

# Quantitative Methods 2

Tutorial 3

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# Assignment 1

1. Assignment tab on LMS
2. Group registration due by 5 pm, Friday 16/8
3. Assignment 1 due by 4 pm, Friday 23/8

# Tutorial 3

1. Obtain and interpret descriptive statistics
2. Evaluate the (normal) distribution of a series
3. Hypothesis testing using
  - Parametric tests
  - Non-parametric test

# Population vs. sample

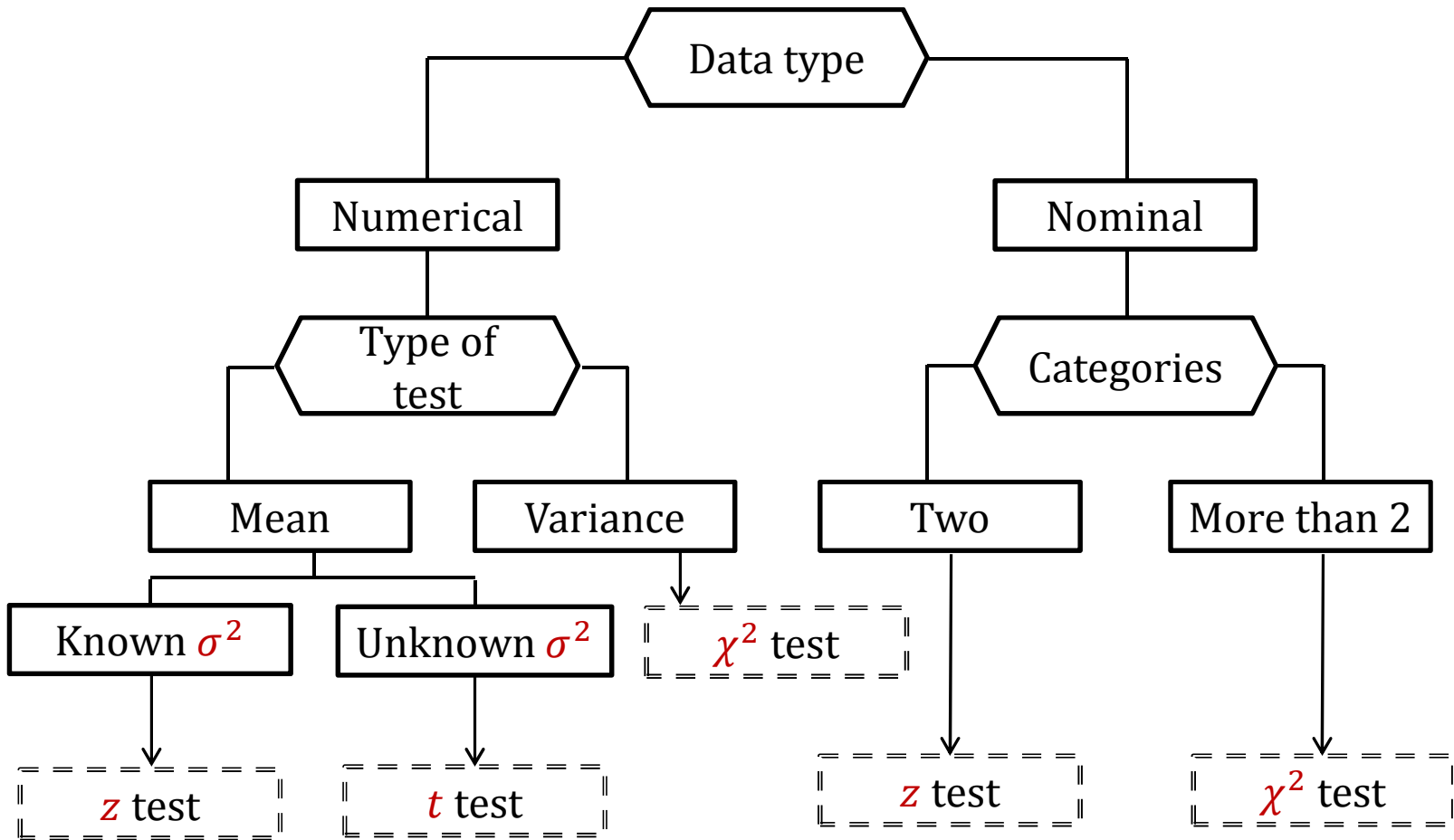
- Population vs. sample quantities

Quantities	Population parameters	Sample statistics
Mean	$\mu$	$\bar{x}$
Variance	$\sigma^2$	$s^2$
Standard deviation	$\sigma$	$s$

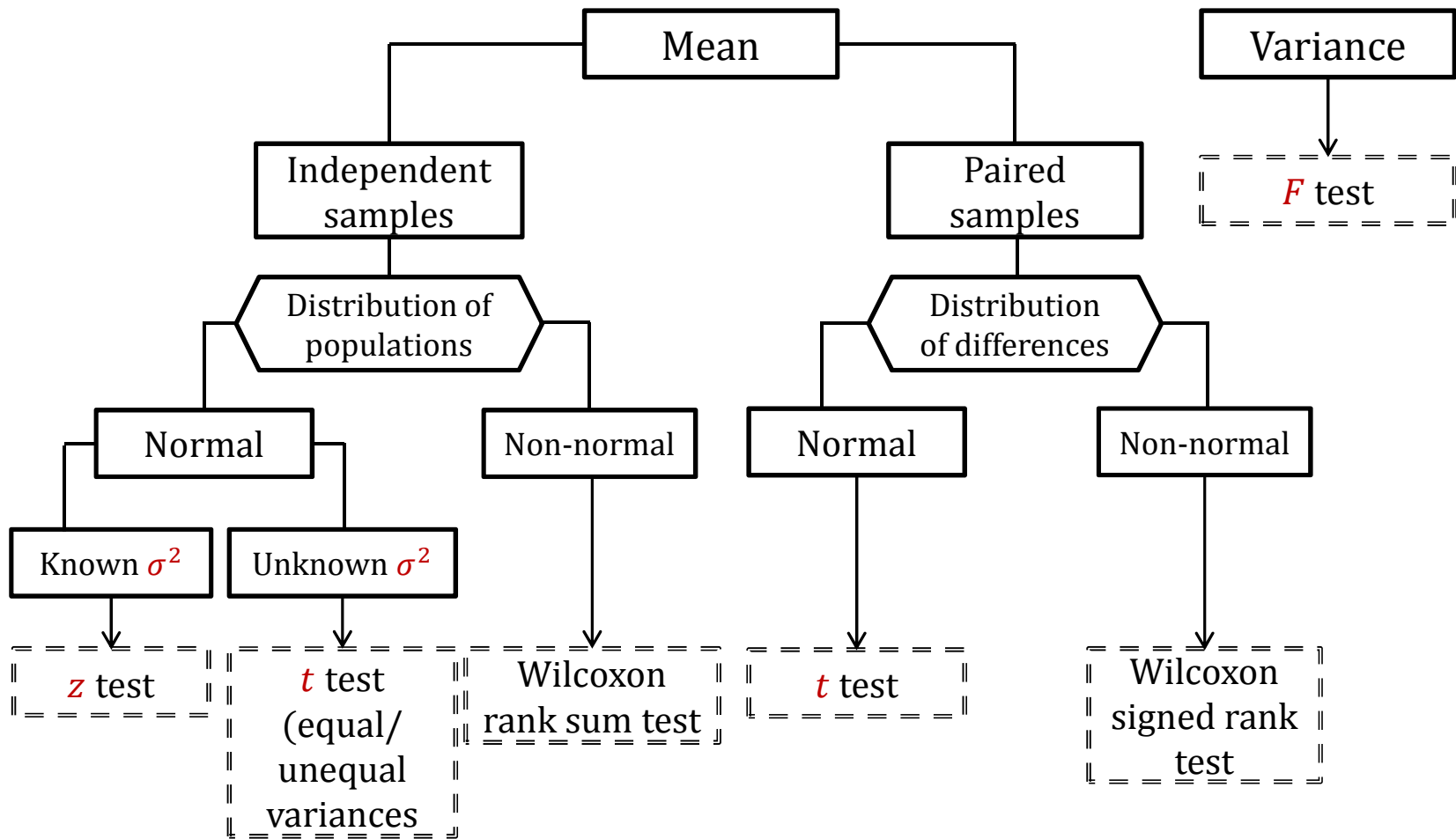
# Hypothesis testing

- Always be about population parameters (not about sample statistics)
- Test of no difference (null hypothesis  $H_0$ ) vs. difference (alternative hypothesis  $H_A$ )
  - $H_A$  never contains the equal sign
- Purpose: To see if the difference is real or due to randomness
  - Find the **probability** of obtaining a sample with the estimated statistic, assuming the population value stated in  $H_0$  is true.
  - If such a probability is small enough then  $H_0$  is false or rejected and vice versa.
  - How small is enough? It's determined in the significance level  $\alpha$

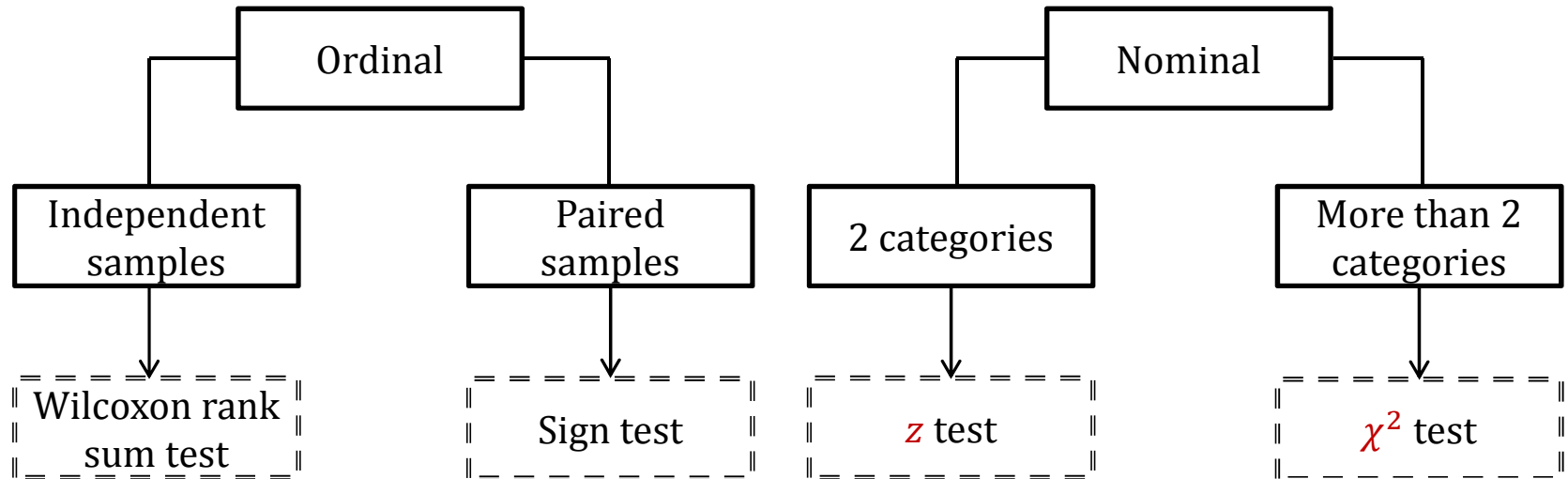
# One population



# Two populations - Numerical

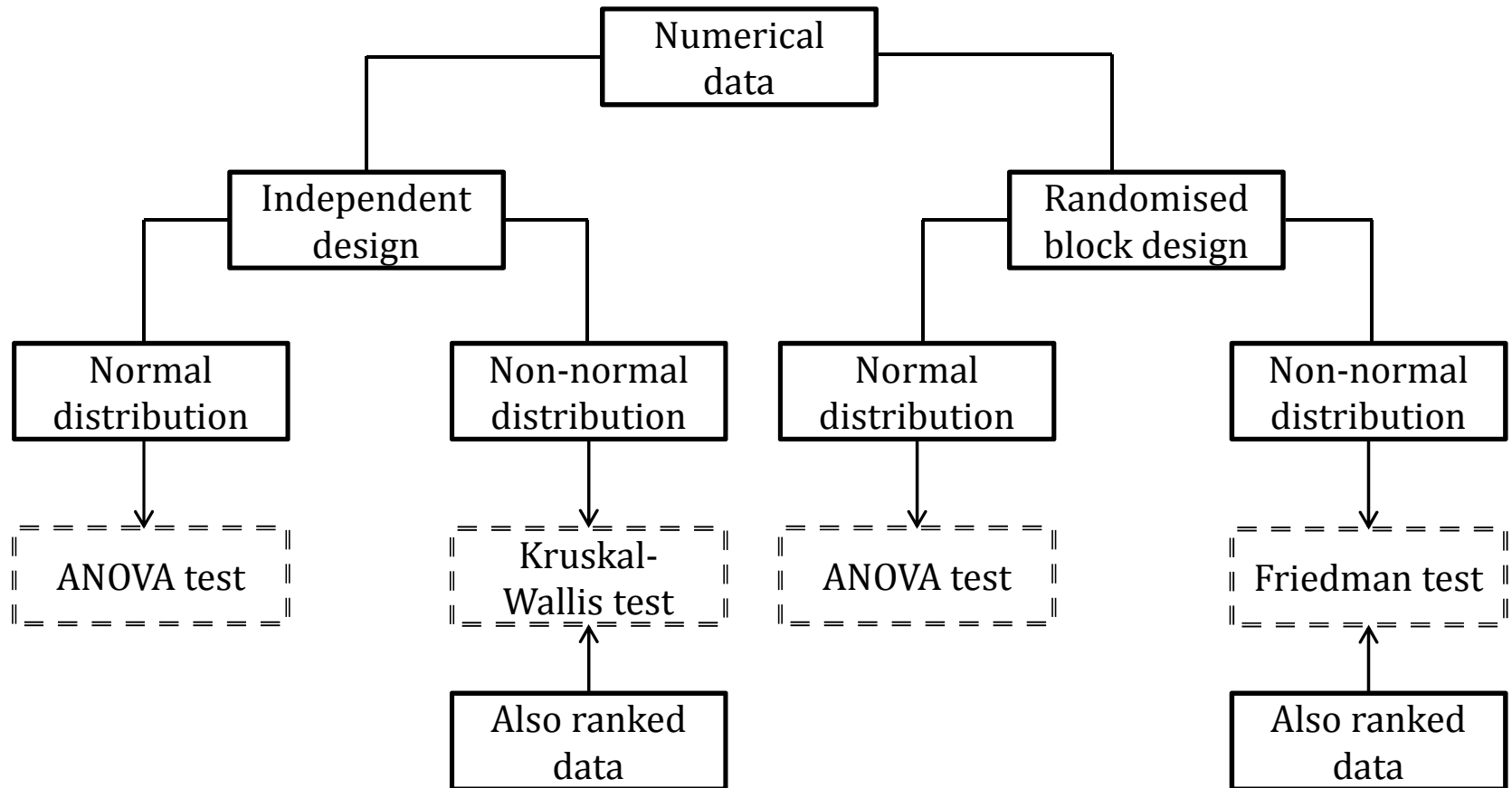


# Two populations - Ordinal and Nominal





# More than two populations



# Hypothesis testing

- Choose the appropriate test based on:
  - Levels of measurement
  - Types of variable
  - Distribution of the underlying population
  - Relevant population parameter (e.g.,  $\sigma^2$ ) known/unknown

Level of measurement	Type of variable	
	Quantitative	Qualitative
Nominal		
Ordinal		
Interval		
Ratio		

# Hypothesis testing

- Significance level  $\alpha$ 
  - Pr(Type I Error): Probability of rejecting  $H_0$  when it is true (an error). That is, the extent to which we can accept that error to happen.
  - Define the distance between the sample mean and the null hypothesis parameter
  - Conventional values of  $\alpha$ : 0.1, 0.05, 0.01
- p-value (of a test statistic)
  - The smallest  $\alpha$  that leads to a rejection of  $H_0$
  - If  $\alpha < \text{p-value}$ , we can't reject  $H_0$ , and vice versa
- The decision (i.e., to reject  $H_0$  or not) is made based on the comparison
  - Significance level vs. p-value; OR
  - Critical value vs. test statistic

# Hypothesis testing

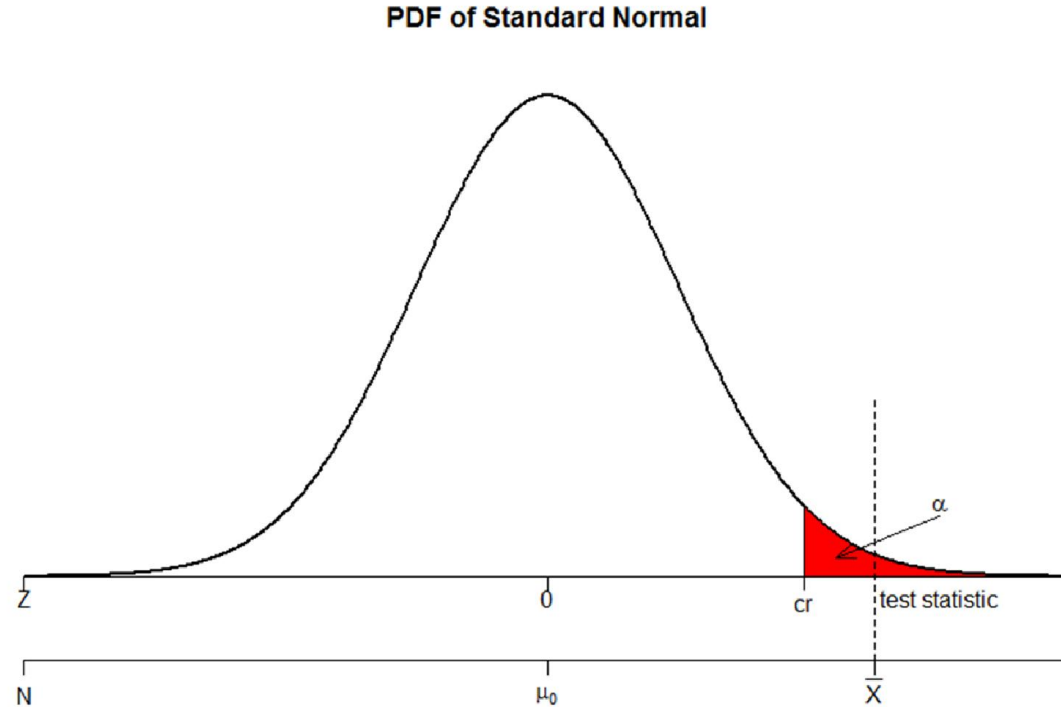
1. Set up  $H_0$  and  $H_A$
2. Determine the appropriate test, test statistic and its sampling distribution
3. Specify  $\alpha$
4. Define the decision rule using either
  - Test statistic vs. critical values, or
  - p-value vs.  $\alpha$
  - Unless asked to use both
5. Calculate the test statistic
6. Make a decision and draw a conclusion
  - Based on the comparison of either
    - Test statistic vs. critical values, or
    - p-value vs.  $\alpha$
    - Whatever use, make it consistent with step 4
  - Decision: Reject or not able to reject  $H_0$
  - Conclusion: Statement in relation to the population values.

# Normality test

- Visual assessment
- Descriptive measures: **relative comparisons**
  - Mean vs. median
  - Skewness
    - $\widehat{SK}$  vs. 0
    - $|\widehat{SK}|$  vs.  $2s_{\widehat{SK}}$
  - Kurtosis
    - $\widehat{K}$  vs. 3
    - $|\widehat{K} - 3|$  vs.  $2s_{\widehat{K}}$
- Hypothesis test: Jacque-Bera

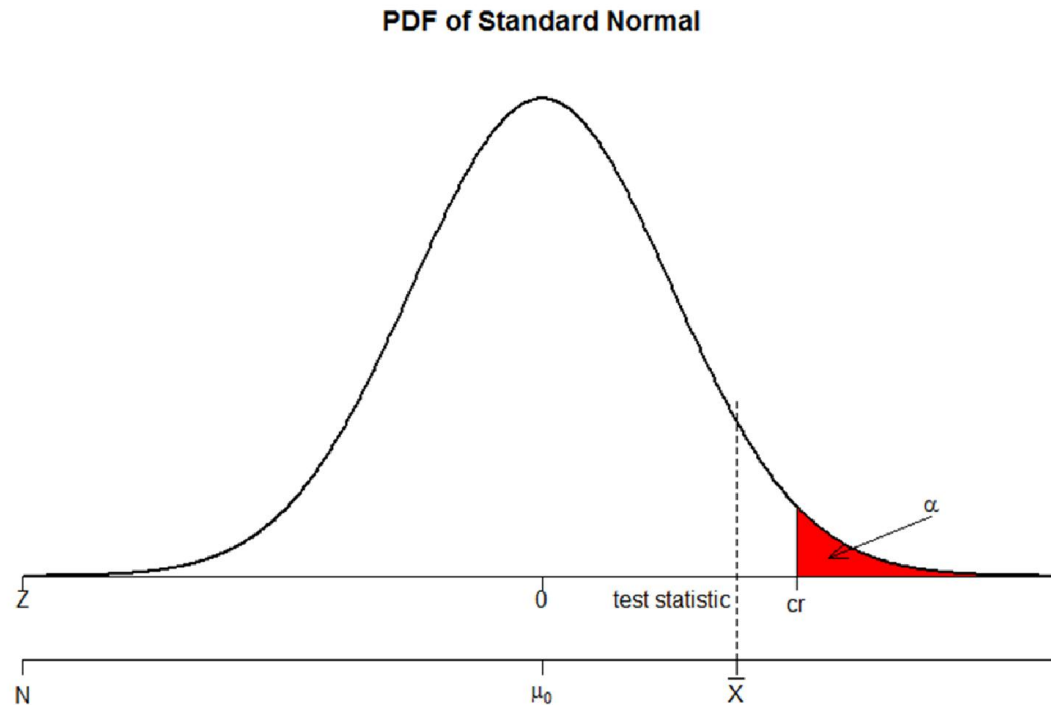
# One-sided hypothesis

- $H_0: \mu = \mu_0$
- $H_A: \mu > \mu_0$
- Reject  $H_0$ ?
  - Yes



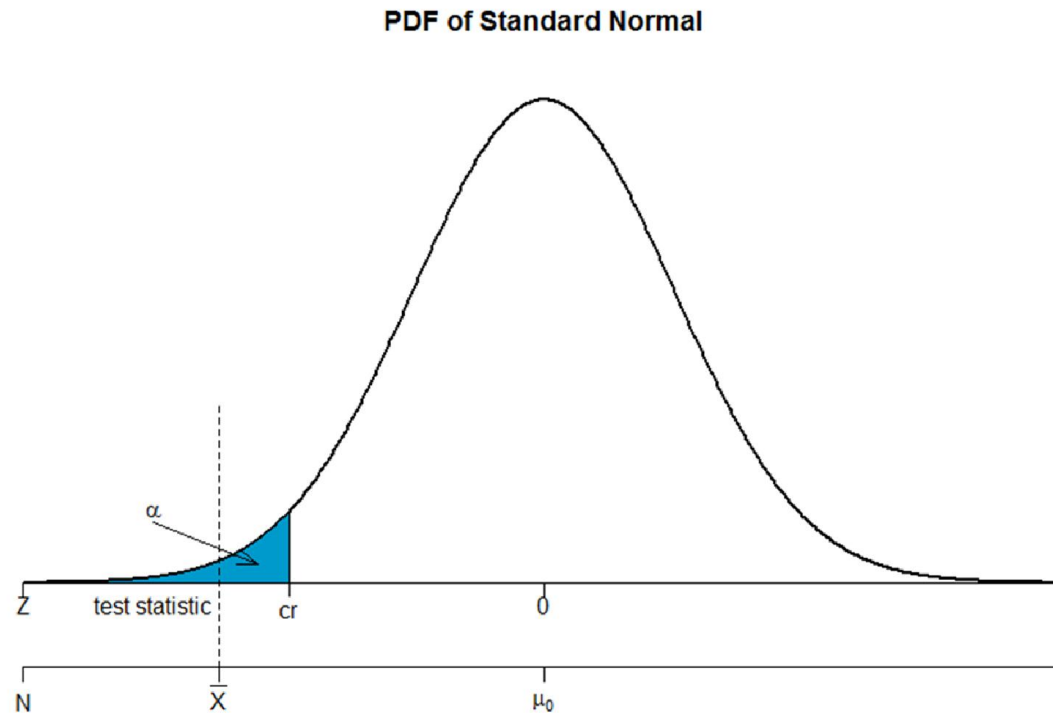
# One-sided hypothesis

- $H_0: \mu = \mu_0$
- $H_A: \mu > \mu_0$
- Reject  $H_0$ ?
  - No



# One-sided hypothesis

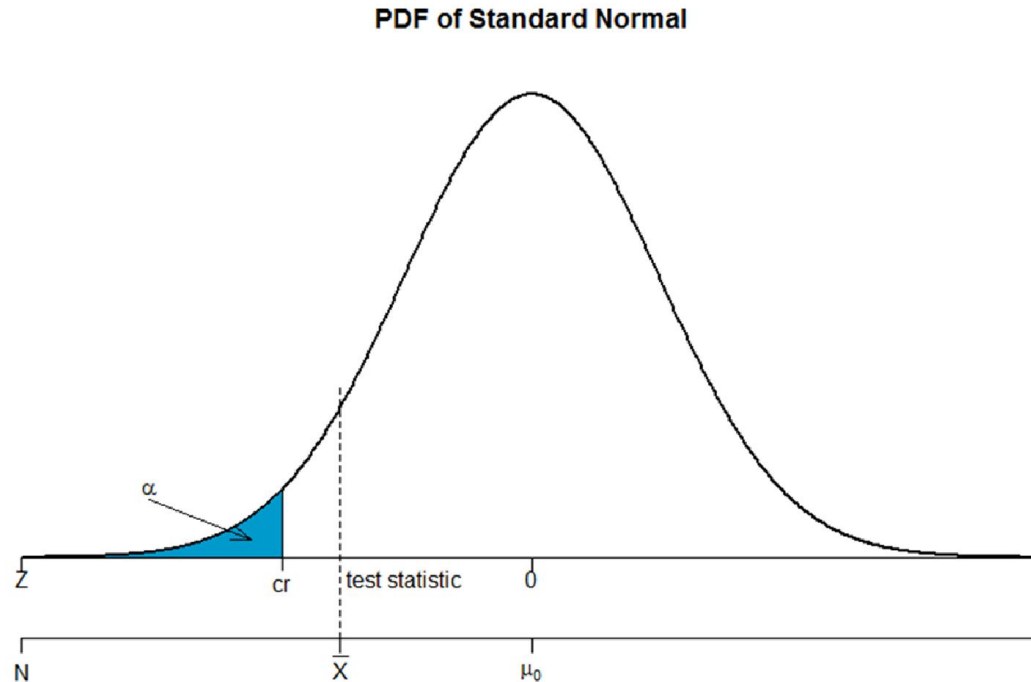
- $H_0: \mu = \mu_0$
- $H_A: \mu < \mu_0$
- Reject  $H_0$ ?
  - Yes





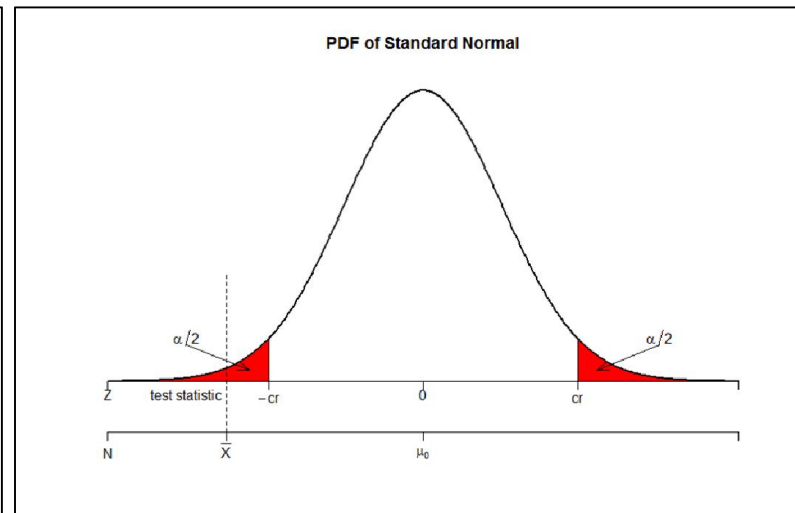
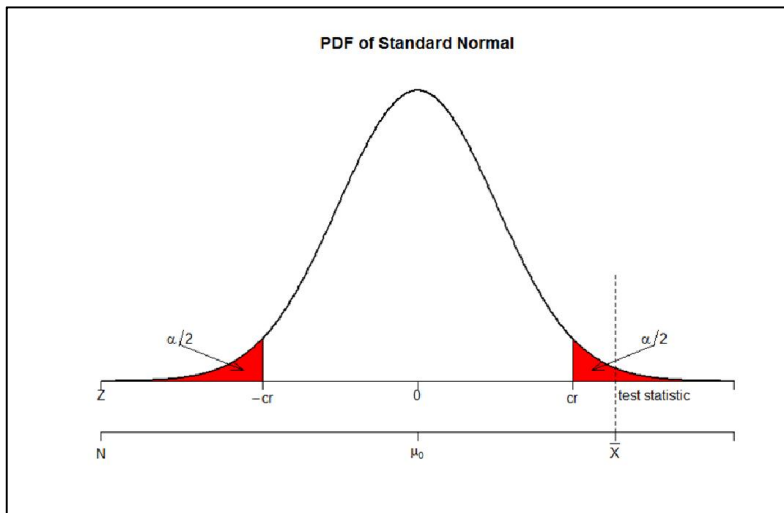
# One-sided hypothesis

- $H_0: \mu = \mu_0$
- $H_A: \mu < \mu_0$
- Reject  $H_0$ ?
  - No



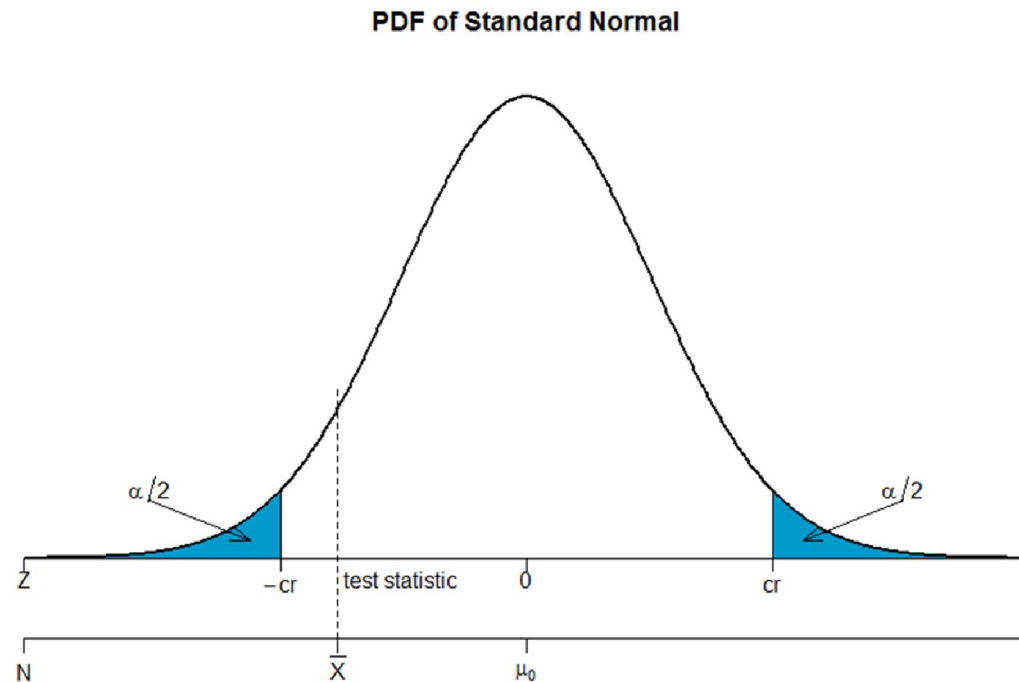
# Two-sided hypothesis

- $H_0: \mu = \mu_0; H_A: \mu \neq \mu_0$
- Reject  $H_0$ ?
  - Yes



# Two-sided hypothesis

- $H_0: \mu = \mu_0$
- $H_A: \mu \neq \mu_0$
- Reject  $H_0$ ?
  - No



# Critical values

1. Exercise 3: Two sided test:

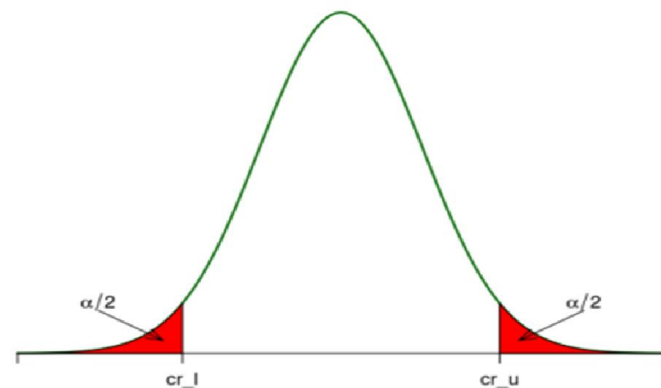
- $\alpha = 0.05$
- $H_0: \mu = 0.25$  ;  $H_A: \mu \neq 0.25$

$$t_{cr\_u} = t_{\alpha/2, n-1} = t_{0.025, 39} \text{ [EViews: scalar tcr\_u=@qtdist(0.975,39)]}$$

$$t_{cr\_l} = t_{1-\alpha/2, n-1} = t_{0.975, 39} \text{ [EViews: scalar tcr\_l=@qtdist(0.025,39)]}$$

Use the  $t$  distribution table: Find  $t_{cr\_u}$  and apply the symmetry:  $t_{cr\_l} = -t_{cr\_u}$

Degrees of freedom	$t_{0.100}$	$t_{0.050}$	$t_{0.025}$	$t_{0.010}$	$t_{0.005}$
29	1.311	1.699	2.045	2.462	2.756
30	1.310	1.697	2.042	2.457	2.750
35	1.306	1.690	2.030	2.438	2.724
40	1.303	1.684	2.021	2.423	2.704
45	1.301	1.679	2.014	2.412	2.690



# Critical values

## 2. Exercise 3: One-sided test

- Right tail test:  $H_0: \mu = 0.25$  ;  $H_A: \mu > 0.25$

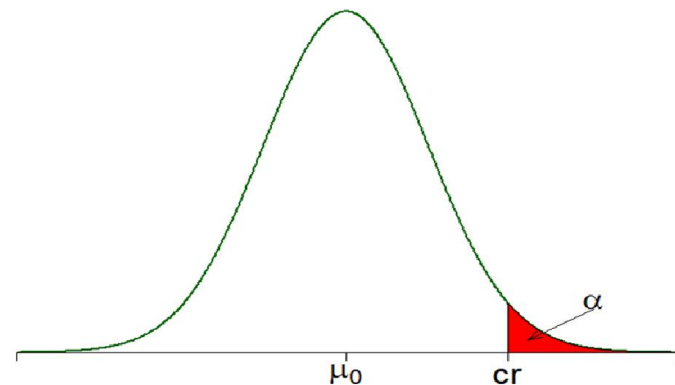
$$t_{cr\_rt} = t_{\alpha, n-1} = t_{0.01, 39} \text{ [EViews: scalar tcr\_rt=@qtdist(0.99,39)]}$$

- Left tail test:  $H_0: \mu = 0.25$  ;  $H_A: \mu < 0.25$

$$t_{cr\_lt} = t_{1-\alpha, n-1} = t_{0.99, 39} \text{ [EViews: scalar tcr\_lt=@qtdist(0.01,39)]}$$

Use the  $t$  distribution table:  $t_{cr\_lt} = -t_{cr\_rt}$

Degrees of freedom	$t_{0.100}$	$t_{0.050}$	$t_{0.025}$	$t_{0.010}$	$t_{0.005}$
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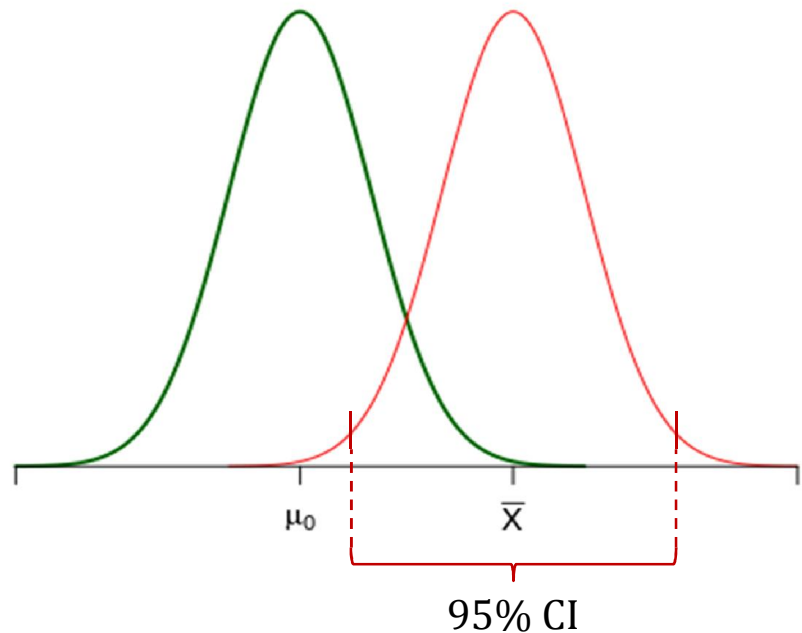
# Confidence interval

- Not a probability
- Always agree with the p-value of the statistic with regard to statistical significance
- Recall significance level defines the distance between the sample mean and the null hypothesis parameter
- The confidence level defines the distance for how close the confidence limits are to sample mean

# Confidence interval

- Reject  $H_0$  at 5% sig. level
- 95% CI doesn't contain  $\mu_0$

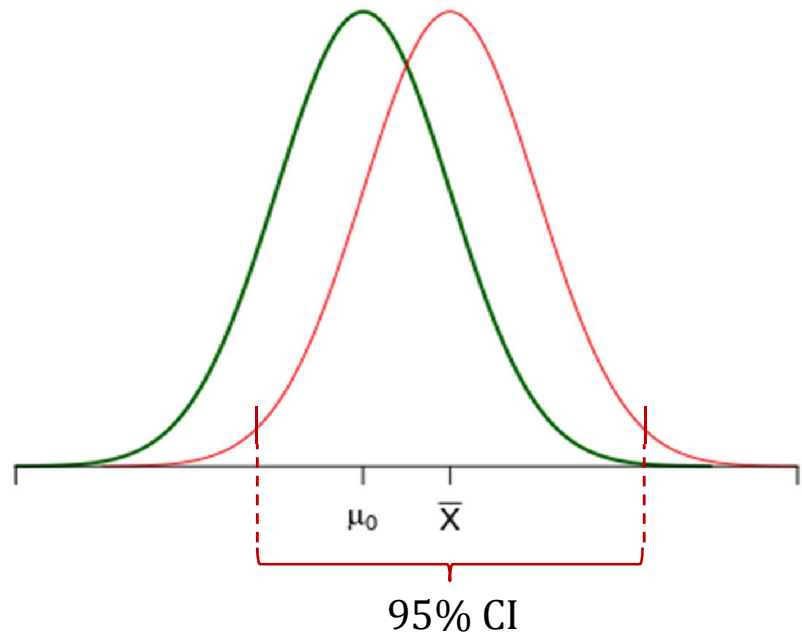
CI and statistical significance



# Confidence interval

- Not to reject  $H_0$  at 5% sig. level
- 95% CI contains  $\mu_0$

CI and statistical significance





# EViews command

- Create a text object to save and modify commands.  
Object → New Object → Text
- Examples of finding critical values associated with significance levels, and p-values associated with test statistics
  - Standard normal distribution (Z test)
  - t distribution (t test)
  - Chi-squared distribution

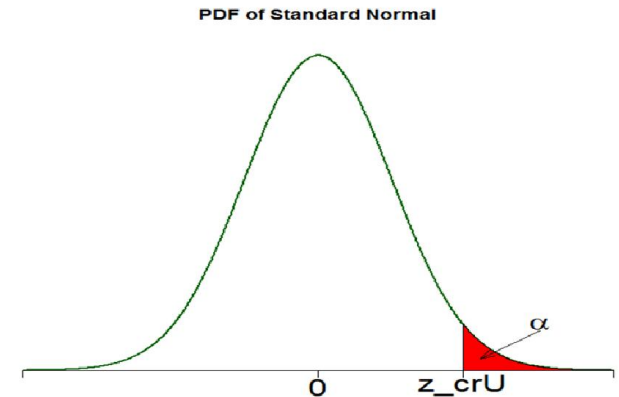
# Standard normal distribution

## 1. Critical value

### a/ One-sided, upper test

$$H_0: \mu = \mu_0; H_A: \mu > \mu_0$$

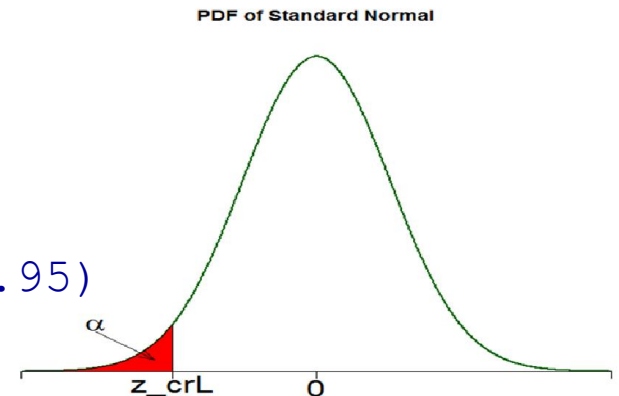
- $\alpha = 0.05$
- Reject  $H_0$  if z statistic  $> z_{\text{critical}}$
- Find  $z_{\text{critical}}$ : `scalar z_crU=@qnorm(0.95)`



### b/ One-sided, lower test

$$H_0: \mu = \mu_0; H_A: \mu < \mu_0$$

- $\alpha = 0.05$
- Reject  $H_0$  if z statistic  $< z_{\text{critical}}$
- Find  $z_{\text{critical}}$ : `scalar z_crL = -@qnorm(0.95)`



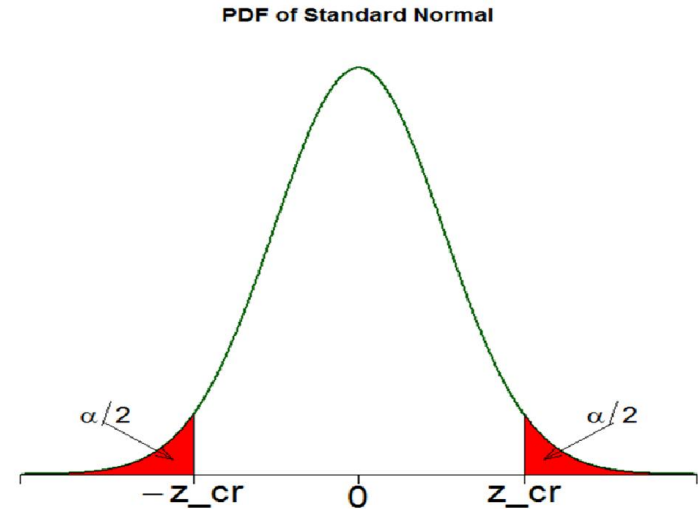
# Standard normal distribution

## 1. Critical value (cont.)

### c/ Two-sided test

$$H_0: \mu = \mu_0; H_A: \mu \neq \mu_0$$

- $\alpha = 0.05$
- Reject  $H_0$  if:  $z \text{ statistic} > z_{\text{critical}}$   
OR  $z \text{ statistic} < -z_{\text{critical}}$
- Find  $z_{\text{critical}}$ :  
`scalar z_cr = @qnorm(0.975)`
- Your turn:
  - Find  $z_{\text{critical}}$  for  $\alpha = 0.1$  and  $\alpha = 0.01$  for one- and two-sided tests
  - Compare values between EViews and standard normal distribution table (they should be equal)



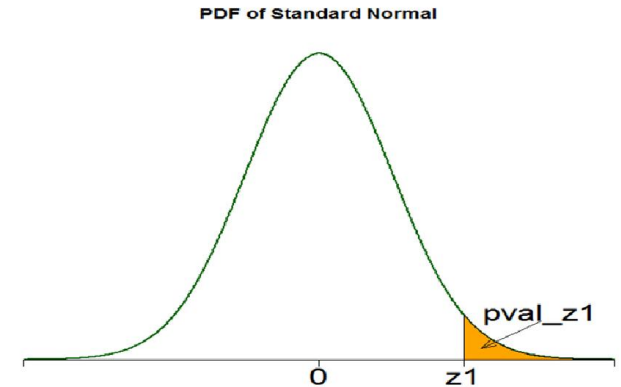
# Standard normal distribution

## 2. p-value of $z$ statistic

### a/ One-sided test

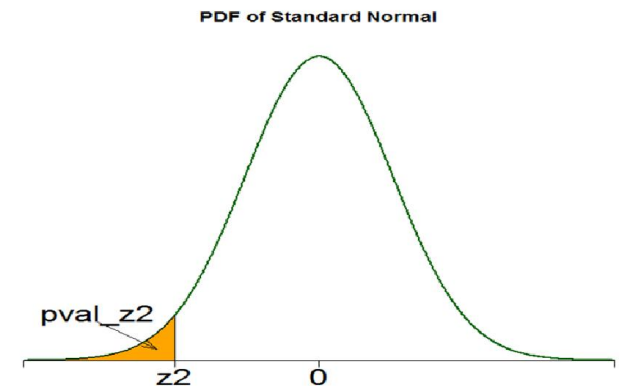
- Suppose  $z$  statistic = 1.68

```
scalar pval_z1 = 1-@cnorm(1.68)
```



- Suppose  $z$  statistic = -1.68

```
scalar pval_z2 = @cnorm(-1.68)
```



- In general:

```
p-value(z) = 1 - p-value(-z)
```

# Standard normal distribution

## 2. p-value of **z statistic** (cont.)

### b/ Two-sided test

- Suppose **z statistic** = 1.68

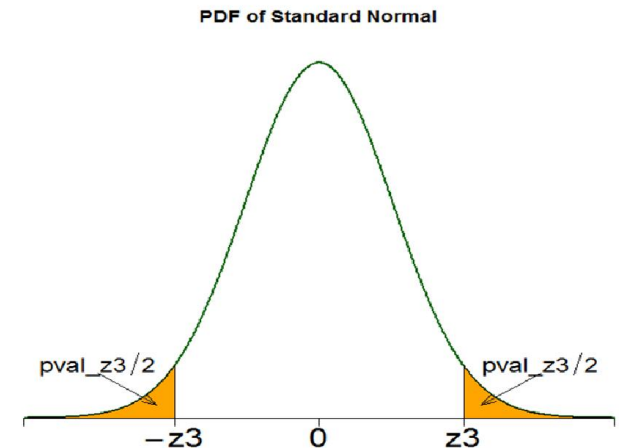
```
scalar pval_z3=2*(1-@cnorm(1.68))
```

- Suppose **z statistic** = -1.68

```
scalar pval_z3 = 2*@cnorm(-1.68)
```

- Your turn:

- Find these p-values.
- Compare EViews and standard normal distribution table outcomes
- Also, compare one-sided and two-sided p-values



# t distribution

## 1. Critical value:

### a/ One-sided, upper test

$$H_0: \mu = \mu_0; H_A: \mu > \mu_0$$

- $\alpha = 0.05$ , suppose degrees of freedom (d.f) = 49
- Reject  $H_0$  if  $t$  statistic  $> t_{\text{critical}}$
- Find  $t_{\text{critical}}$ :

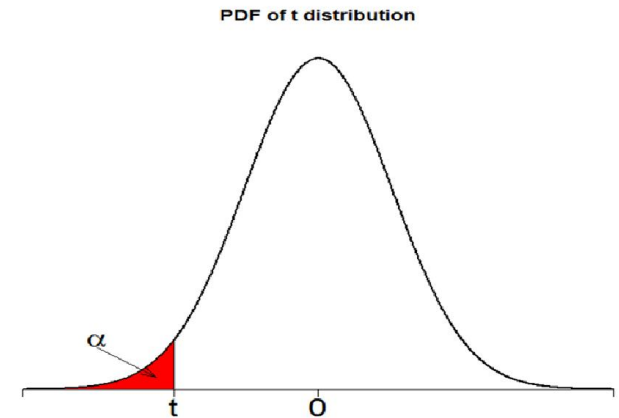
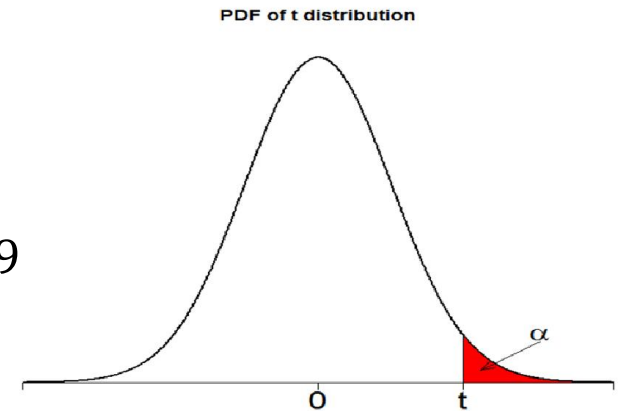
```
scalar t_crU = @qtdist(0.95,49)
```

### b/ One-sided, lower test

$$H_0: \mu = \mu_0; H_A: \mu < \mu_0$$

- $\alpha = 0.05$ , d.f = 49
- Reject  $H_0$  if  $t$  statistic  $< t_{\text{critical}}$
- Find  $t_{\text{critical}}$ :

```
scalar t_crL = -@qtdist(0.95,49)
```



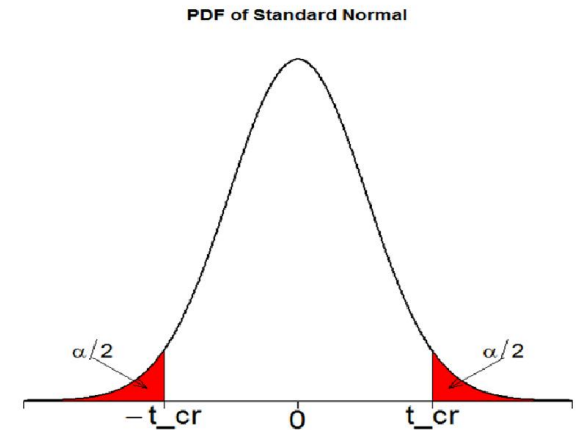
# t distribution

## 1. Critical value (cont.)

### c/ Two-sided test

$$H_0: \mu = \mu_0; H_A: \mu \neq \mu_0$$

- $\alpha = 0.05$ , d.f.=49
- Reject  $H_0$  if:  $t \text{ statistic} > t_{\text{critical}}$   
OR  $t \text{ statistic} < -t_{\text{critical}}$
- Find  $t_{\text{critical}}$ : `scalar t_cr = @qtdist(0.975,49)`
- Your turn:
  - Find  $t_{\text{critical}}$  for  $\alpha = 0.1$  and  $\alpha = 0.01$  for one- and two-sided tests. Also, vary d.f and observe the difference.
  - Compare values between EViews and t distribution table (they should be equal)



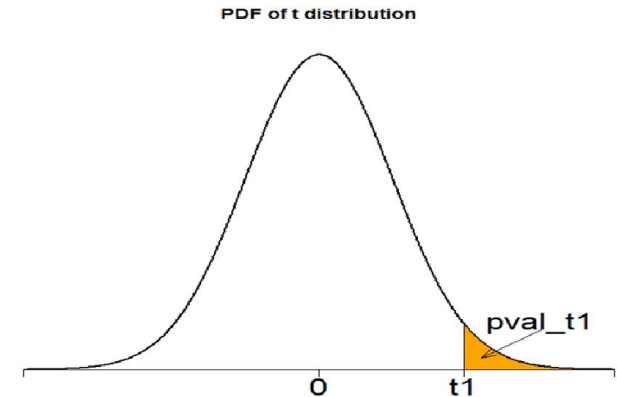
# t distribution

2. p-value of *t statistic* (still, d.f=49)

a/ One-sided test

- Suppose *t statistic* = 1.68

```
scalar pval_t1 = 1-@ctdist(1.68,49)
```

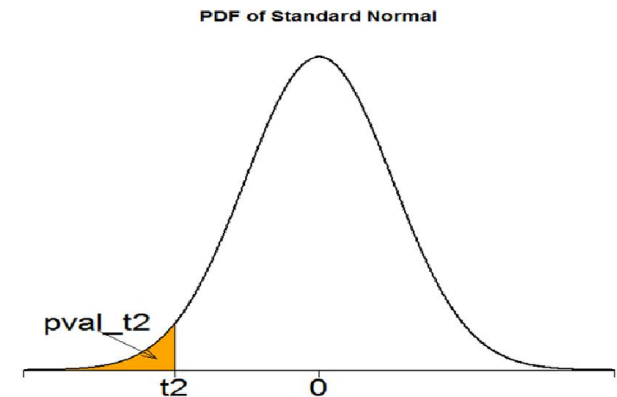


- Suppose *t statistic* = -1.68

```
scalar pval_t2 = @ctdist(-1.68,49)
```

- In general:

```
p-value(t) = 1 - p-value(-t)
```





# t distribution

## 2. p-value of *t* statistic (cont.)

### b/ Two-sided test

- Suppose *t* statistic = 1.68

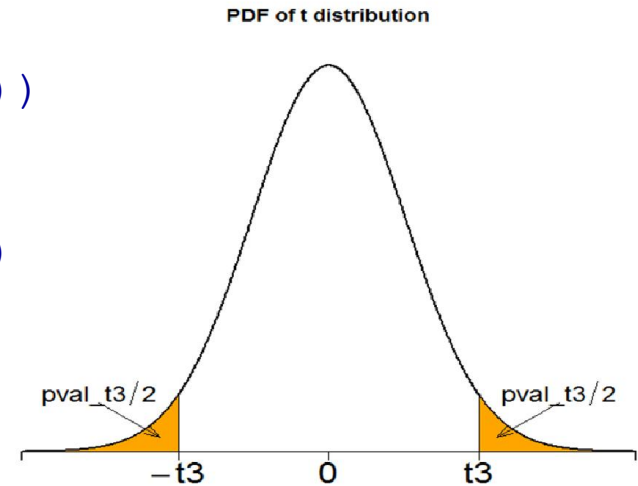
```
scalar pval_t3=2*(1-@ctdist(1.68,49))
```

- Suppose *t* statistic = -1.68

```
scalar pval_t3 = 2*@ctdist(-1.68,49)
```

- Your turn:

- Find these p-values.
- Compare EViews and t distribution table outcomes
- Also, compare one-sided and two-sided p-values



# Chi-squared distribution

## 1. Critical value

- $\chi^2$  is always positive (by definition)

$$H_0: \chi^2 = \chi_0^2; H_A: \chi^2 > \chi_0^2$$

- $\alpha = 0.05$ , degrees of freedom (d.f) = 8

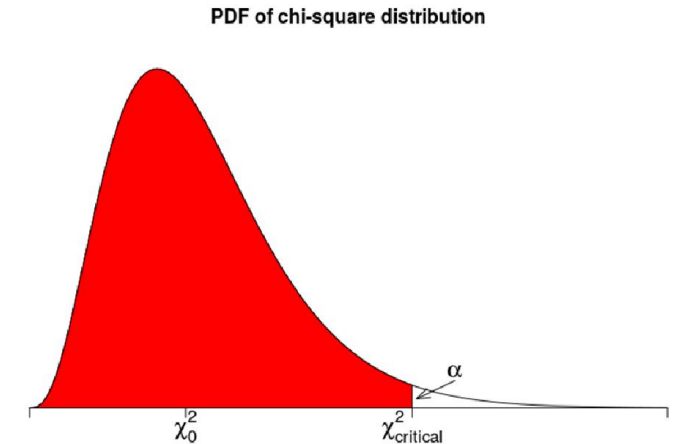
- Reject  $H_0$  if  $\chi^2 > \chi_{\text{critical}}^2$

- Find  $\chi_{\text{critical}}^2$ :

```
scalar cr_chi = @qchisq(0.95, 8)
```

- Your turn:

- Find the critical value. Vary  $\alpha$  and d.f and observe the difference
- What if  $H_A: \chi^2 < \chi_0^2$ ? Reject  $H_0$  if  $\chi^2 < \chi_{\text{critical}}^2$
- Compare EViews and chi-squared distribution table outcomes



# Chi-squared distribution

## 1. p-value of chi-squared statistic

- Suppose  $\chi^2$  statistic = 20, d.f = 8

- Find p-value  $\chi^2_{\text{critical}}$ :

```
scalar pval_chi = 1-@cchisq(20,8)
```

- Notice the difference of command compared with z and t distributions

- Your turn:

- Find the p value. Vary  $\chi^2$  and d.f and observe the difference
- Compare EViews and chi-squared distribution table outcomes

