

Today's Agenda

1) Recap

2) 2- test

toy examples → scratch / scipy

3) T-test

Case Study → youtube data

Giving a certain drug will increase IQ

H_0 : "drug is ineffective"
enough people

Population mean IQ = $\bar{\mu}$

$$H_0: \tilde{\mu} = \mu$$

$$H_a: \tilde{\mu} \neq \mu$$

$\hat{\mu}$ → IQ if you take drug
both of these we do not know
 $(H_a: \hat{\mu} > \mu)$

Population mean : I Q score " μ " pop std : σ

We make an assumption $H_0: \tilde{\mu} = \mu$

We get some samples / data \rightarrow those who taken drugs (100 of 200)

We need an appropriate "test statistic" from the samples

Related to the sample mean \rightarrow pop. mean

① We need the distribution T under H_0

② We need to compute p-value \rightarrow accept Vs reject
 $(P(T \geq T_{\text{obs}}))$

T: "test statistic"

$$H_0: \hat{\mu} = \mu$$

This is the sample mean

$$[\mathcal{X}_1 \ \mathcal{X}_2 \ \mathcal{X}_3 \ \dots \ \mathcal{X}_n]$$

Sample mean $\bar{m} = \frac{\mathcal{X}_1 + \mathcal{X}_2 + \dots + \mathcal{X}_n}{n}$

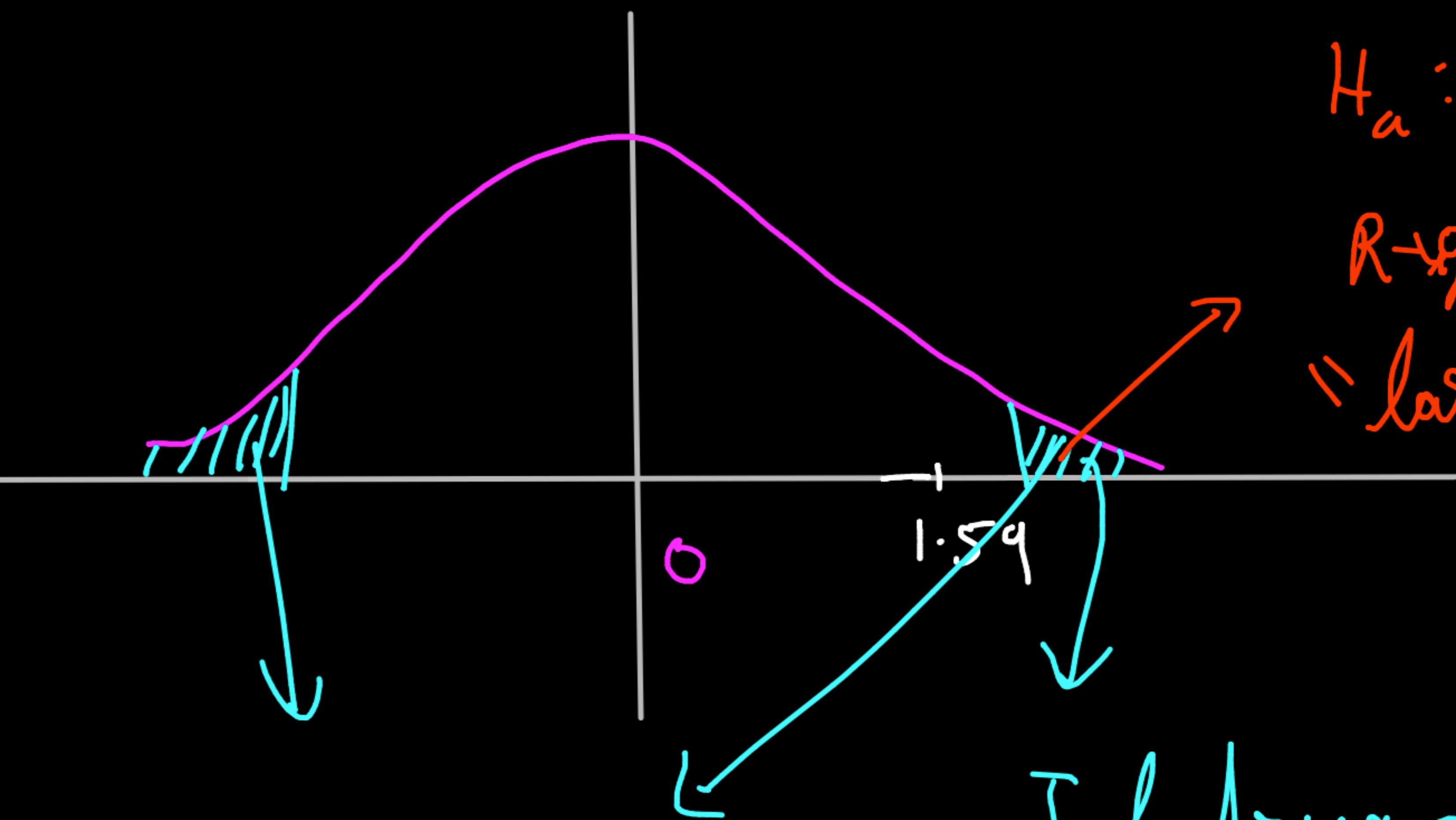
Standardization

$$T = \frac{\bar{m} - \mu}{\sigma / \sqrt{n}} \sim N(0, 1)$$

$$\sim N(\text{mean} = \mu, \text{std dev} = \frac{\sigma}{\sqrt{n}})$$

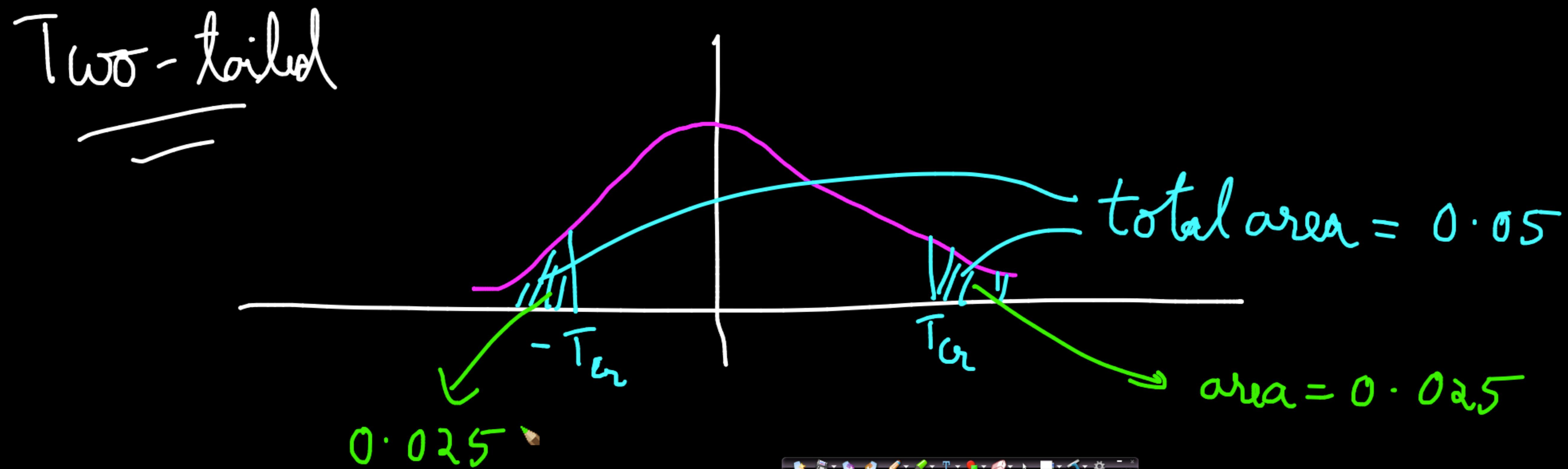
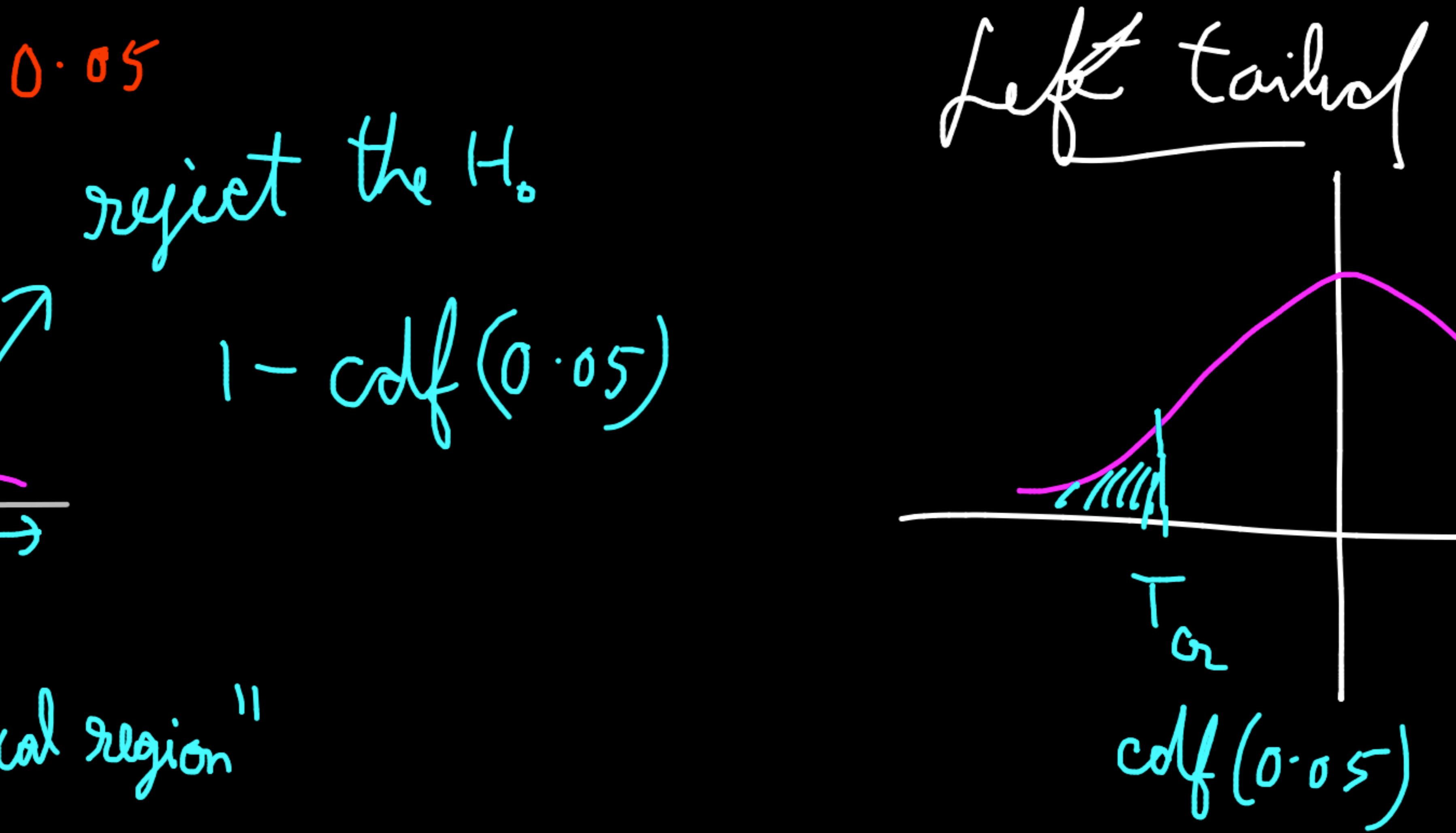
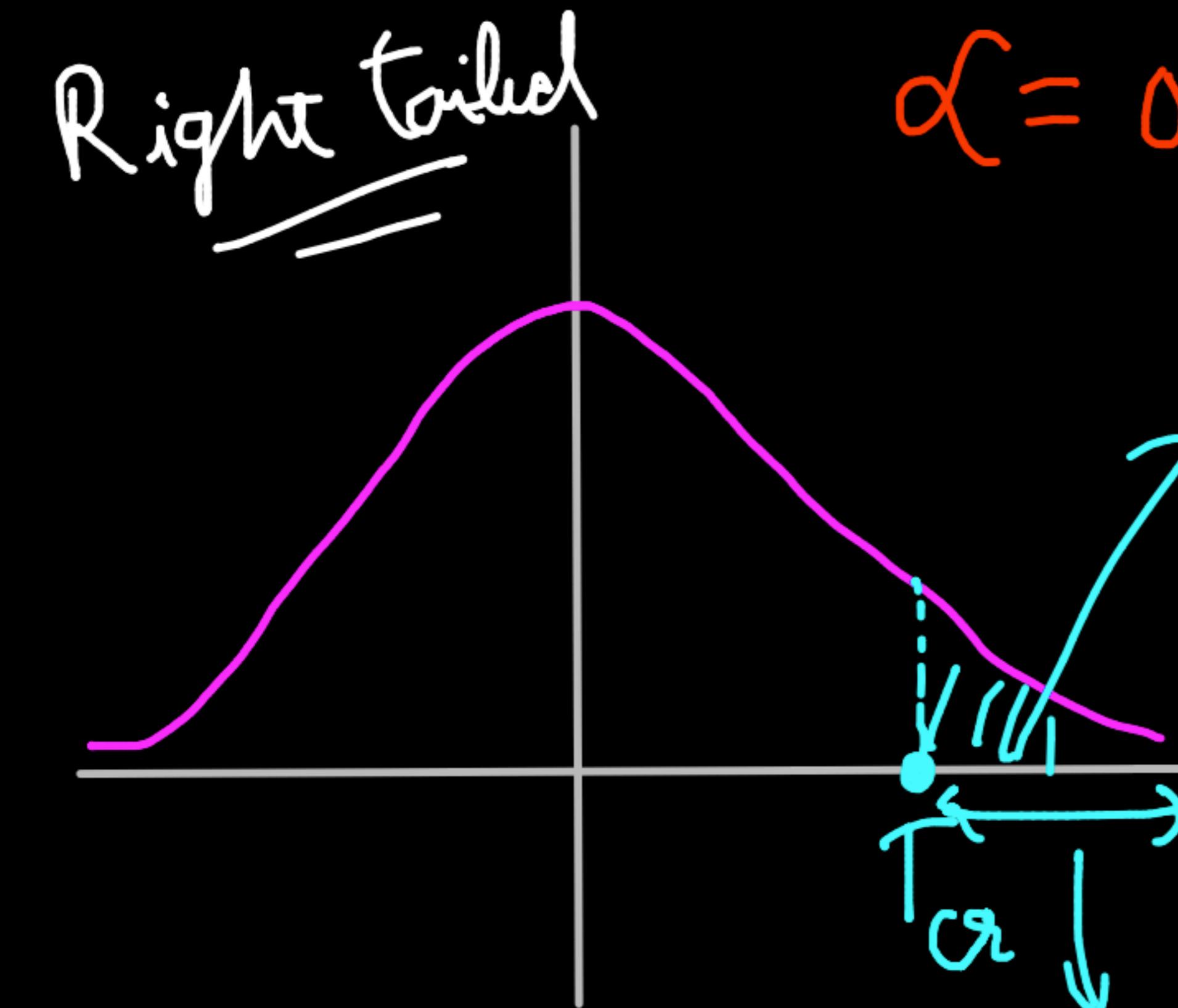
$$T = \frac{m - \mu}{\sigma / \sqrt{n}}$$

$$\sim N(0, 1)$$



$H_a: \tilde{\mu} > \mu$
Right tail
"larger"

If drug worked
observed statistic
will lie in the
right extreme



Hypothesis test : We make a statement about the population
using samples

Test statistic : Determined from the sample

Critical region : If "Observed test statistic" is in the critical region, we reject H_0

2 types of errors

→ Type I: H_0 is true, we reject → False positive
→ Type II: H_0 is false, we accept → False negative

Compare 2 drugs for recovery times: 2 sets of samples

Med 1 → recovery times $\chi_1, \chi_2, \chi_3, \dots, \chi_{n_1}$ (say 150 patients)

Med 2 → recovery times $y_1, y_2, y_3, \dots, y_{n_2}$ (say 200 people)

μ_1 → pop. mean of med 1

σ_1 → pop std

$$H_0: \mu_1 = \mu_2$$

μ_2 → pop mean for med 2

σ_2 → pop std dev

$$\bar{m}_1 = \frac{\chi_1 + \chi_2 + \dots + \chi_{n_1}}{n_1}$$

$$\bar{m}_2 = \frac{y_1 + y_2 + \dots + y_{n_2}}{n_2}$$

Options:

$$m_1 - m_2$$

$$m_1/m_2$$

$$m_1 - m_2 / S \cdot E^{\alpha}$$

1) Med |
=====

μ_1, σ_1 → population parameter
 n_1, m_1, s_1 → no. / sample mean / Sample stdn

2) Med

μ_2, σ_2 → pop
 n_2, m_2, s_2 → Sample

$$H_0: \mu_1 = \mu_2$$

$$H_a: \mu_1 \neq \mu_2$$

$$m_1 \sim N\left(\mu_1, \frac{\sigma_1^2}{\sqrt{n_1}}\right)$$

$$E[m_1] = \mu_1, \quad \text{Var}[m_1] = \frac{\sigma_1^2}{n_1}$$

$$m_2 \sim N\left(\mu_2, \frac{\sigma_2^2}{\sqrt{n_2}}\right)$$

$$E[m_2] = \mu_2, \quad \text{Var}[m_2] = \frac{\sigma_2^2}{n_2}$$

Guess: $m_1 - m_2$

$$E[m_1 - m_2] = \mu_1 - \mu_2$$

$$\text{Var}[m_1 - m_2] = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

Standardizing

$$\frac{(m_1 - m_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim$$

$$N(0, 1)$$

$$X \rightarrow \text{Var}[X] = \sigma^2$$

$$\text{Var}[aX] = a^2 \text{Var}[X] = a^2 \sigma^2$$

$$\text{Var}[-X] = (-1)^2 \text{Var}[X] = \sigma^2$$

$$\text{Var}[X+Y] = \text{Var}[X] + \text{Var}[Y] \quad (\text{if independent})$$

$$\text{Var}[m_1 - m_2] = \text{Var}[m_1] + \text{Var}[m_2]$$

Summary

- 1) One set of samples Vs population parameter
 $\mu = 100 \text{ (IO)}$
 x_1, x_2, \dots, x_n
- 2) two sets of samples
 x_1, x_2, \dots, x_{n_1} y_1, y_2, \dots, y_{n_2}

Test-Statistic :

One-sample

$$\frac{\bar{m} - \mu}{(\sigma)/\sqrt{n}}$$

$$H_0: \tilde{\mu} = \mu$$

Two-sample

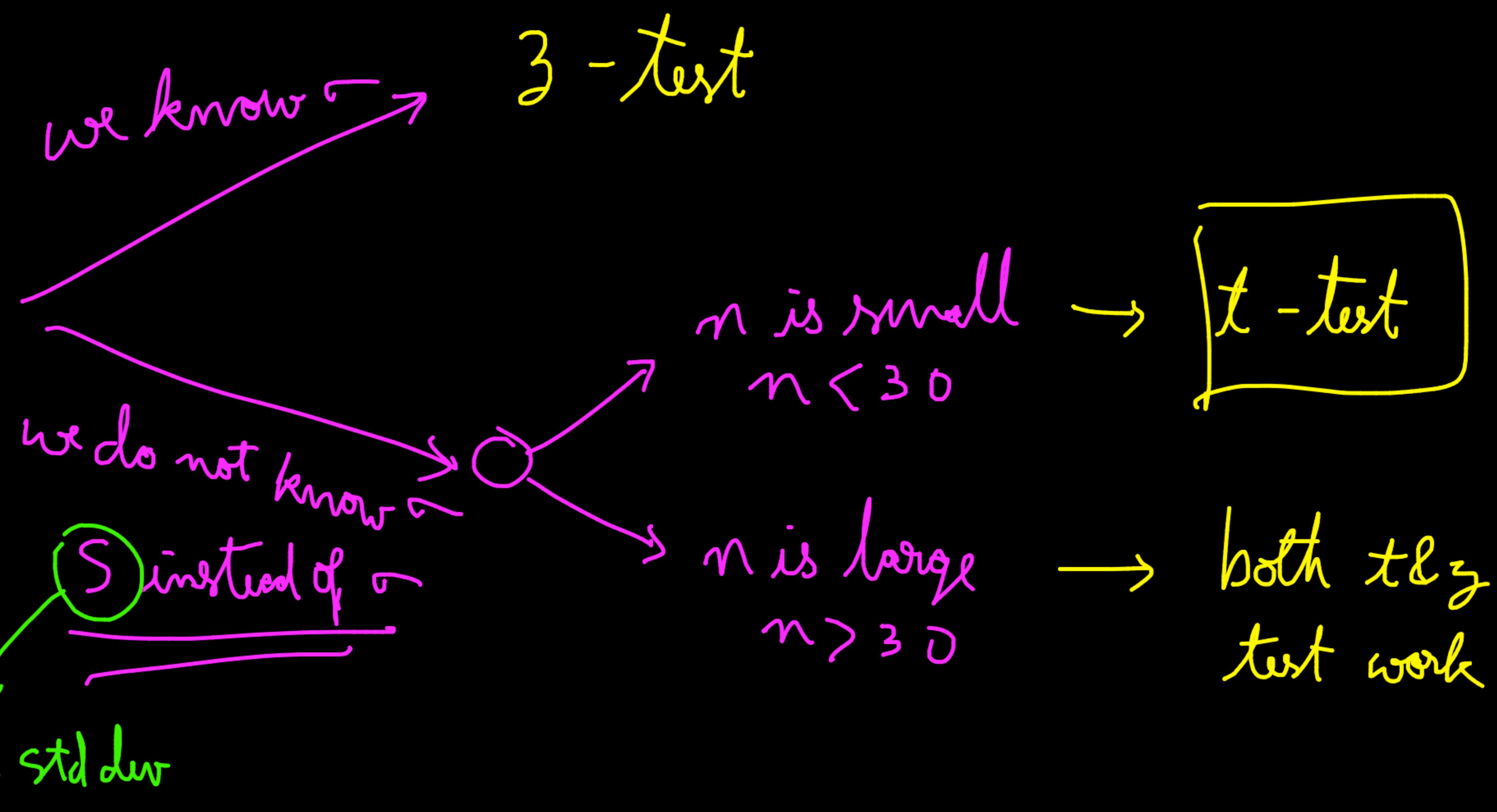
$$\frac{\bar{m}_1 - \bar{m}_2}{\sqrt{\frac{(\sum)^2}{n_1} + \frac{(\sum)^2}{n_2}}}$$

$$H_0: M_1 = M_2$$

(S) instead

$$\begin{aligned} n < 30 &\rightarrow t\text{-test} \\ n > 30 &\rightarrow z\text{-test} \end{aligned}$$

2 possibilities
regarding σ



Sample std dev

$$T = \frac{m - \mu}{\sqrt{s^2/n}}$$

no longer Gaussian

s': sample std dev

"t-distribution"

degrees of freedom $n-1$

As $n \rightarrow \infty$, $t \rightarrow$ Gaussian

$$T = \frac{m - \mu}{\sqrt{s/\sqrt{n}}}$$

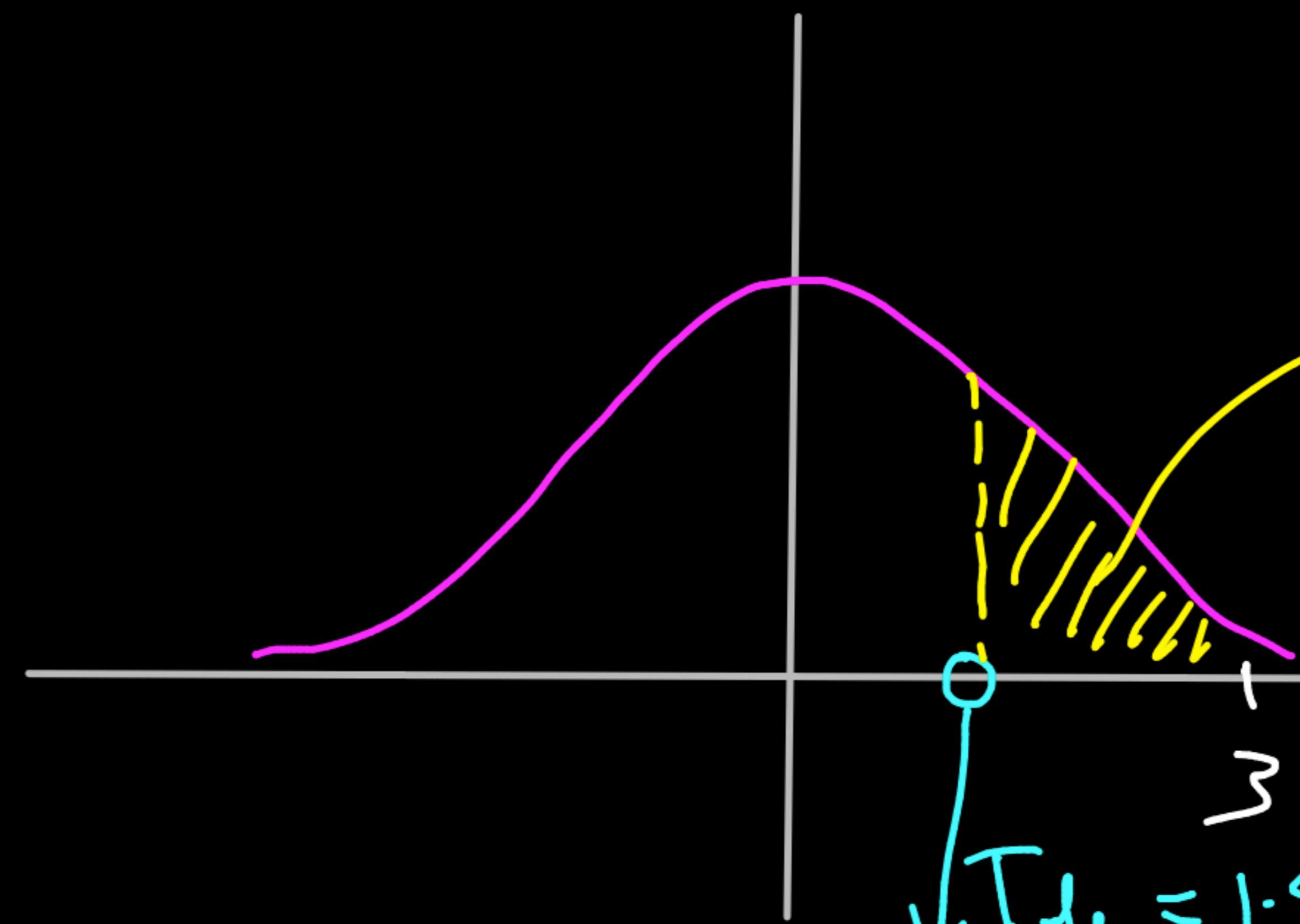
no longer Gaussian

s': sample std dev

"t-distribution"

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area is the p-value $P[T \geq T_{\text{obs}}]$

$$P[T \geq 1.5]$$

100 - 116,
↓
standard.

If test statistic is
Standard Gaussian
"Z score"