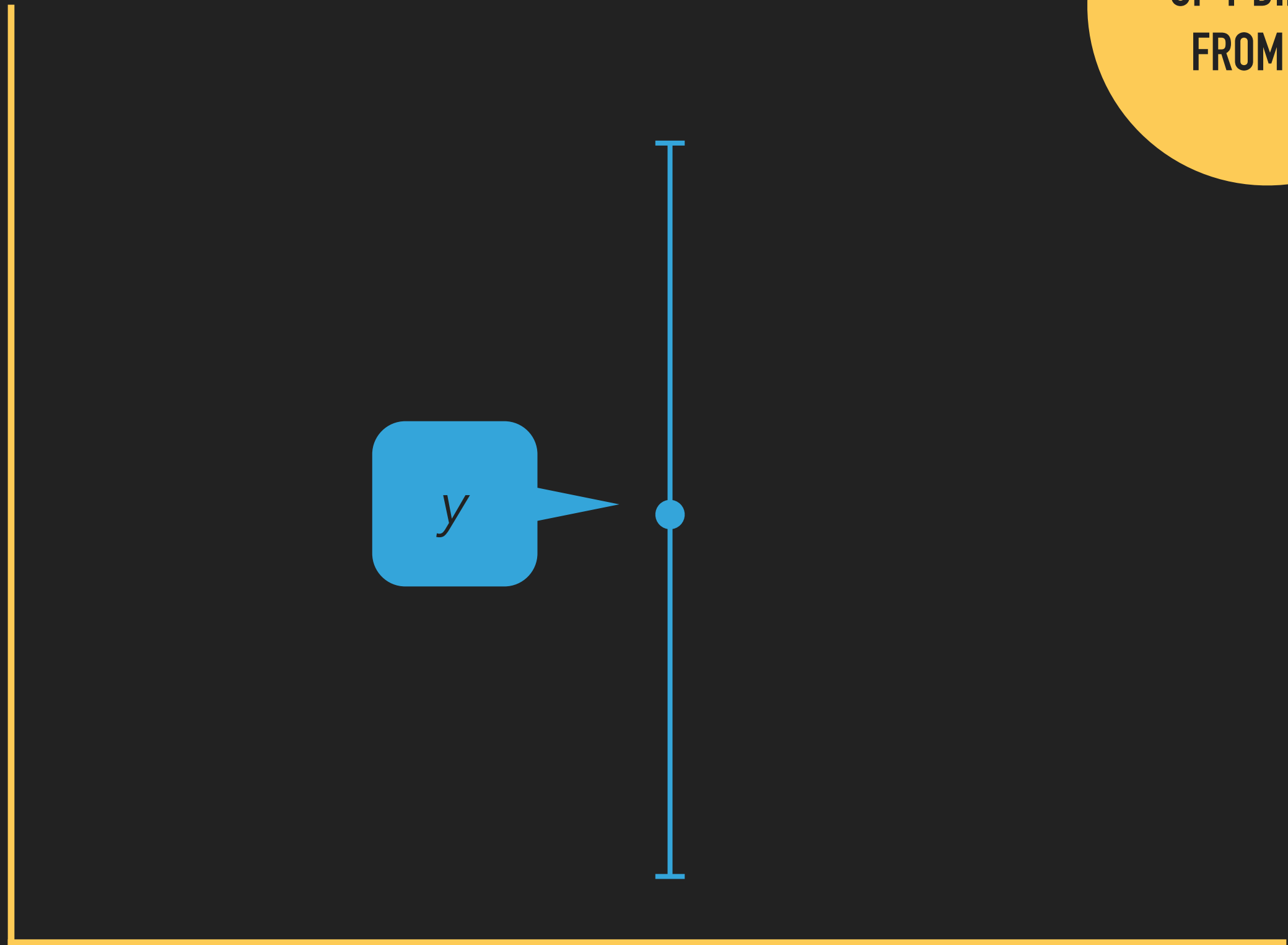


# DIFFERENCE IN MEANS



# DIFFERENCE IN MEANS

DOES THE MEAN  
OF Y DIFFER  
FROM  $\mu$ ?



# HYPOTHESES

- ▶  $H_0$  = there is no significant difference between the mean of  $y$  and the population
- ▶  $H_1$  = there is a significant difference between the mean of  $y$  and the population



# ASSUMPTIONS

- ▶ continuous data ( $y$ )
- ▶ the distribution of  $y$  is approximately normal
- ▶ degrees of freedom ( $\nu$ ) =  $n-1$

# FORMULA

$$t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}}$$

# FORMULA

$$t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}}$$

standard error

### 3. INDEPENDENT SAMPLES

---

# FIND THE PROBABILITY OF T

```
display ttail(df,t)*2
```

```
. display ttail(72,3.6308)*2
```

```
.0005255
```

```
. display ttail(72,1.6308)*2
```

```
.1072996
```

### 3. INDEPENDENT SAMPLES

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# FIND THE PROBABILITY OF T

```
display (1-ttail(df,-t))*2
```

```
▪ display (1-ttail(72,-3.6308))*2
```

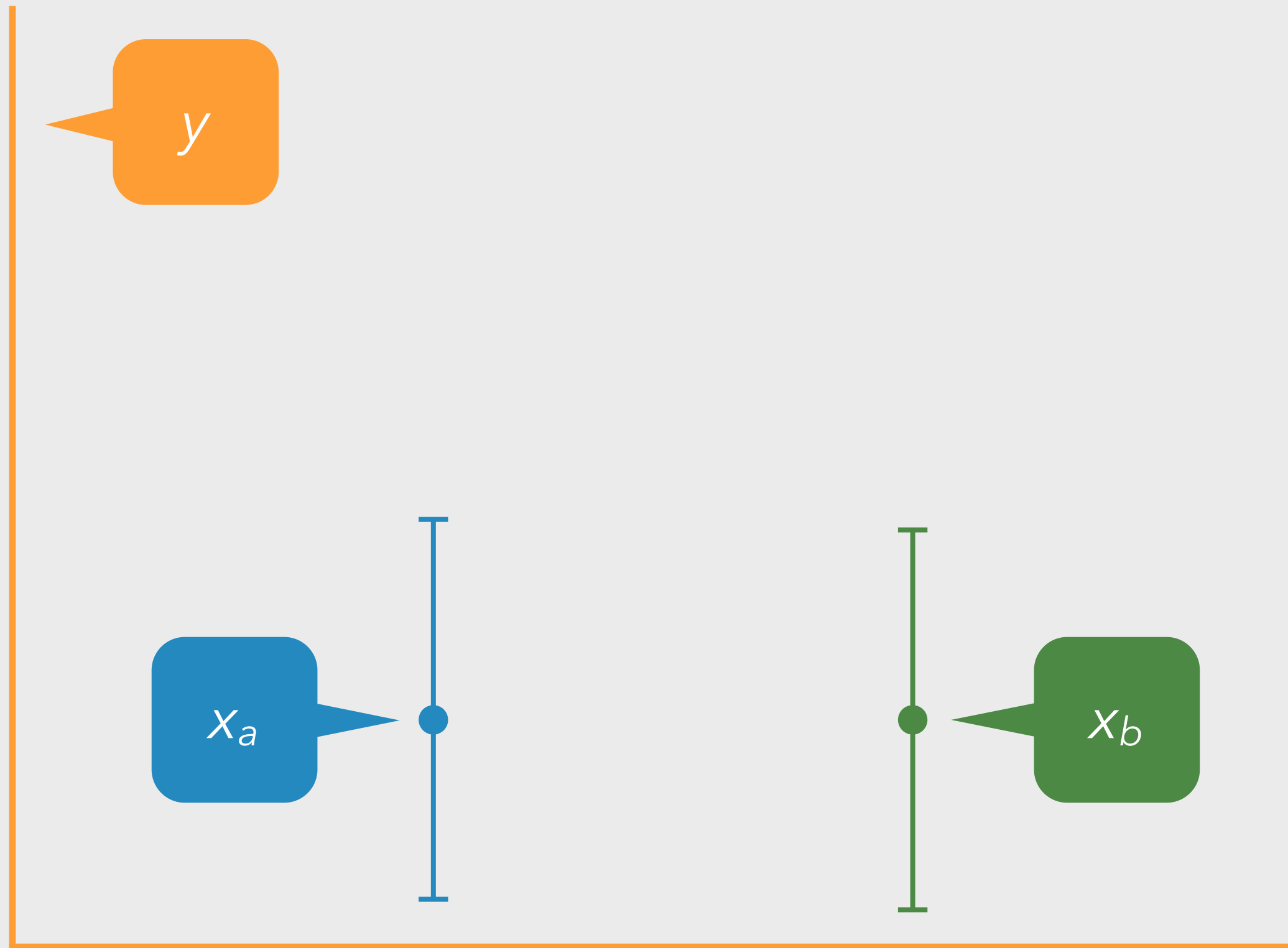
```
▪ .0005255
```

```
▪ display (1-ttail(72,-1.6308))*2
```

```
▪ .1072996
```

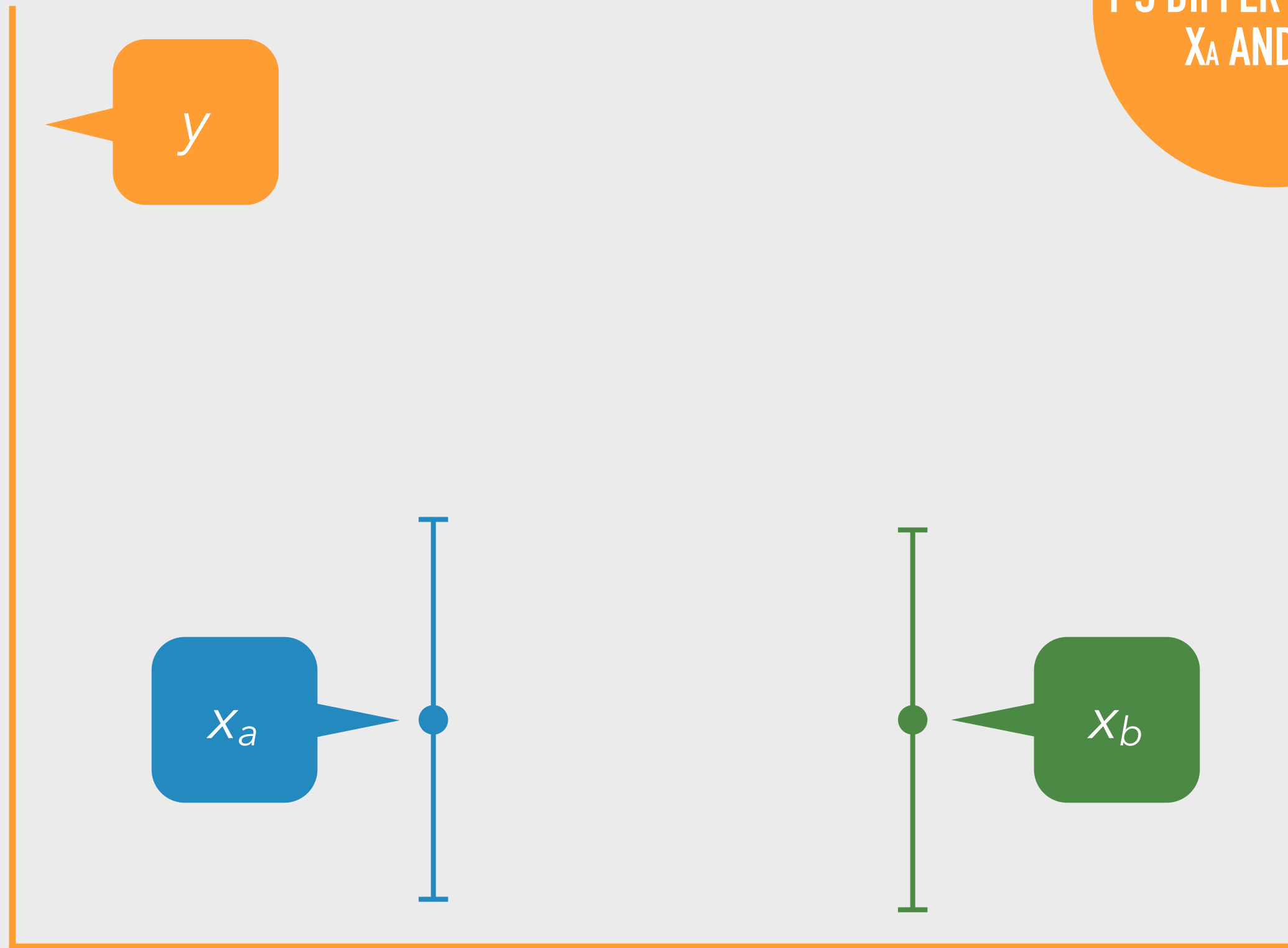
# 4 INDEPENDENT SAMPLES

# DIFFERENCE OF MEANS



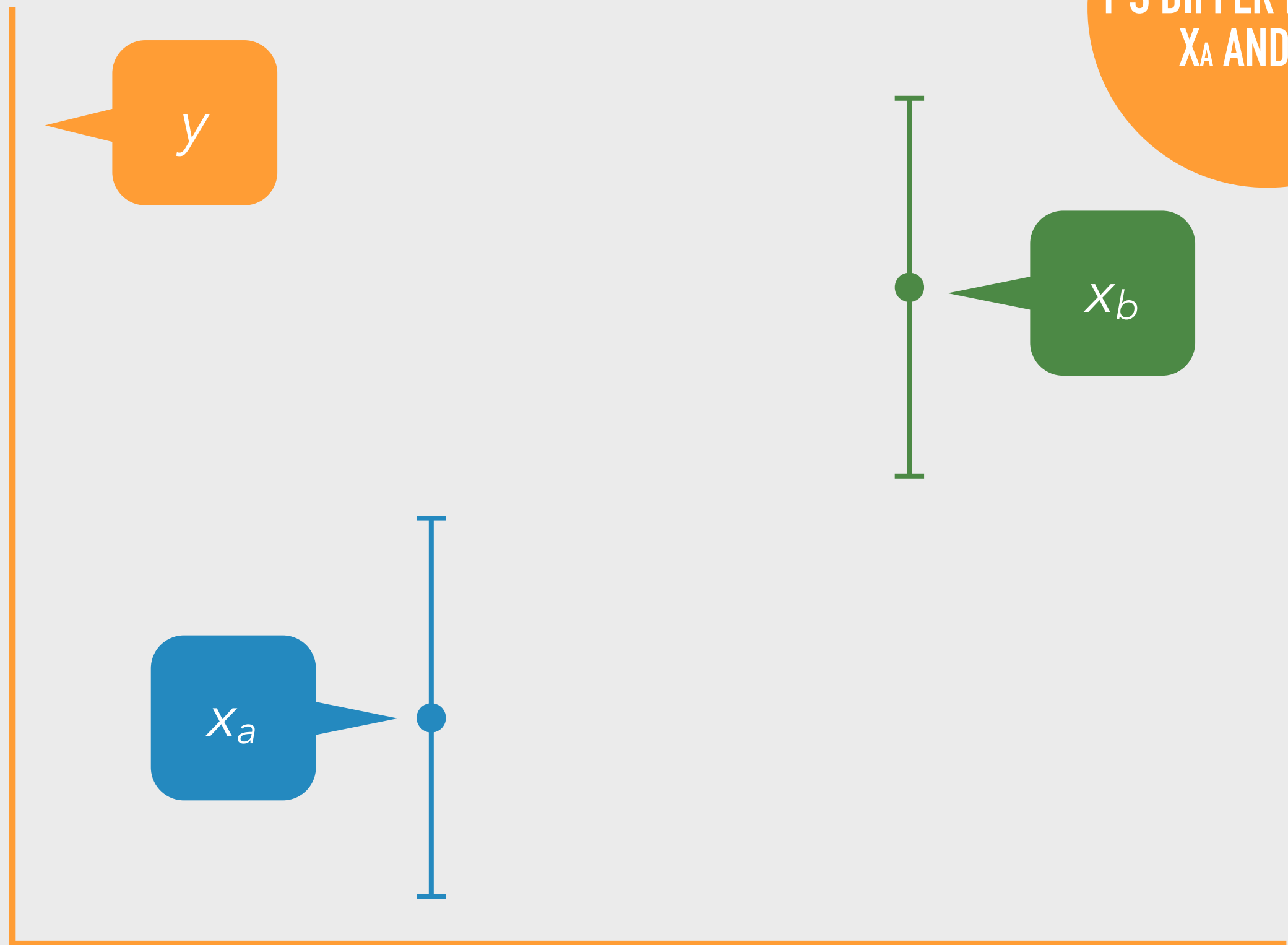
# DIFFERENCE OF MEANS

DO THE MEAN  
Y'S DIFFER BETWEEN  
 $X_A$  AND  $X_B$ ?



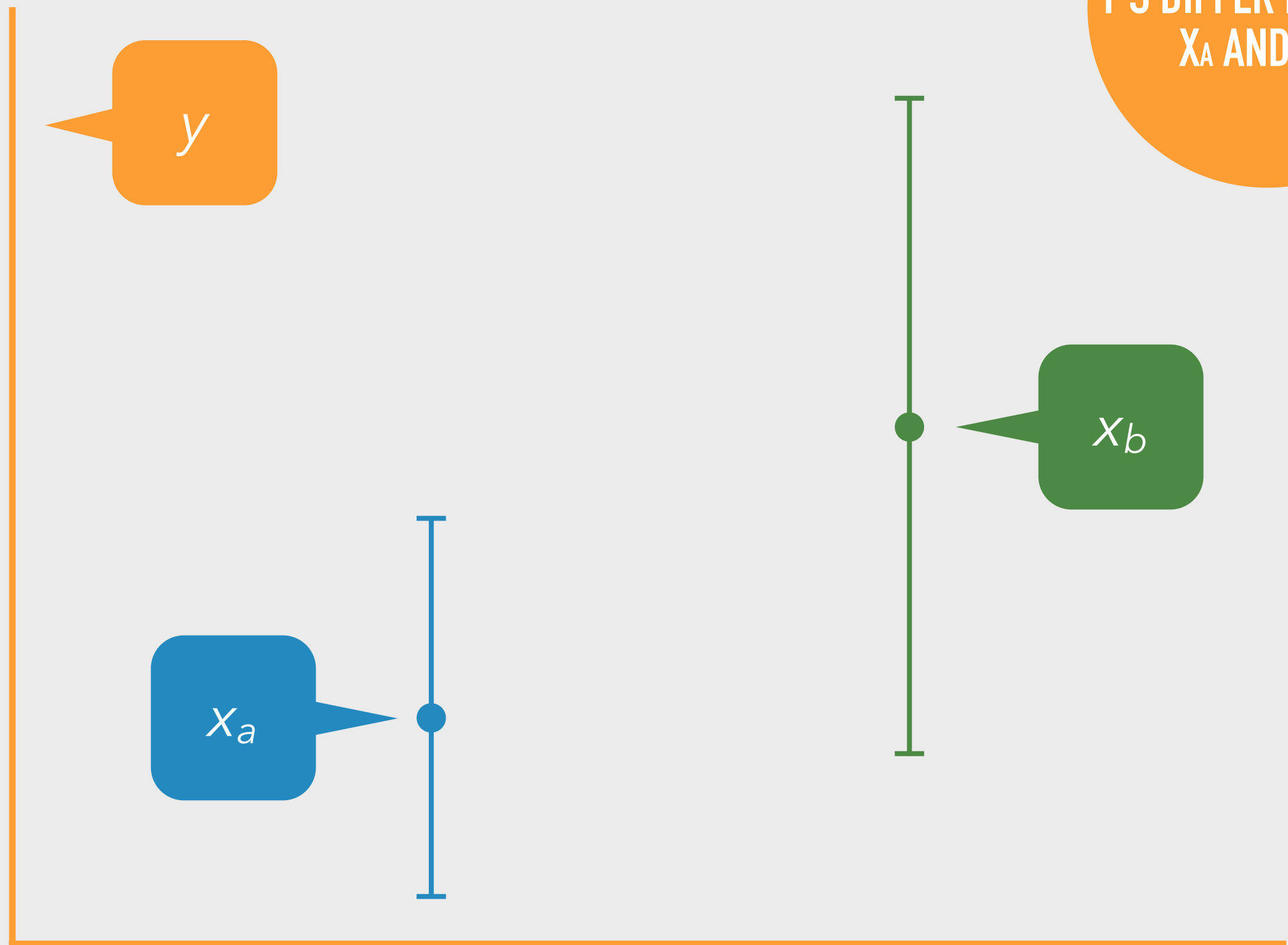


# DIFFERENCE OF MEANS



DO THE MEAN  
Y'S DIFFER BETWEEN  
 $X_A$  AND  $X_B$ ?

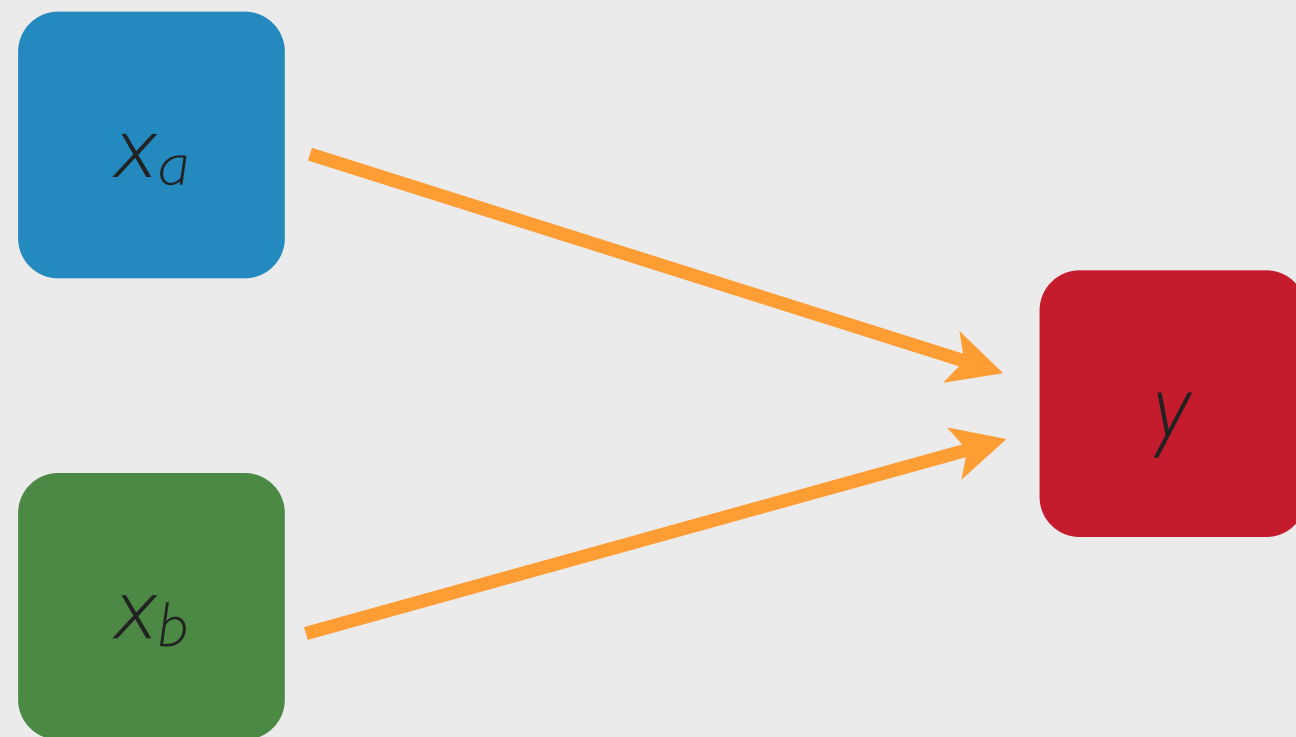
# DIFFERENCE OF MEANS



## 4. INDEPENDENT SAMPLES

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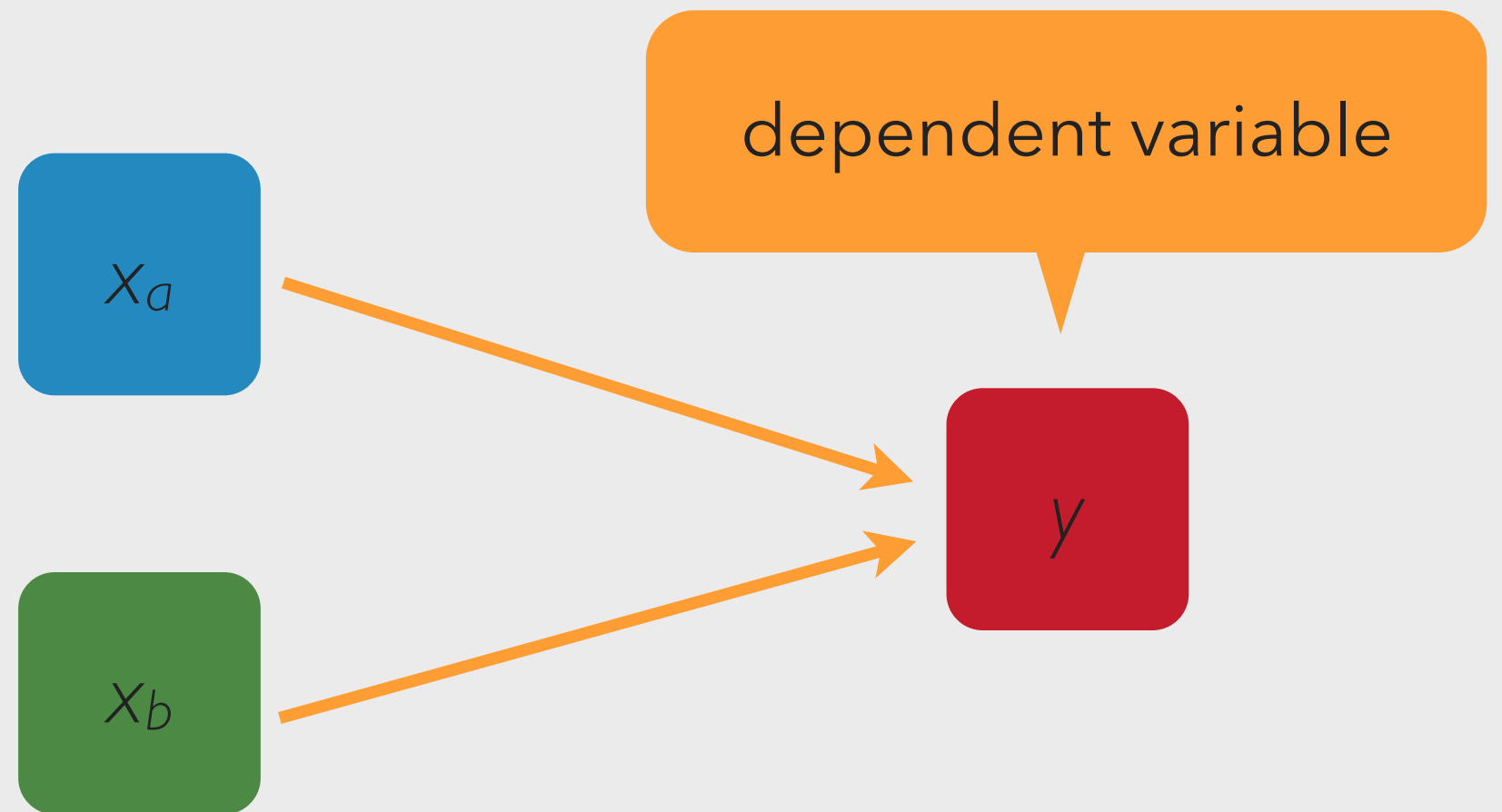
# MODEL



## 4. INDEPENDENT SAMPLES

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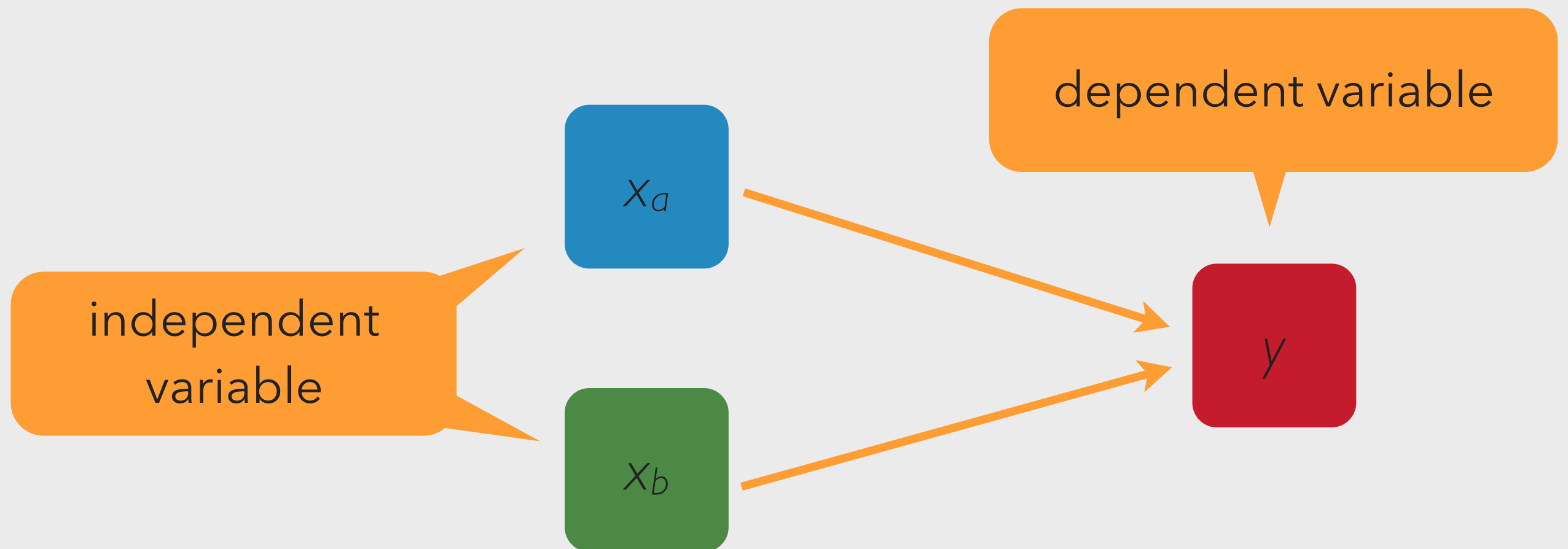
# MODEL



## 4. INDEPENDENT SAMPLES

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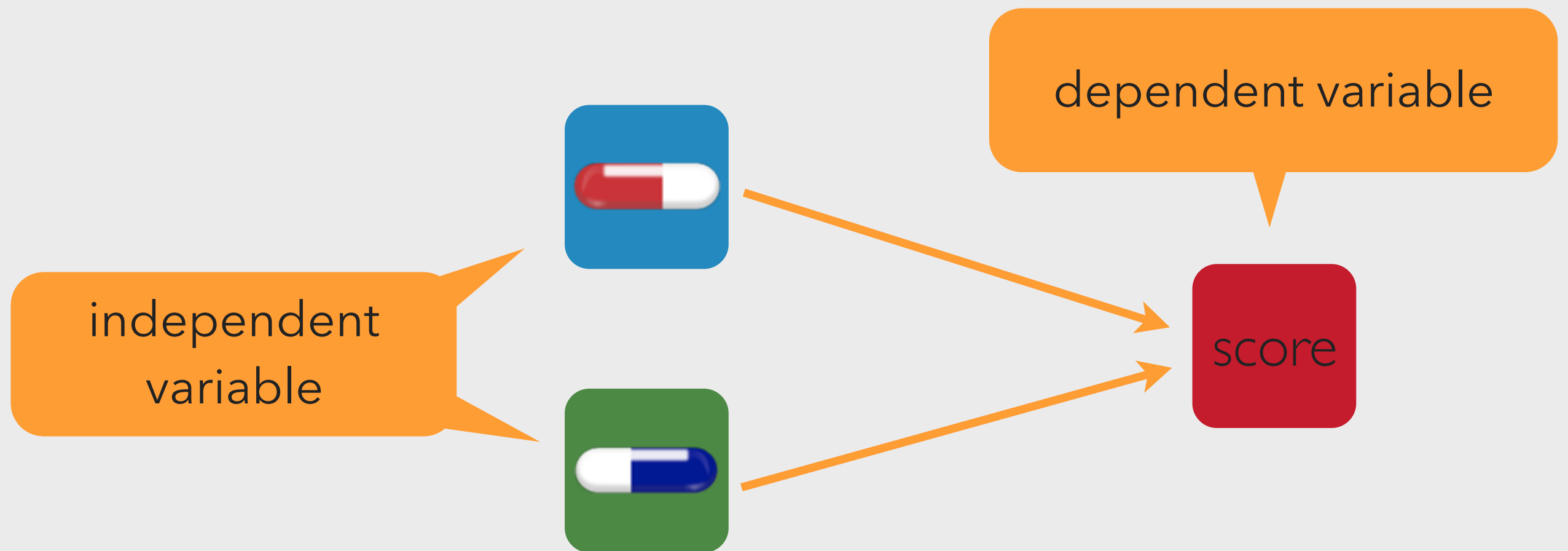
# MODEL



## 4. INDEPENDENT SAMPLES

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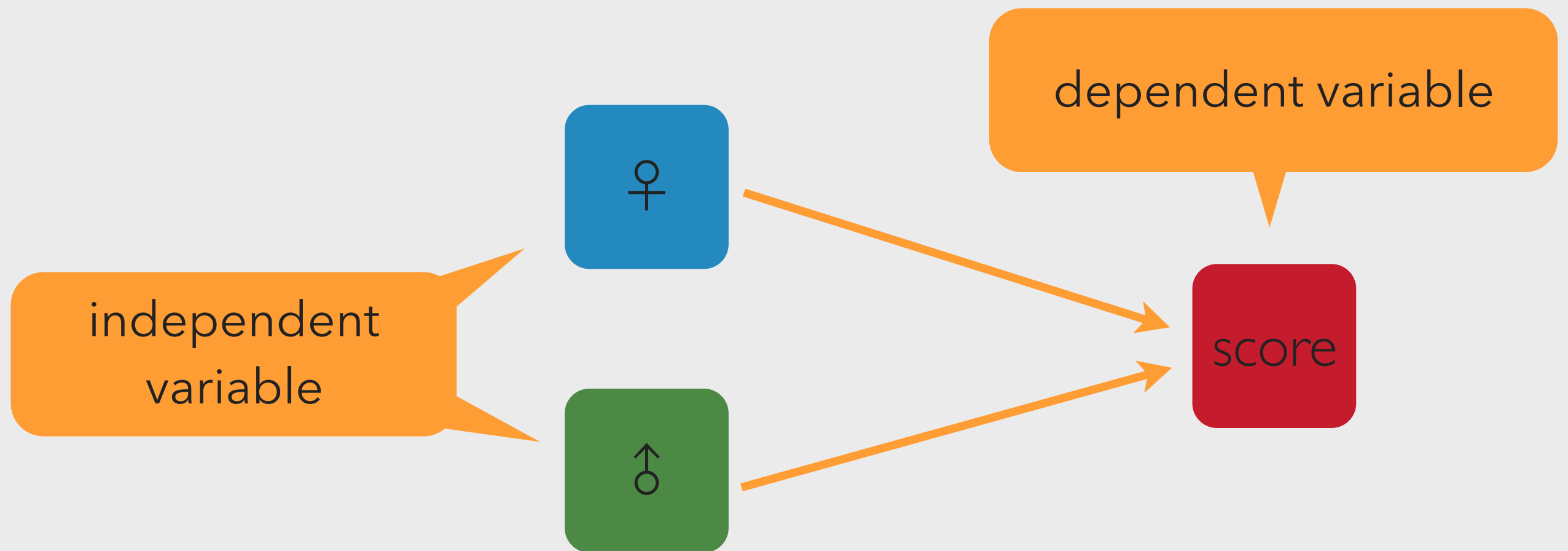
# MODEL



## 4. INDEPENDENT SAMPLES

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# MODEL



## 4. INDEPENDENT SAMPLES

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# HYPOTHESES

- ▶  $H_0$  = there is no difference in the mean of  $y$  between  $x_a$  and  $x_b$
- ▶  $H_1$  = there is a difference in the mean of  $y$  between  $x_a$  and  $x_b$



# ASSUMPTIONS

- ▶ dependent variable ( $y$ ) is continuous
- ▶ the distribution of  $y$  is approximately normal
- ▶ independent variable is binary ( $x_a$  and  $x_b$ )
- ▶ homogeneity of variance between  $x_a$  and  $x_b$
- ▶ observations are independent
- ▶  $v = n_a + n_b - 2$

## 4. INDEPENDENT SAMPLES

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# EQUATION ASSUMING HOMOGENEITY OF VARIANCE

$$t = \frac{\bar{X}_a - \bar{X}_b}{\sqrt{\frac{s_p^2}{n_a} + \frac{s_p^2}{n_b}}}$$

## 4. INDEPENDENT SAMPLES

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# EQUATION ASSUMING HOMOGENEITY OF VARIANCE

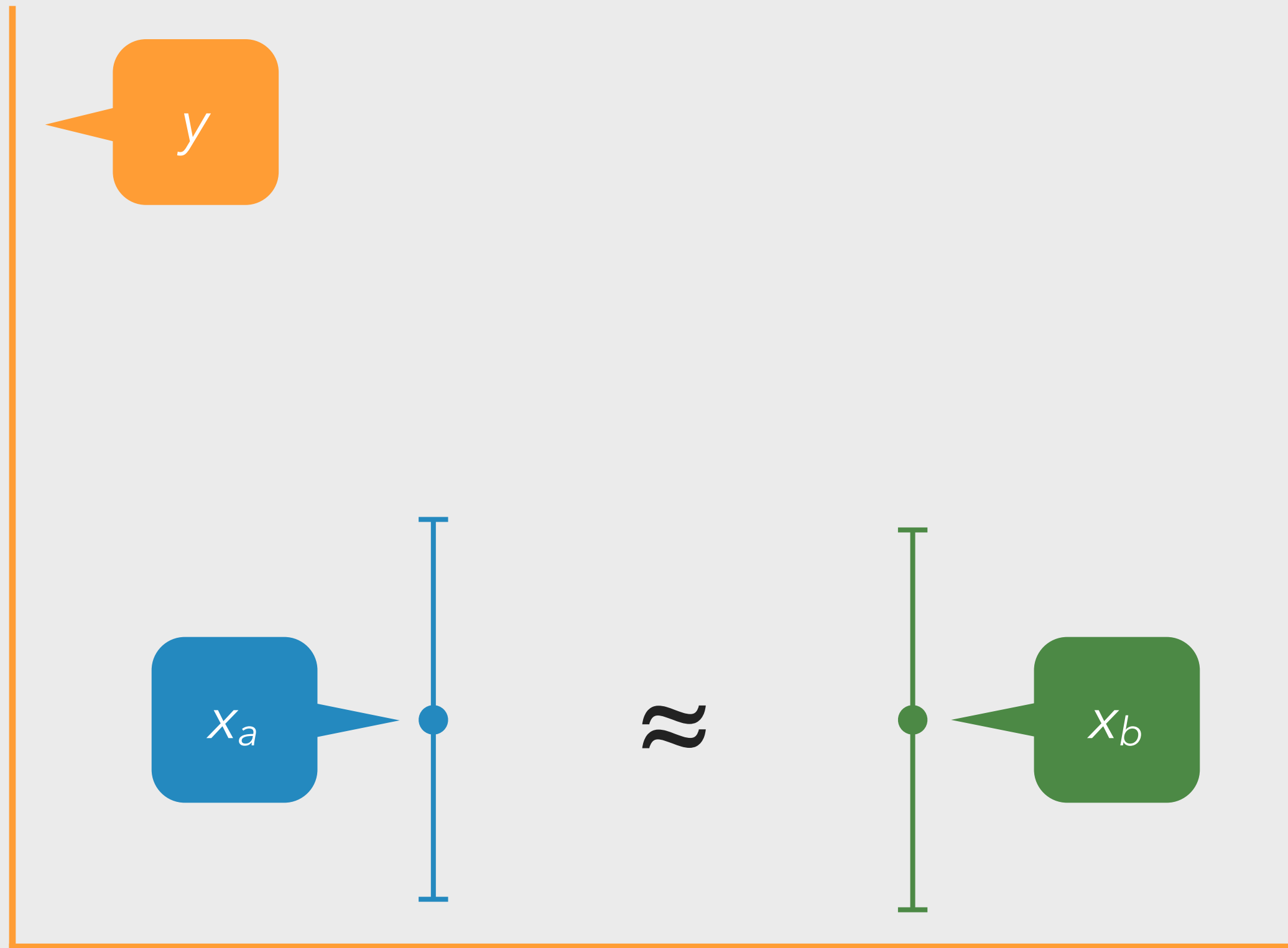
$$t = \frac{\bar{X}_a - \bar{X}_b}{\sqrt{\frac{s_p^2}{n_a} + \frac{s_p^2}{n_b}}}$$

pooled variance

# POOLED VARIANCE EQUATION

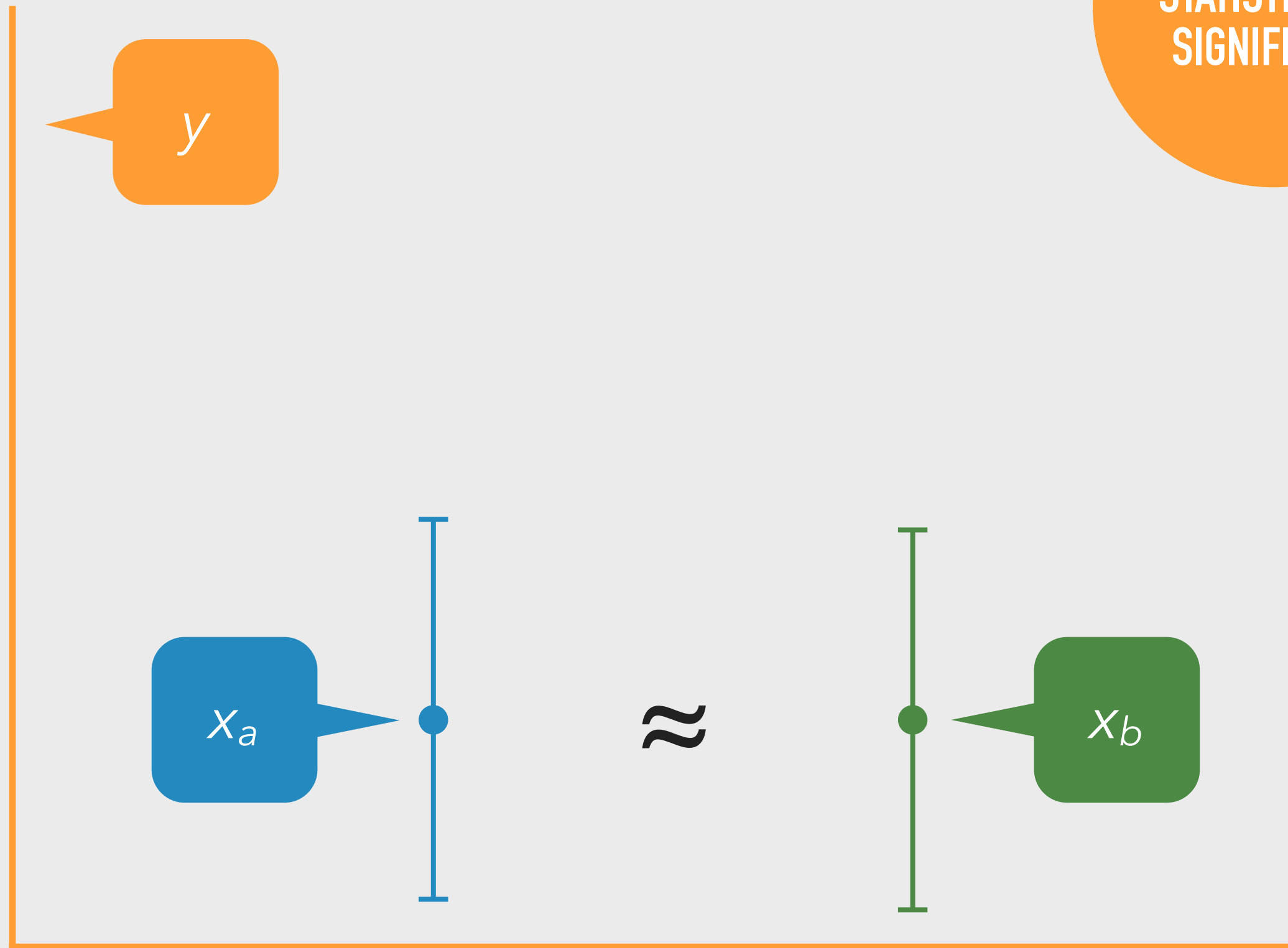
$$s_p^2 = \frac{(n_a - 1) s_a^2 + (n_b - 1) s_b^2}{n_a + n_b - 2}$$

# DIFFERENCE OF MEANS

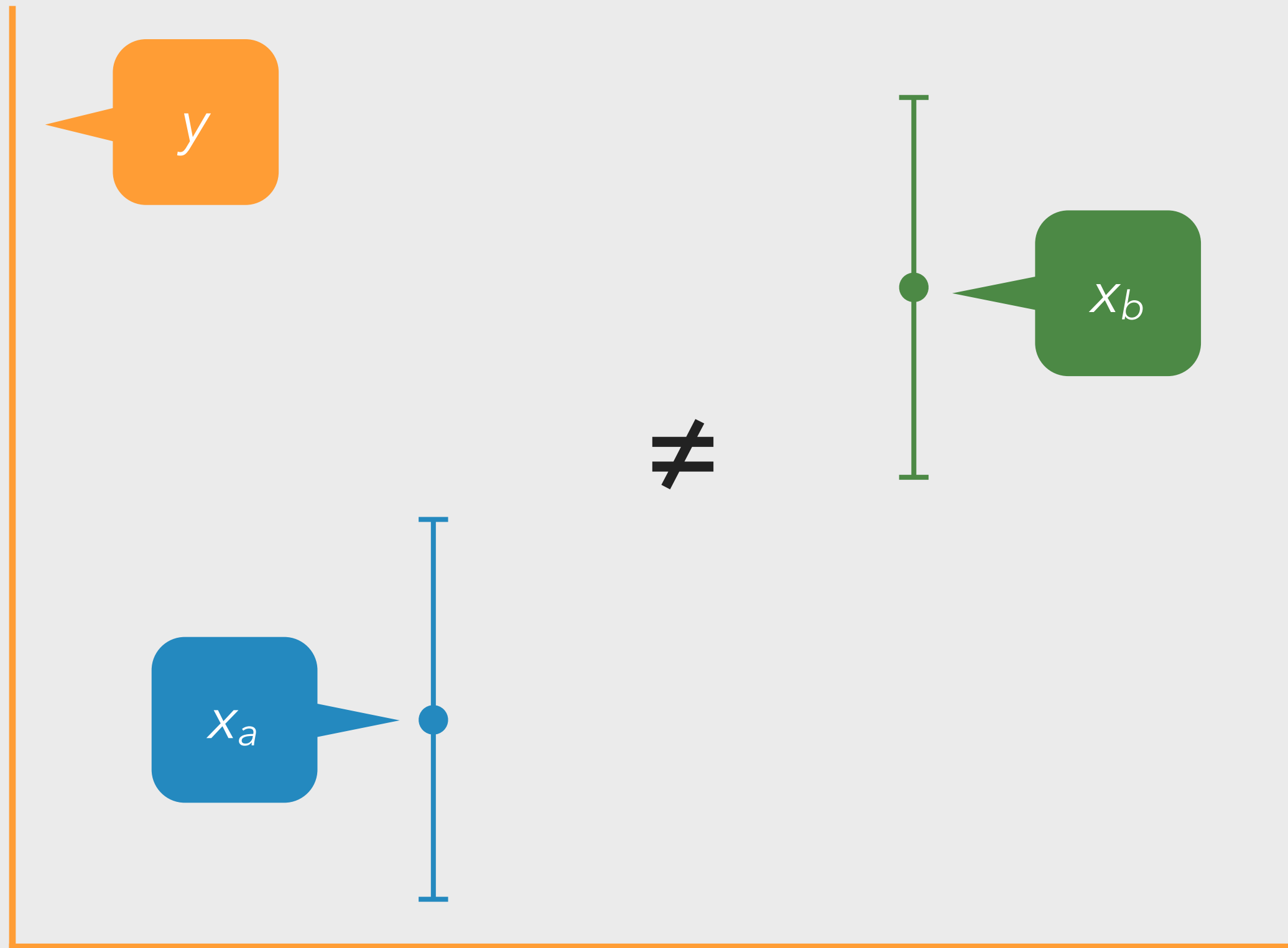


# DIFFERENCE OF MEANS

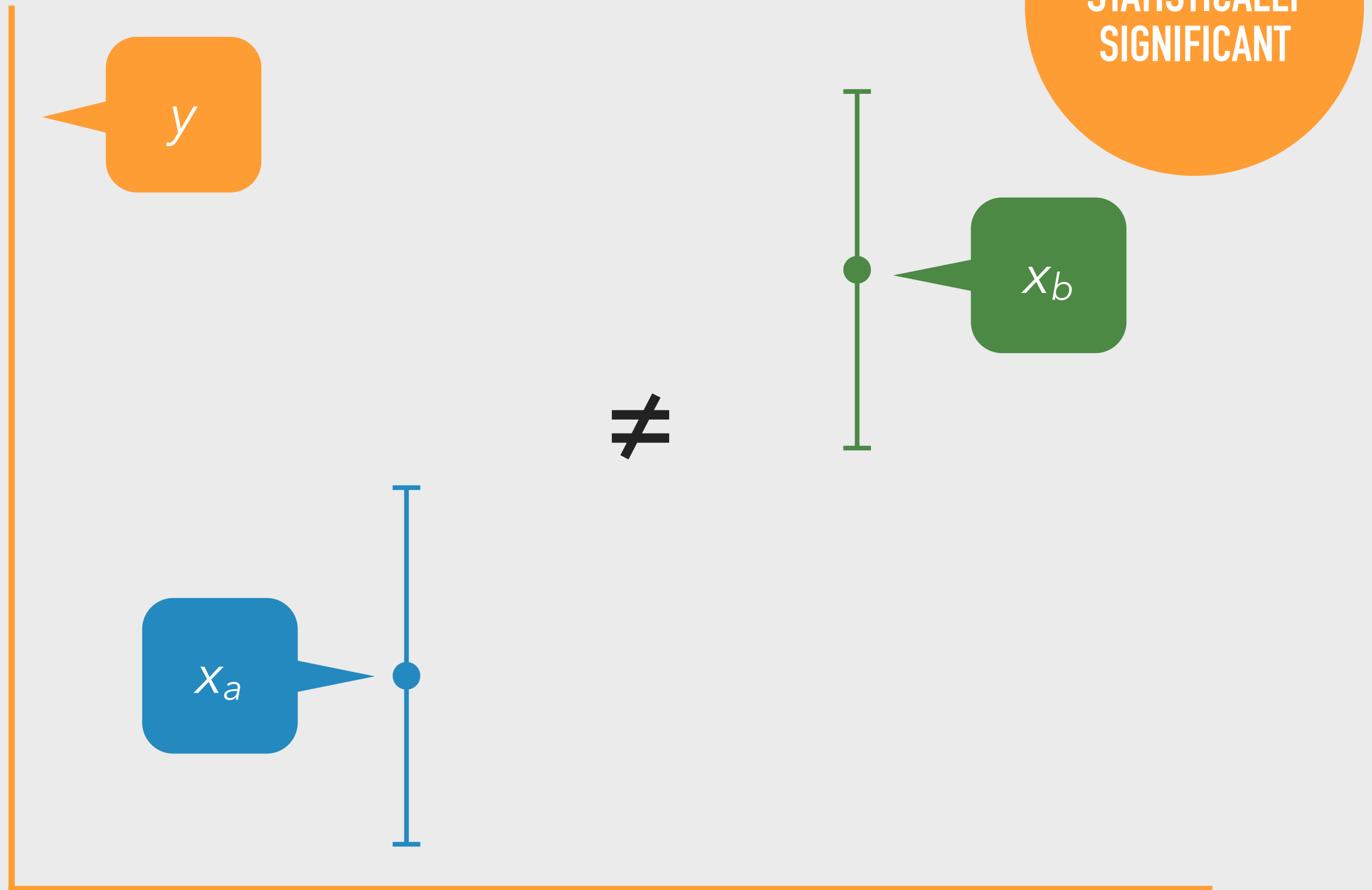
T-TEST IS **NOT**  
STATISTICALLY  
SIGNIFICANT



# DIFFERENCE OF MEANS

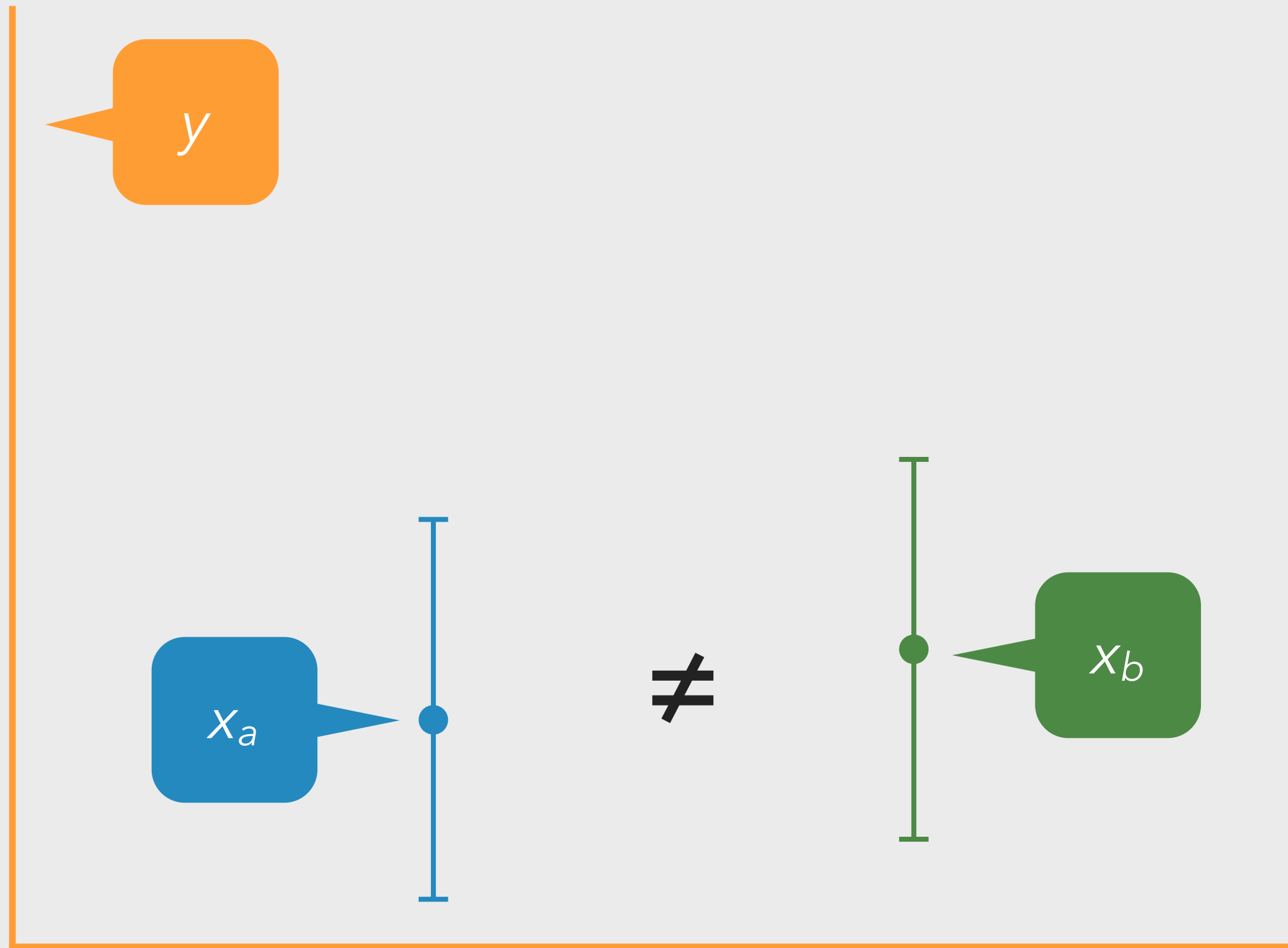


# DIFFERENCE OF MEANS



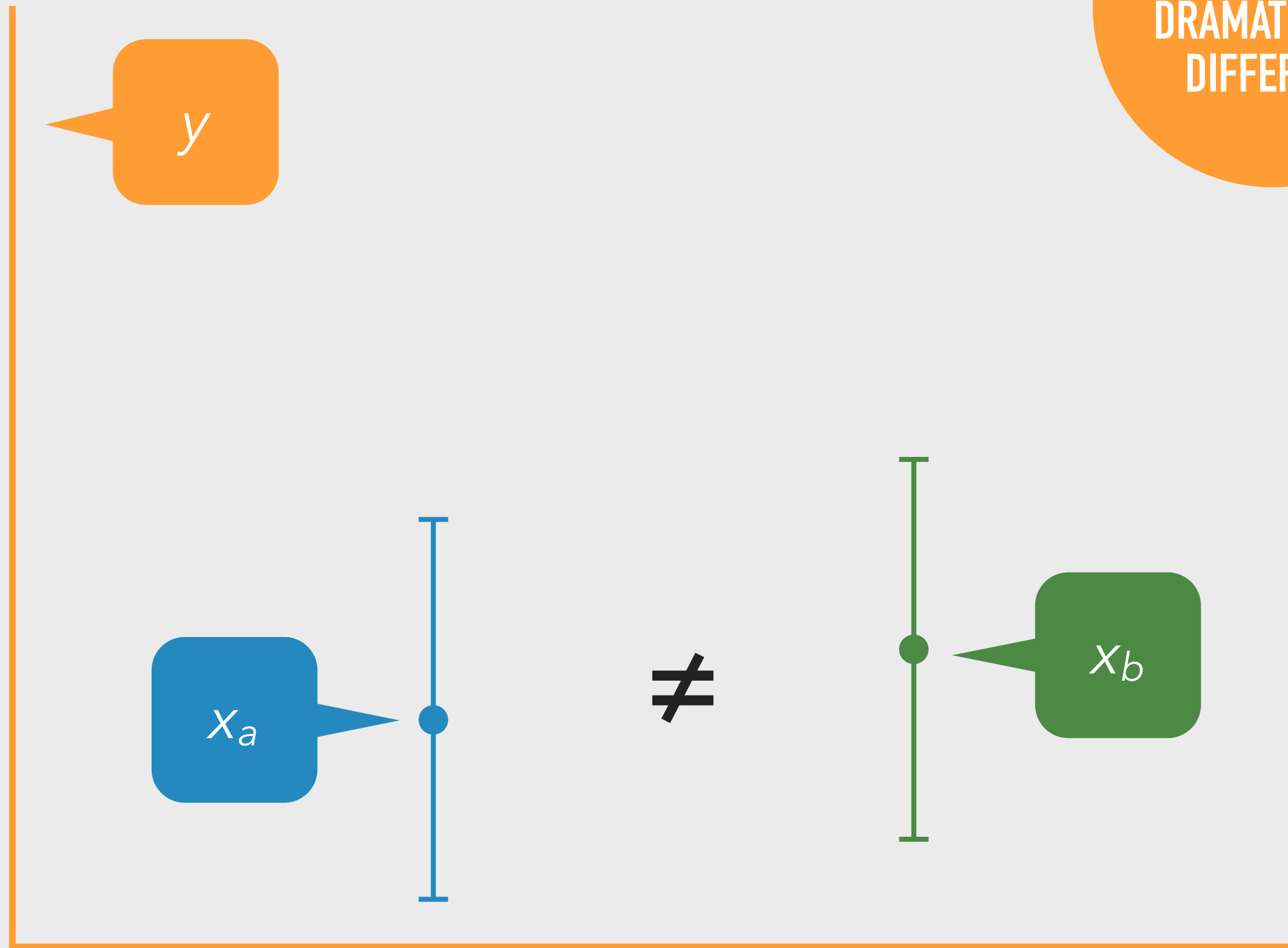


# DIFFERENCE OF MEANS



# DIFFERENCE OF MEANS

THE MEANS DO  
NOT HAVE TO BE  
DRAMATICALLY  
DIFFERENT



# ASSUMPTIONS

- ▶ dependent variable ( $y$ ) is continuous
- ▶ the distribution of  $y$  is approximately normal
- ▶ independent variable is binary ( $x_a$  and  $x_b$ )
- ▶ homogeneity of variance between  $x_a$  and  $x_b$
- ▶ observations are independent
- ▶  $v = n_a + n_b - 2$

# EQUATION IF HOMOGENEITY OF VARIANCE CANNOT BE ASSUMED

$$t = \frac{\bar{X}_a - \bar{X}_b}{\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}}$$

# EQUATION IF HOMOGENEITY OF VARIANCE CANNOT BE ASSUMED

$$t = \frac{\bar{X}_a - \bar{X}_b}{\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}}$$

variance values for  
each subgroup

## 4. INDEPENDENT SAMPLES

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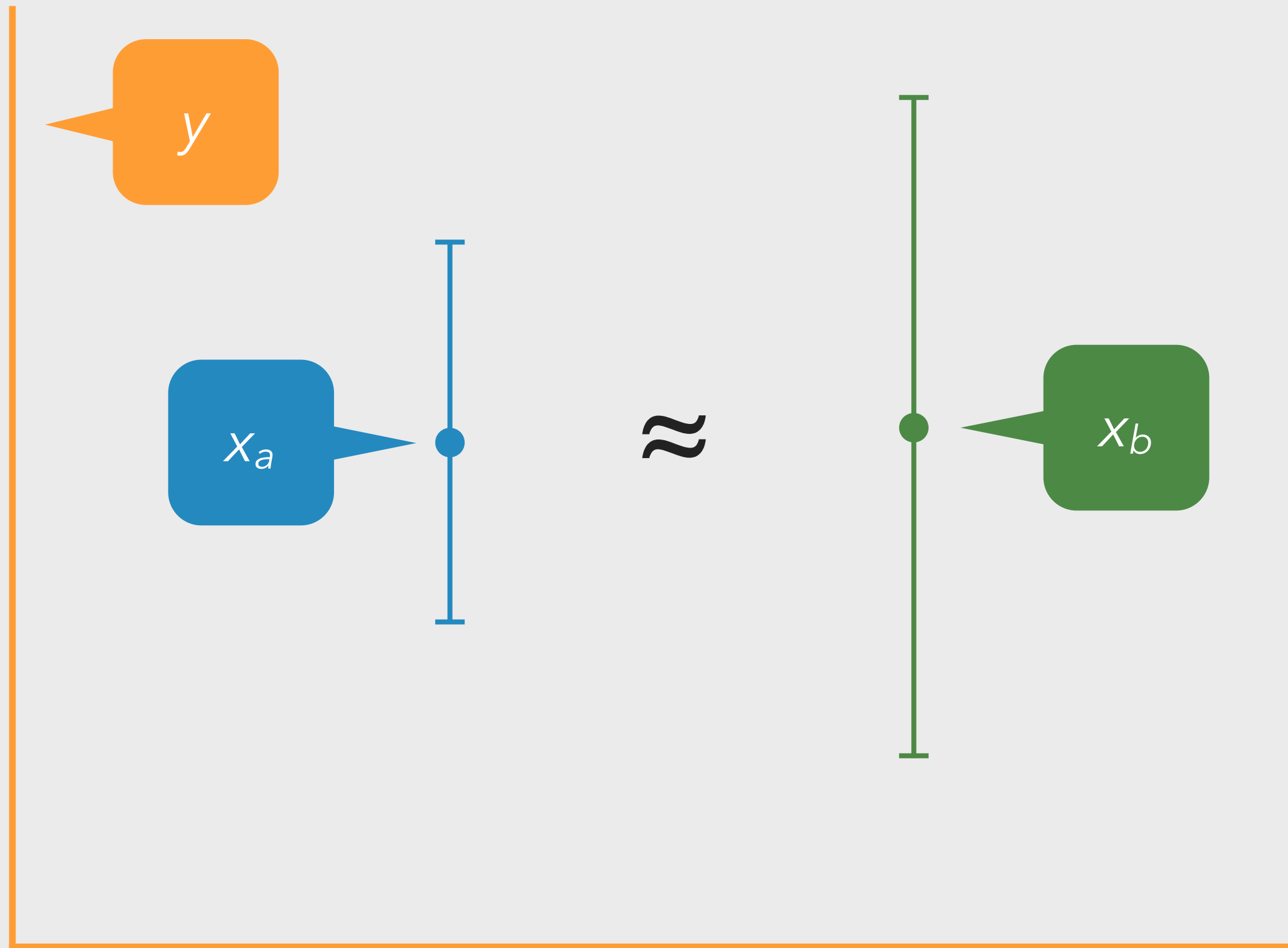
CAUTION! CAUTION! CAUTION!

$$t = \frac{\bar{X}_a - \bar{X}_b}{\sqrt{\frac{s_p^2}{n_a} + \frac{s_p^2}{n_b}}} \neq t = \frac{\bar{X}_a - \bar{X}_b}{\sqrt{\frac{s_a^2}{n_a} + \frac{s_b^2}{n_b}}}$$

# WELCH'S CORRECTED DEGREES OF FREEDOM

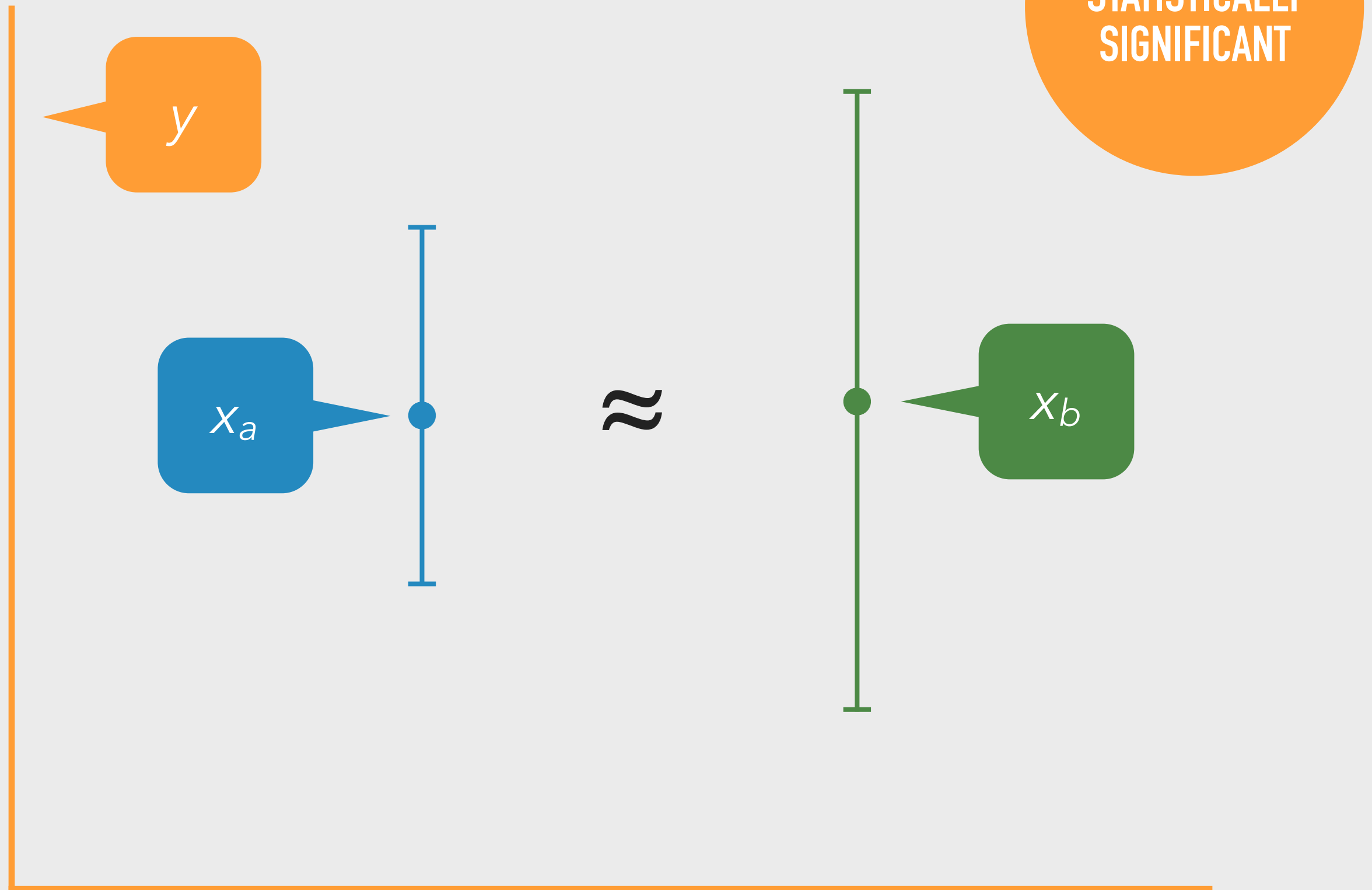
$$v \approx \frac{\left( \frac{s_a^2}{n_a} + \frac{s_b^2}{n_b} \right)^2}{\frac{s_a^4}{(n_a^2)(n_a - 1)} + \frac{s_b^4}{(n_b^2)(n_b - 1)}}$$

# DIFFERENCE OF MEANS

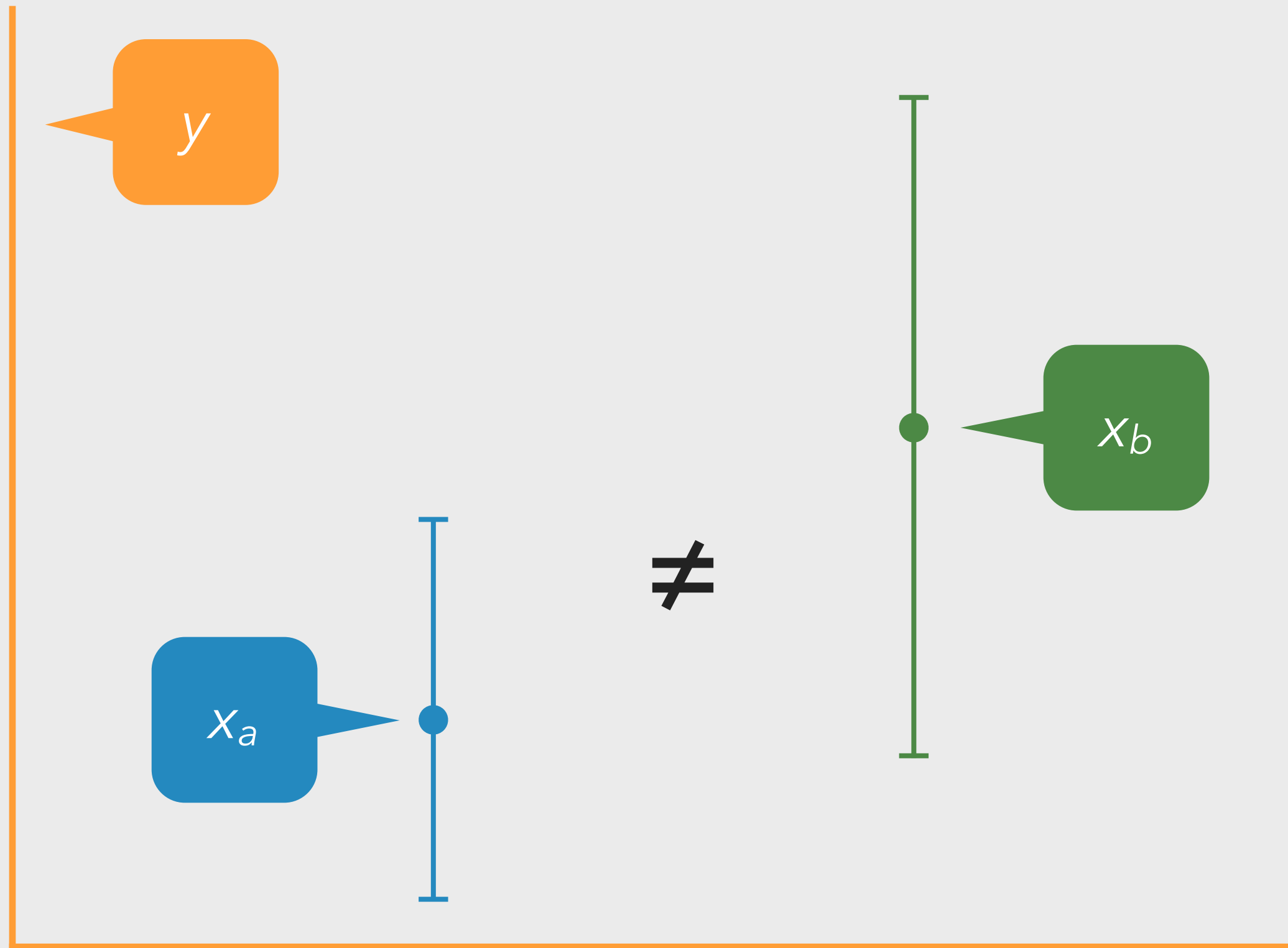




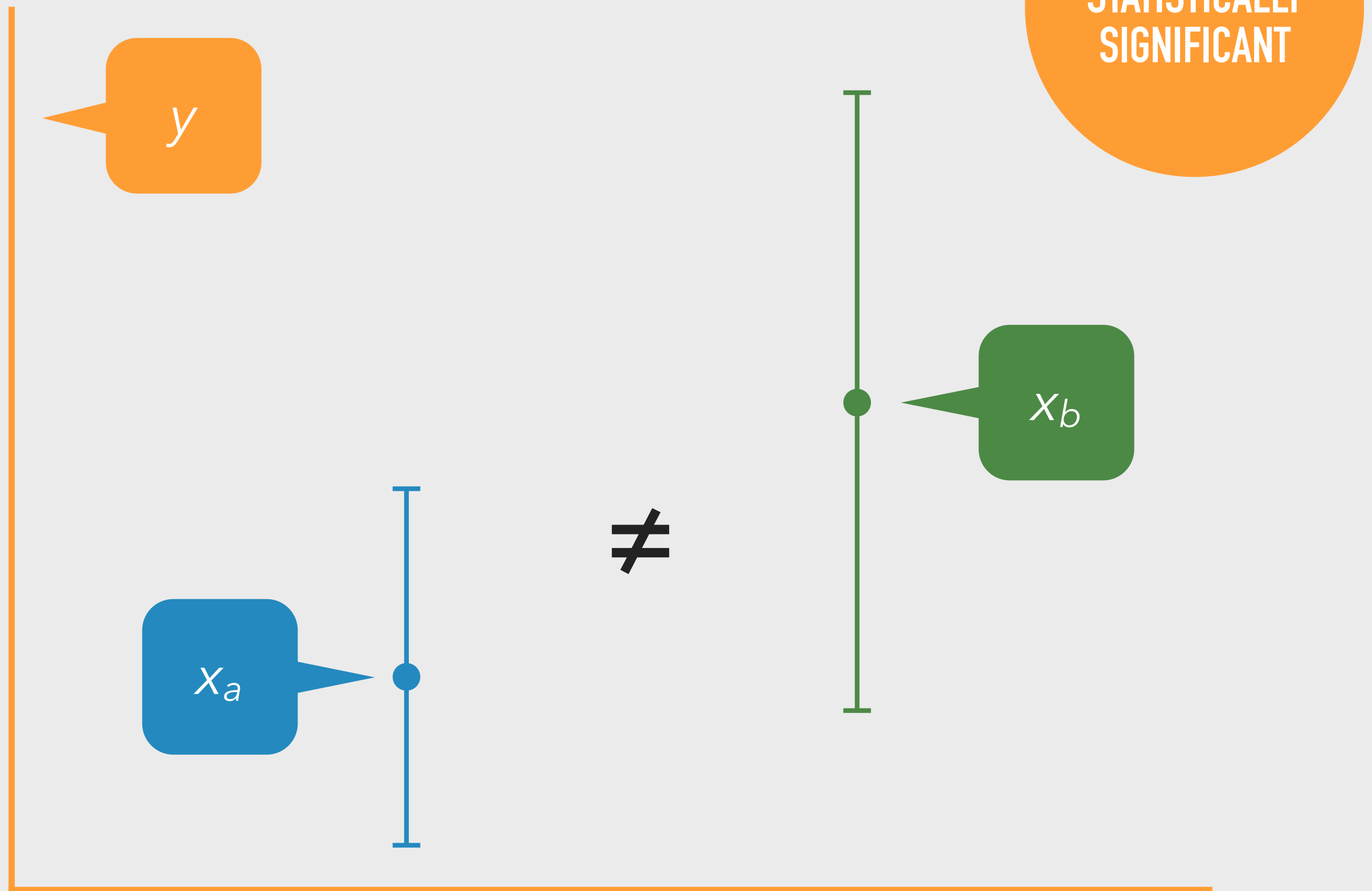
# DIFFERENCE OF MEANS



# DIFFERENCE OF MEANS

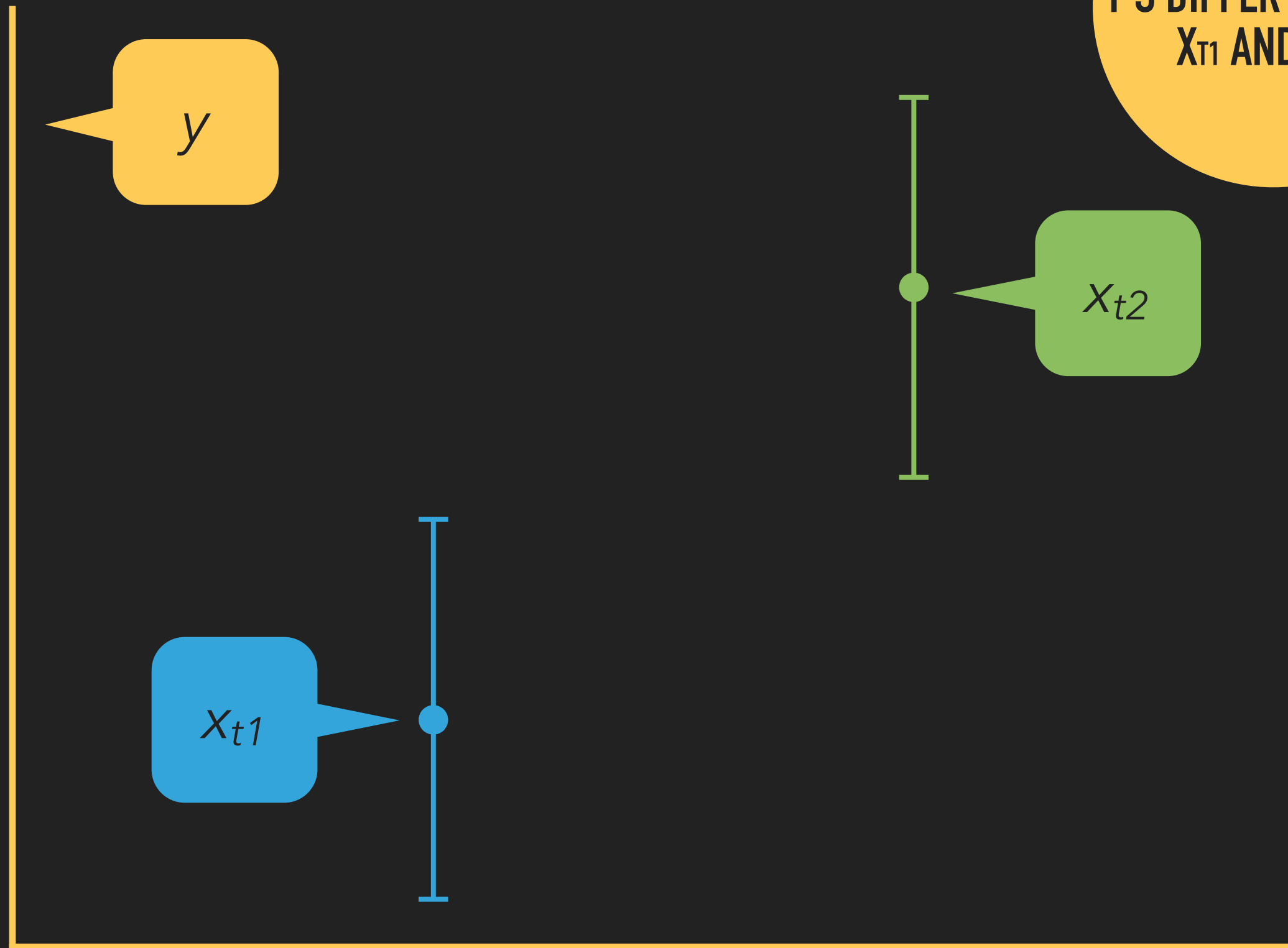


# DIFFERENCE OF MEANS

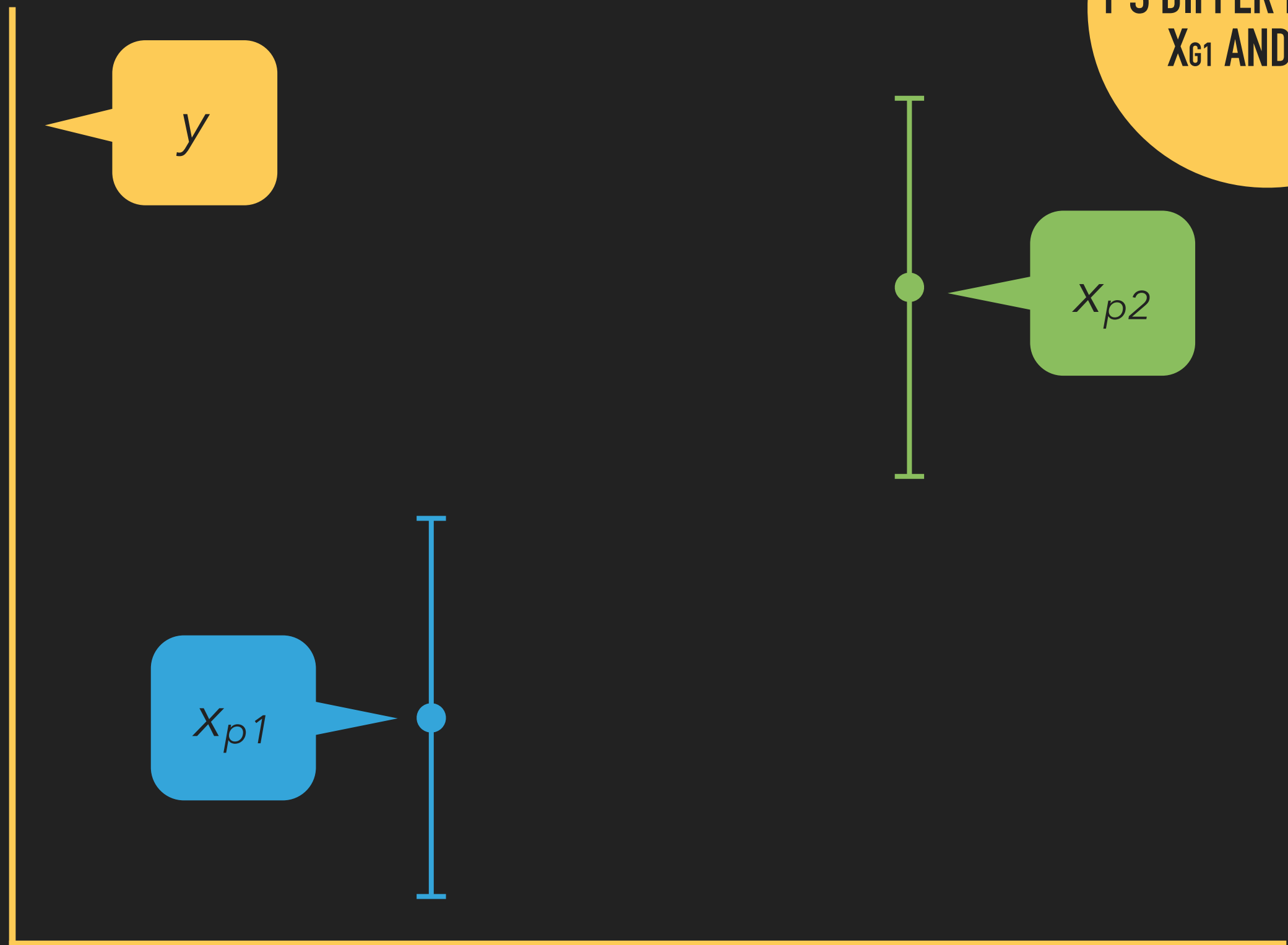


# 5 DEPENDENT SAMPLES

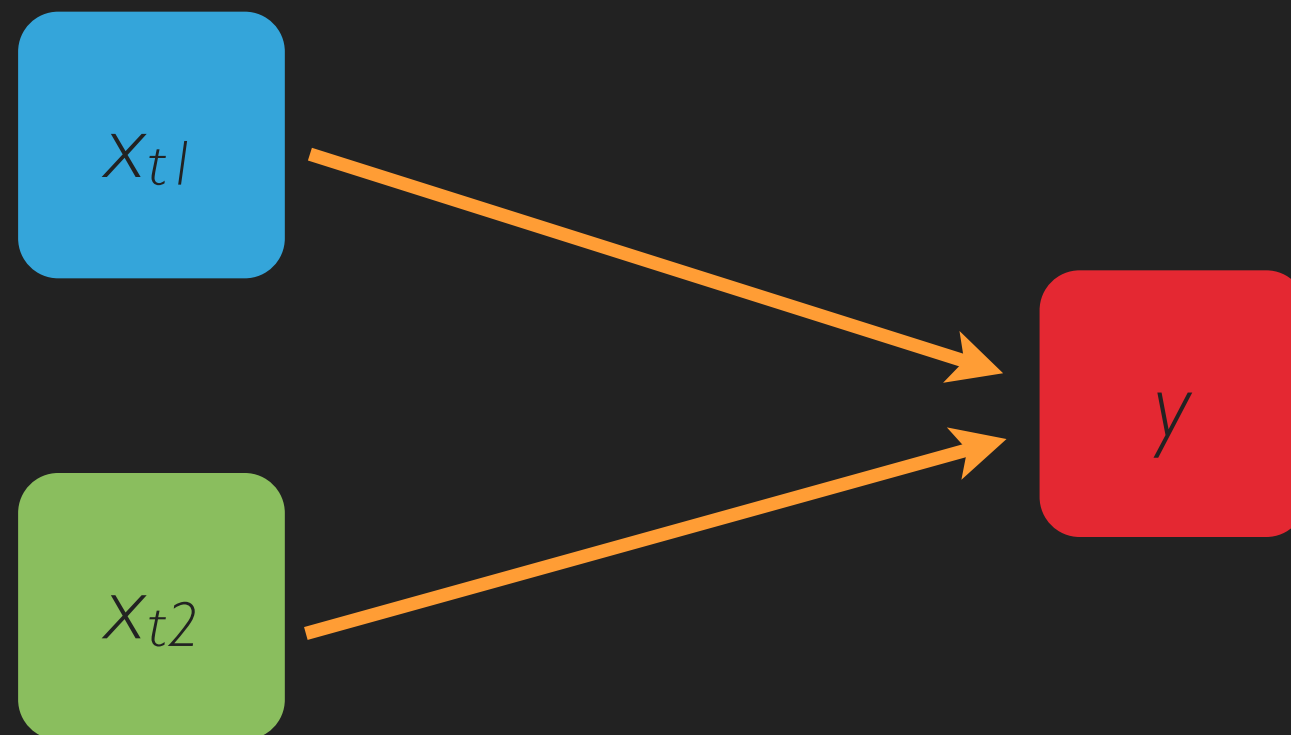
# DIFFERENCE OF MEANS



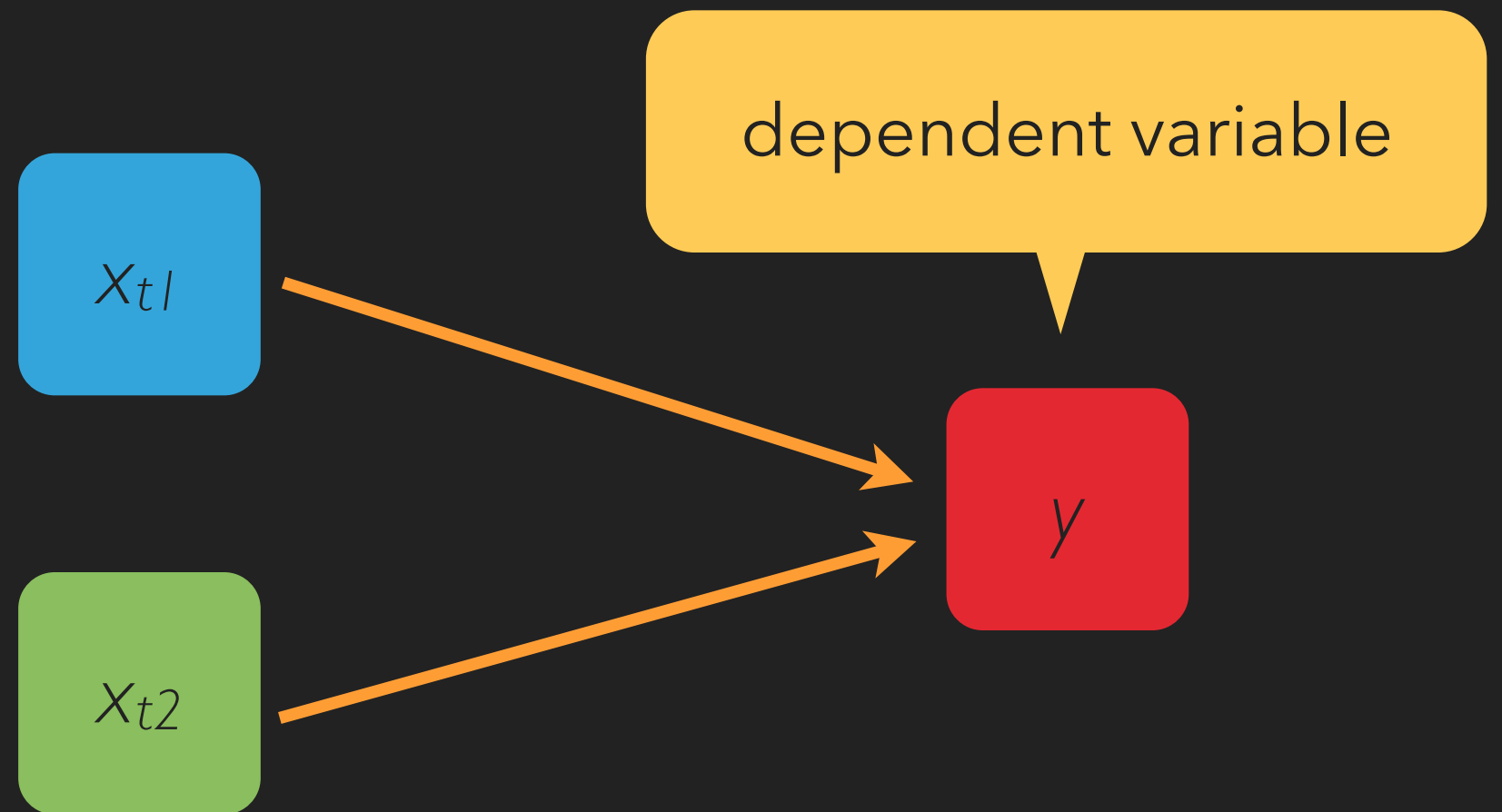
# DIFFERENCE OF MEANS



# MODEL

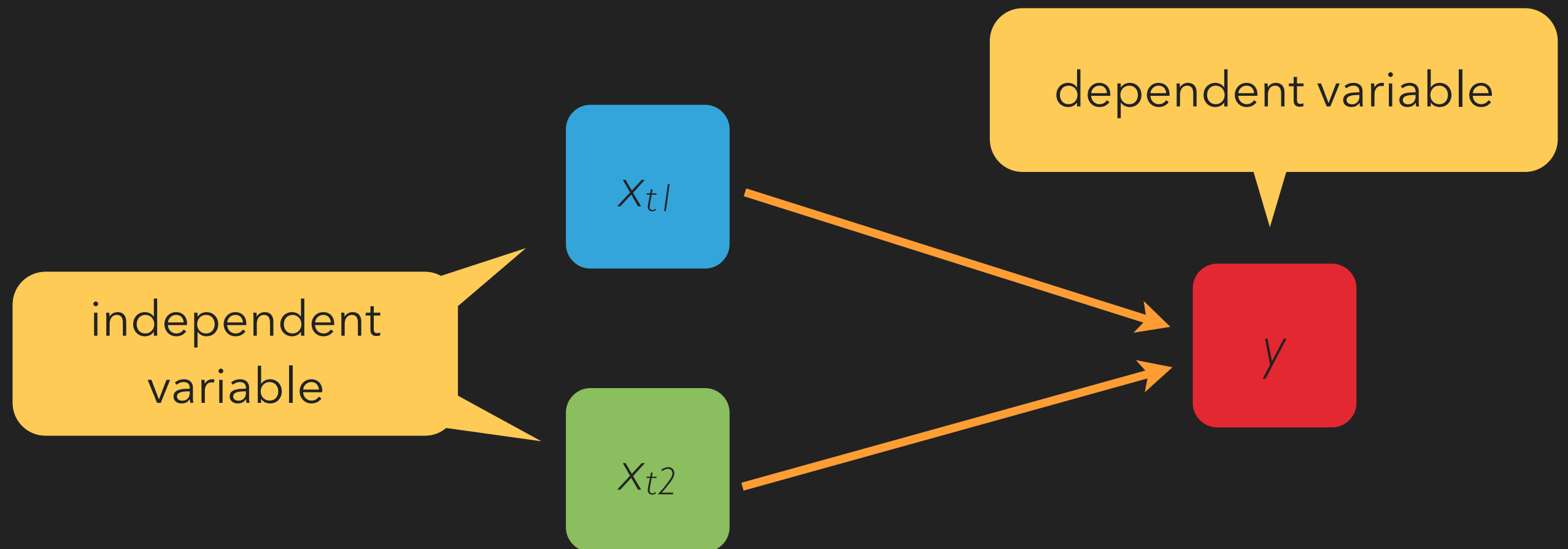


# MODEL

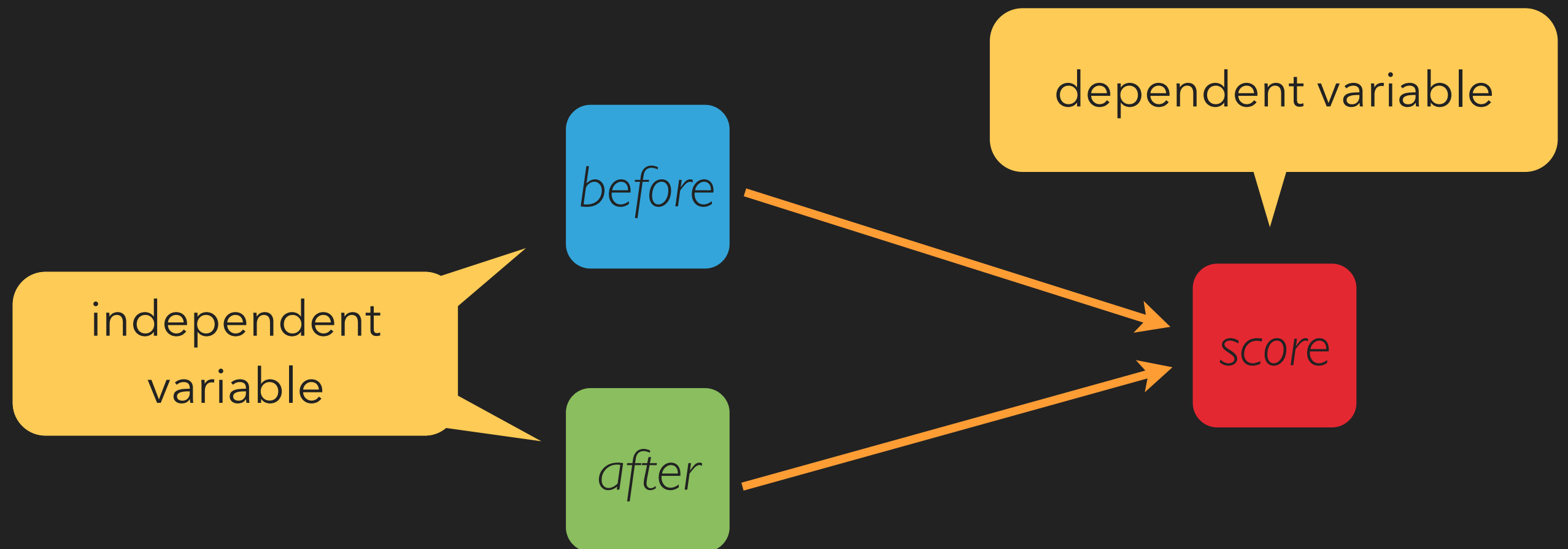




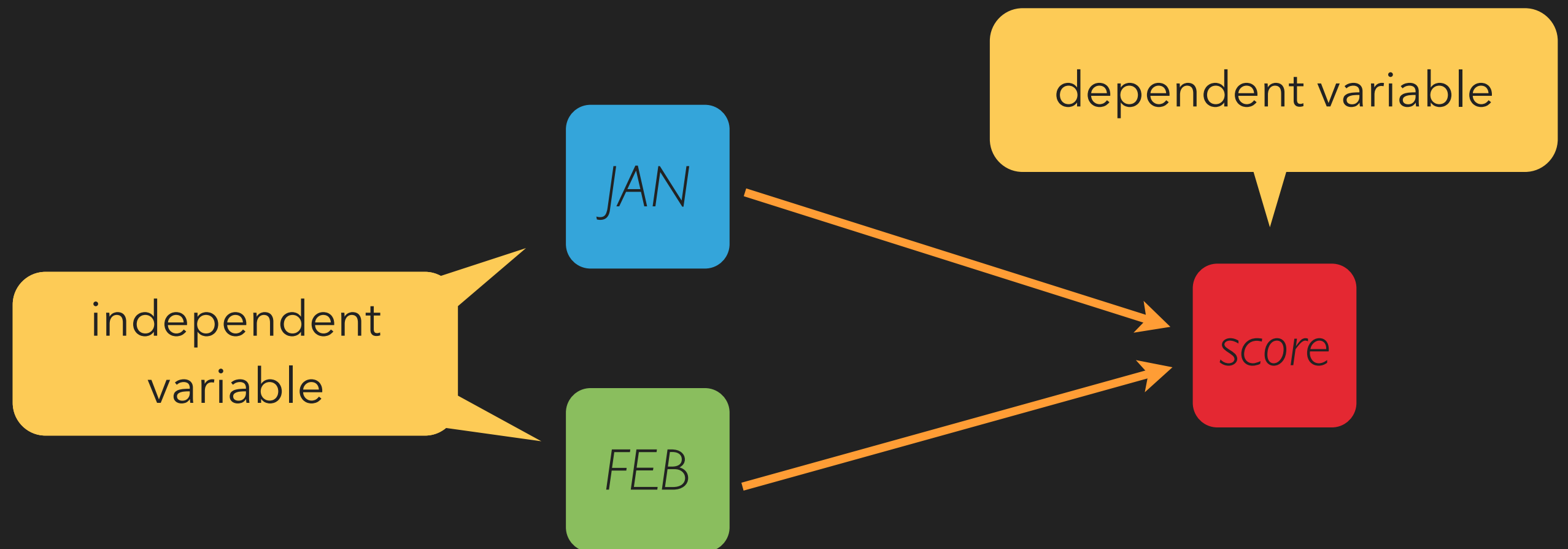
# MODEL



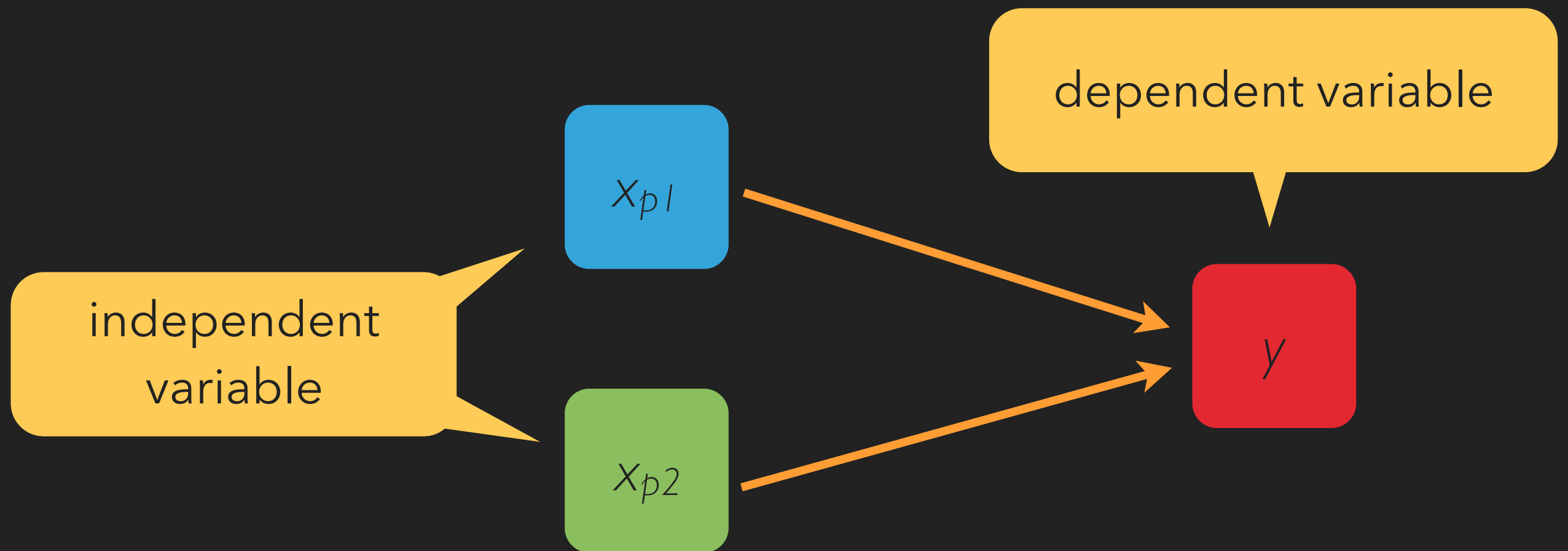
# MODEL



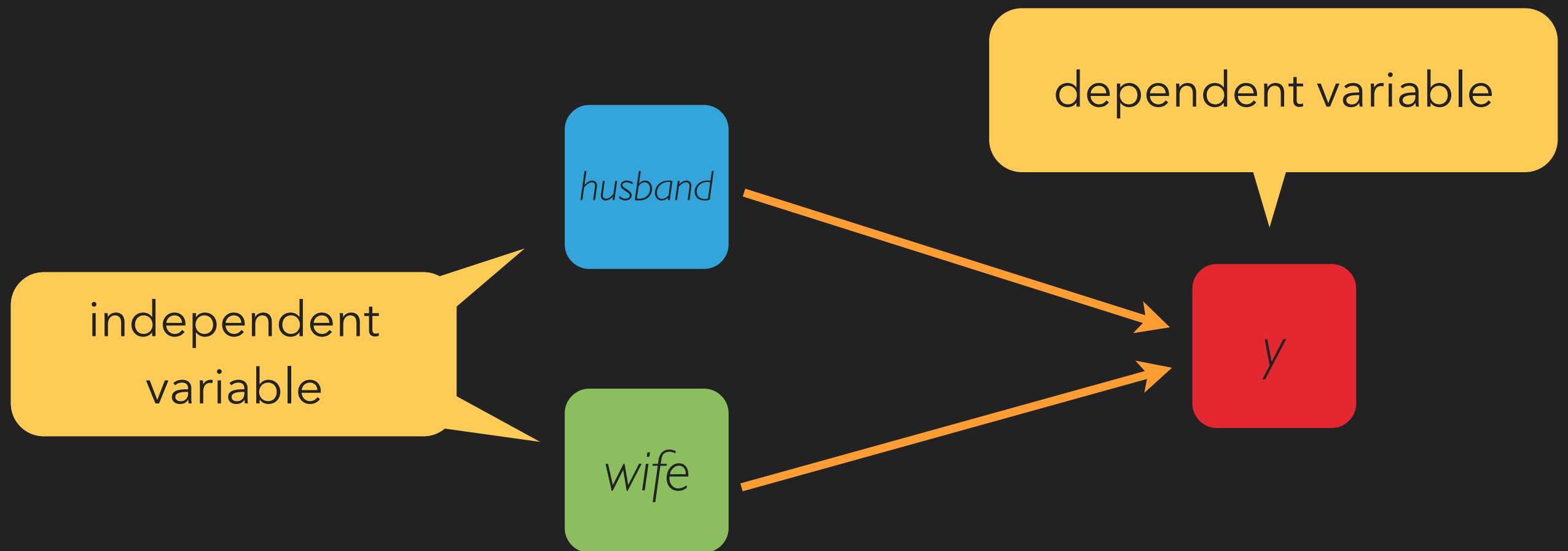
# MODEL



# MODEL



# MODEL



# HYPOTHESES

- ▶  $H_0$  = there is no difference in the mean of  $y$  between  $x_{t1}$  and  $x_{t2}$
- ▶  $H_1$  = there is a difference in the mean of  $y$  between  $x_{t1}$  and  $x_{t2}$

# HYPOTHESES

- ▶  $H_0$  = there is no difference in the mean of  $y$  between  $x_{g1}$  and  $x_{g2}$
- ▶  $H_1$  = there is a difference in the mean of  $y$  between  $x_{g1}$  and  $x_{g2}$

# ASSUMPTIONS

- ▶ dependent variable ( $y$ ) is continuous
- ▶ independent variable is binary ( $x_{g1}$  and  $x_{g2}$ )
- ▶ homogeneity of variance between  $x_{g1}$  and  $x_{g2}$
- ▶ the distribution of the differences between  $x_{g1}$  and  $x_{g2}$  is normally distributed
- ▶ scores are dependent



## 5. DEPENDENT SAMPLES

---

# EQUATION

$$t = \frac{\bar{d}}{\sqrt{\frac{s_d^2}{n}}}$$

# EQUATION

$$t = \frac{\bar{d}}{\sqrt{\frac{s_d^2}{n}}}$$

mean of difference  
between groups

# EQUATION

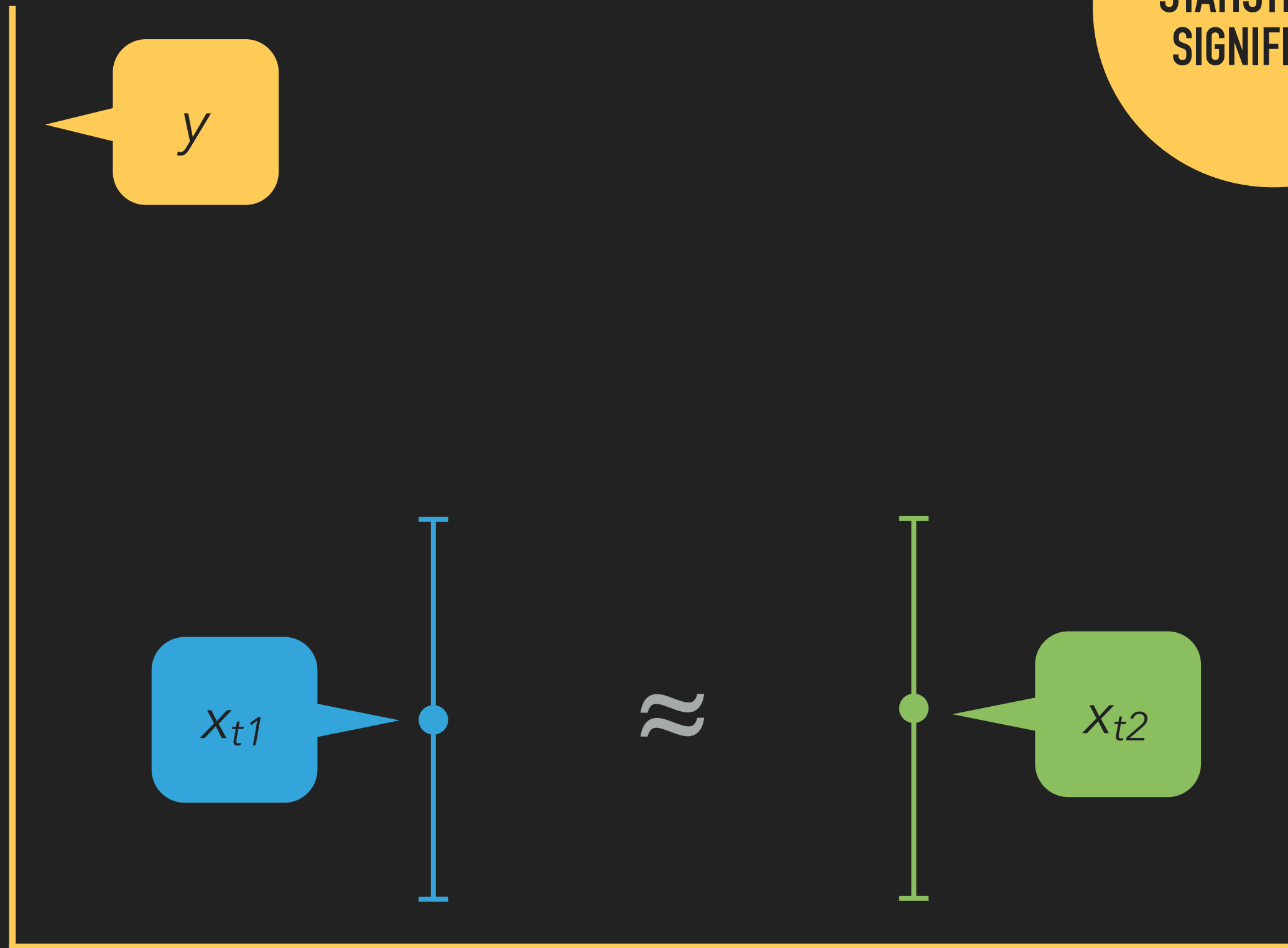
$$t = \frac{\bar{d}}{\sqrt{\frac{s_d^2}{n}}}$$

mean of difference  
between groups

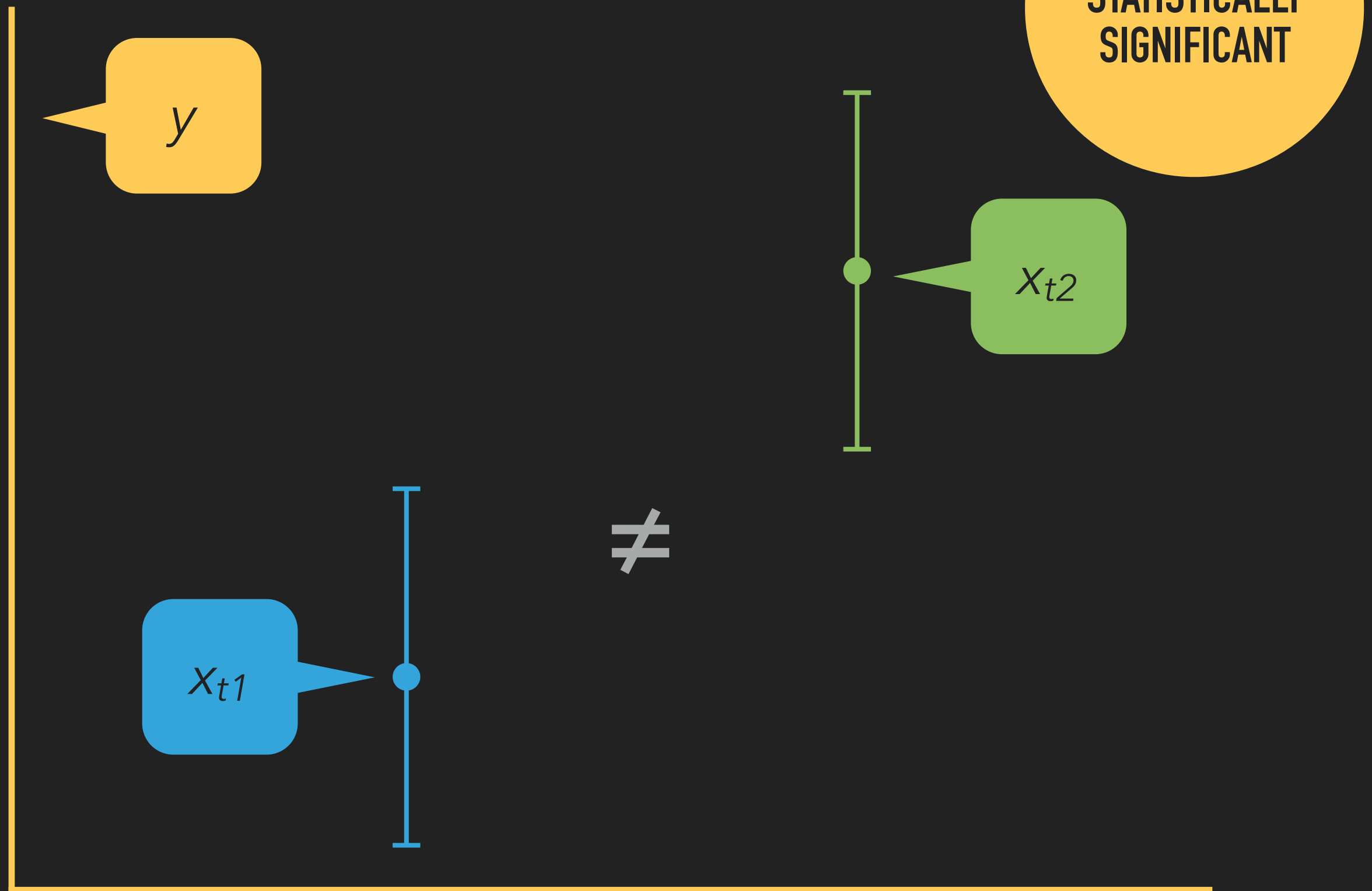
variance of difference  
between groups

# DIFFERENCE OF MEANS

T-TEST IS NOT  
STATISTICALLY  
SIGNIFICANT



# DIFFERENCE OF MEANS



# LONG DATA

<i>participant</i>	<i>score</i>	<i>timePoint</i>
jane	10	before
jane	12	after
john	15	before
john	14	after

WIDE DATA

<i>participant</i>	<i>score1</i>	<i>score2</i>
jane	10	12
john	15	14
joe	12	12
jessica	8	11

# RESHAPING DATA

<i>participant</i>	<i>score</i>	<i>timePoint</i>
jane	10	before
jane	12	after
john	15	before
john	14	after
joe	12	before
joe	12	after
jessica	8	before
jessica	11	after



<i>participant</i>	<i>score1</i>	<i>score2</i>
jane	10	12
john	15	14
joe	12	12
jessica	8	11



# 6 EFFECT SIZES

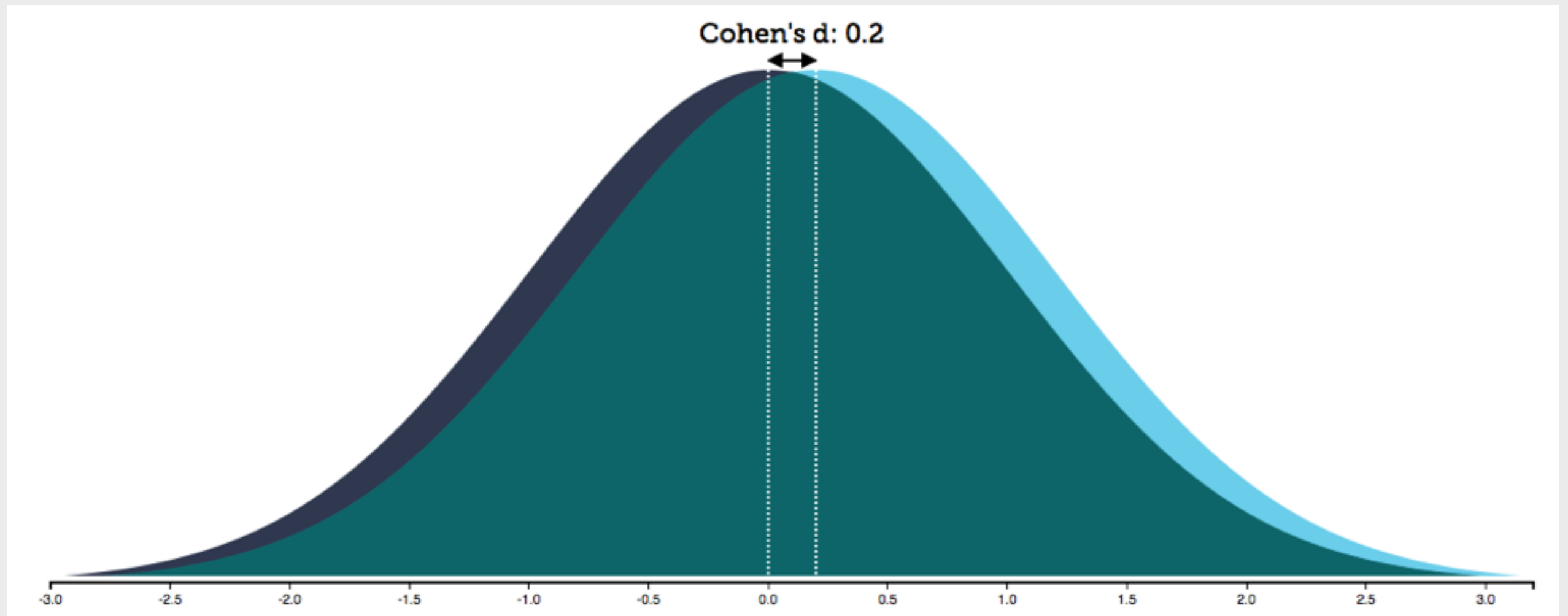
# PROBLEM

Statistical Significance  $\neq$  Real World Significance

## 6. EFFECT SIZES

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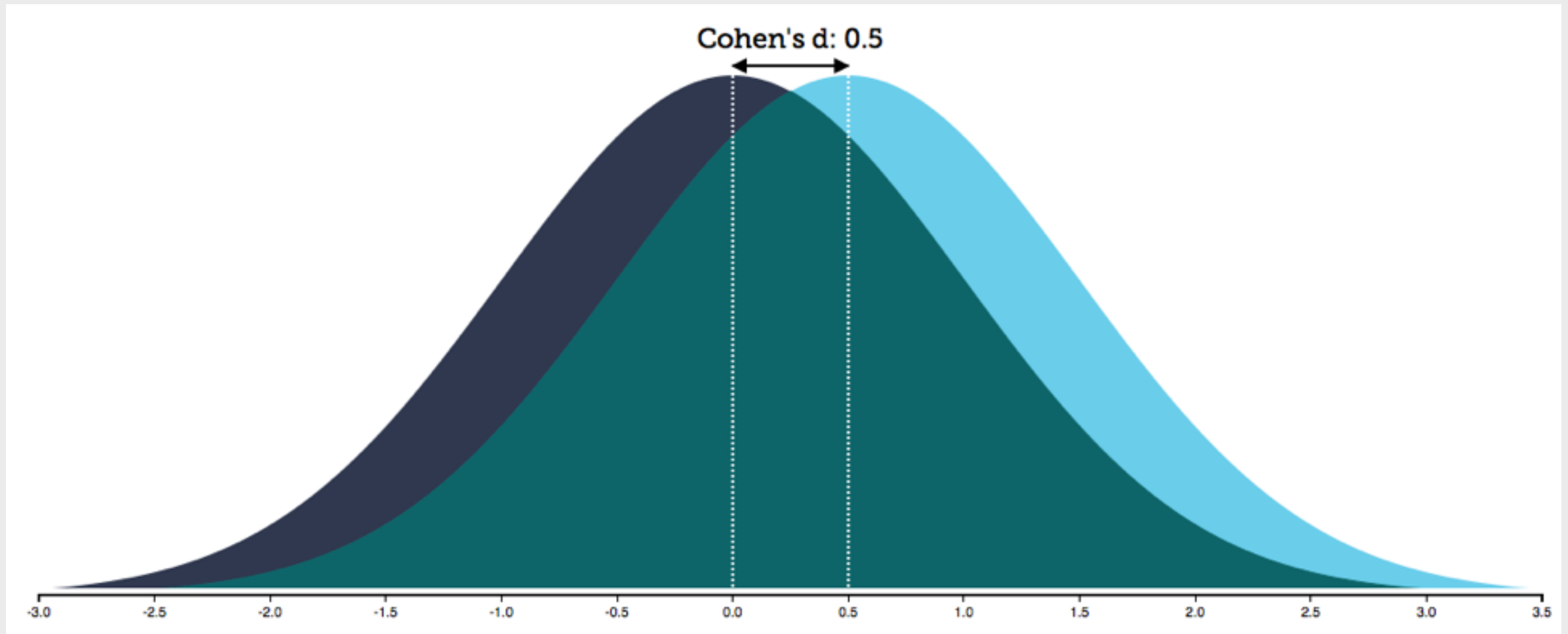
# COHEN'S D INTERPRETATION



## 6. EFFECT SIZES

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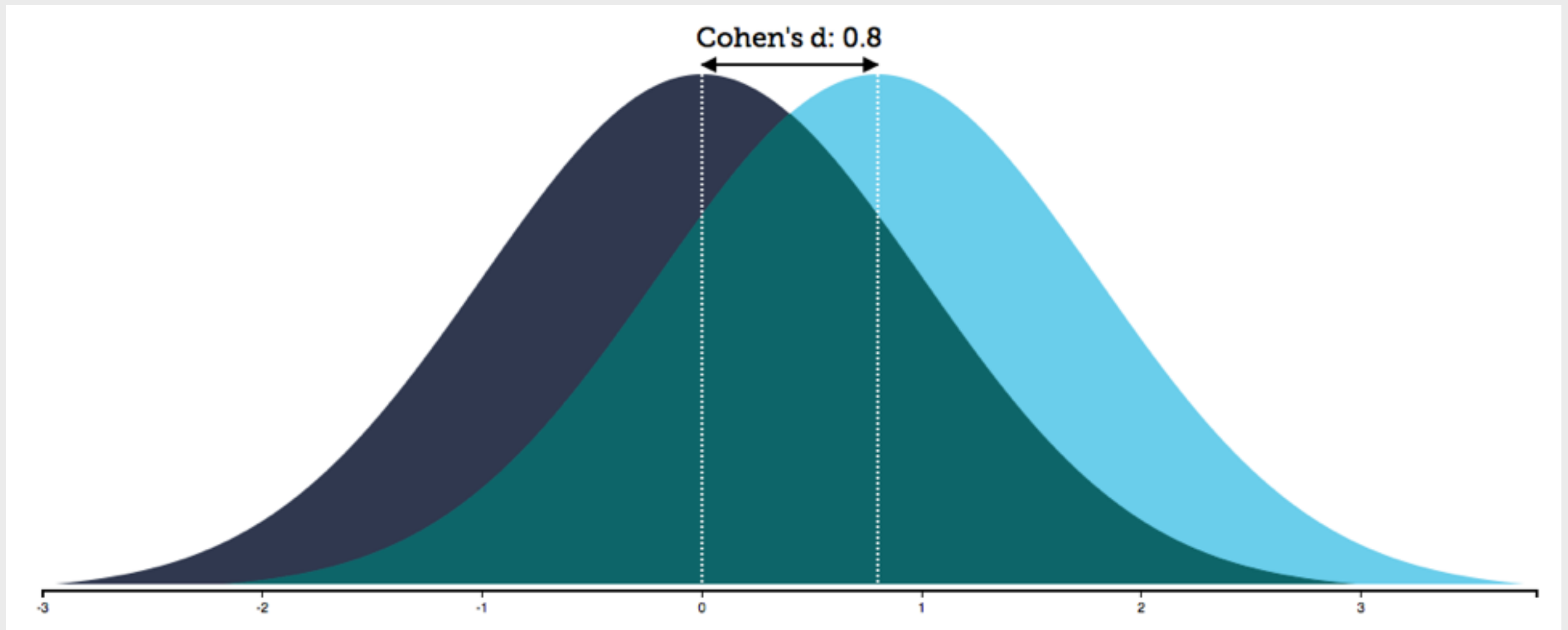
# COHEN'S D INTERPRETATION



## 6. EFFECT SIZES

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# COHEN'S D INTERPRETATION



# COHEN'S D EQUATION

$$d = \frac{\bar{x}_a - \bar{x}_b}{\sqrt{\frac{(n_a - 1)s_a^2 + (n_b - 1)s_b^2}{n_a + n_b - 2}}}$$

# COHEN'S D EQUATION

$$d = \frac{\bar{x}_a - \bar{x}_b}{\sqrt{\frac{(n_a - 1)s_a^2 + (n_b - 1)s_b^2}{n_a + n_b - 2}}}$$

pooled variance

# COHEN'S D EQUATION

$$d = \frac{M_t - M_c}{\sqrt{\frac{(n_t - 1)s_t^2 + (n_c - 1)s_c^2}{n_t + n_c - 2}}}$$



# COHEN'S D EQUATION SIMPLIFIED

$$n_a = n_b$$

$$d = \frac{2t}{\sqrt{v}}$$

# COHEN'S D EQUATION SIMPLIFIED

$$n_a = n_b$$

$$d = \frac{2t}{\sqrt{v}}$$

$$n_a \neq n_b$$

$$d = \frac{t (n_t + n_c)}{\sqrt{v} (\sqrt{n_t} + \sqrt{n_c})}$$

# DOCUMENT DETAILS

Document produced by [Christopher Prener, Ph.D](#) for the Saint Louis University course SOC 5050: QUANTITATIVE ANALYSIS - APPLIED INFERENTIAL STATISTICS. See the [course wiki](#) and the repository [README.md](#) file for additional details.



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