

# PBIO 504

## Introductory Biostatistics

### **Analysis of Variance (ANOVA)**

Textbook chapter 12

# Hypothesis Testing t-test

- When we want to compare means from two different groups, we use the Student t test (also called the 2 sample t test).
- Null hypothesis:  $H_0: \mu_1 = \mu_2$ .
- The test determines whether the difference between the observed means  $\bar{X}_1 - \bar{X}_2$  is too large to have occurred by chance.
- How large is large?

$$T = | \bar{X}_1 - \bar{X}_2 | / S_p, \text{ and } P_{H_0} [|T| \geq t] = 0.05$$

When n is large,  $df = n_1 + n_2 - 2$ ,  $P_{H_0} [|T| \geq z] = 0.05$

# t Tests for more than Two Groups

- Suppose we had three different groups
- We could test the following pairs of groups using a t-test for each pair:
  - *$mean_1$  vs  $mean_2$*
  - *$mean_2$  vs  $mean_3$*
  - *$mean_1$  vs  $mean_3$*

» However

# Recall the Alpha Level

- We reject the null hypothesis if the p-value from the t test is less than  $\alpha$ , the level of chance (probability) that we are willing to accept for wrongly rejecting the null hypothesis.
- If  $\alpha$  is set at 0.05, it will have this value for each and every t-test we perform.
- So if there are 3 groups being compared in our study, there are 3 t tests, each carrying its own  $\alpha$  of 0.05, and so the study wide  $\alpha$  may become as large as  $3 \times 0.05 = 0.15$
- ***Are we willing to accept such a level? No, see later.***

# ANOVA

- ANOVA will address this problem by giving us a single p-value to test the hypothesis:

$$H_0: \text{mean}_1 = \text{mean}_2 = \text{mean}_3 = \dots = \text{mean}_k$$

*$H_A$ : at least one of these means is different from the others*

where **k** is the number of groups.

# Some Features of ANOVA

- **ANOVA** (analysis of variance) gets its name from its ability to determine whether the variability within groups is greater or lesser than between groups. Clearly the latter outcome would be desirable.
- ANOVA incorporates multiple variables: the outcome variable is the dependent variable (which must be a continuous variable), and the explanatory variables are the independent variables. This total set of variables is called the model.
- If there are just 2 groups, the ANOVA p-value will be the same as the p-value of the 2-sample t test if there are no other variables in the model.

# The ANOVA Model

- The general form:  
***dependent variable (y) =***  
***=independent variables ( $x_1 + x_2 + x_3 + \dots + x_k$ )***
- An example would be:  
***blood pressure = agegroup***
- Blood pressure is being modeled as a function of age group, where group 1: 20-40 years old, group 2: 41-60 years old, and group 3: 61+ years old. This is an example of a one-way ANOVA because there is only one independent variable agegroup. Var agegroup has 3 levels.

# One- Way ANOVA Table

Source	df	SS	MS	F	P
Model (m) ( <i>between groups</i> )	$df_m = K-1$	$SS_m$	$MS_m = SS_m / df_m$	$MS_m / MS_e$	P-value from F distribution
Error (e) ( <i>within groups</i> )	$df_e = N-K$	$SS_e$	$MS_e = SS_e / df_e$		
Total ( <i>overall</i> )	$df = N-1$	$SS_m + SS_e$			

Note: N is the total sample size and K is the number of groups of the indep var. (in the previous ex age had 3 groups)



# ANOVA Table

Source	df	SS	MS	F	P
Model (m) ( <i>between groups</i> )	K-1	$SS_m$	$SS_m / df_m$	$MS_m / MS_e$	from F distribution with
Error (e) ( <i>within groups</i> )	N-K	$SS_e$	$SS_e / df_e$		
Total ( <i>overall</i> )	N-1	$SS_m + SS_e$			

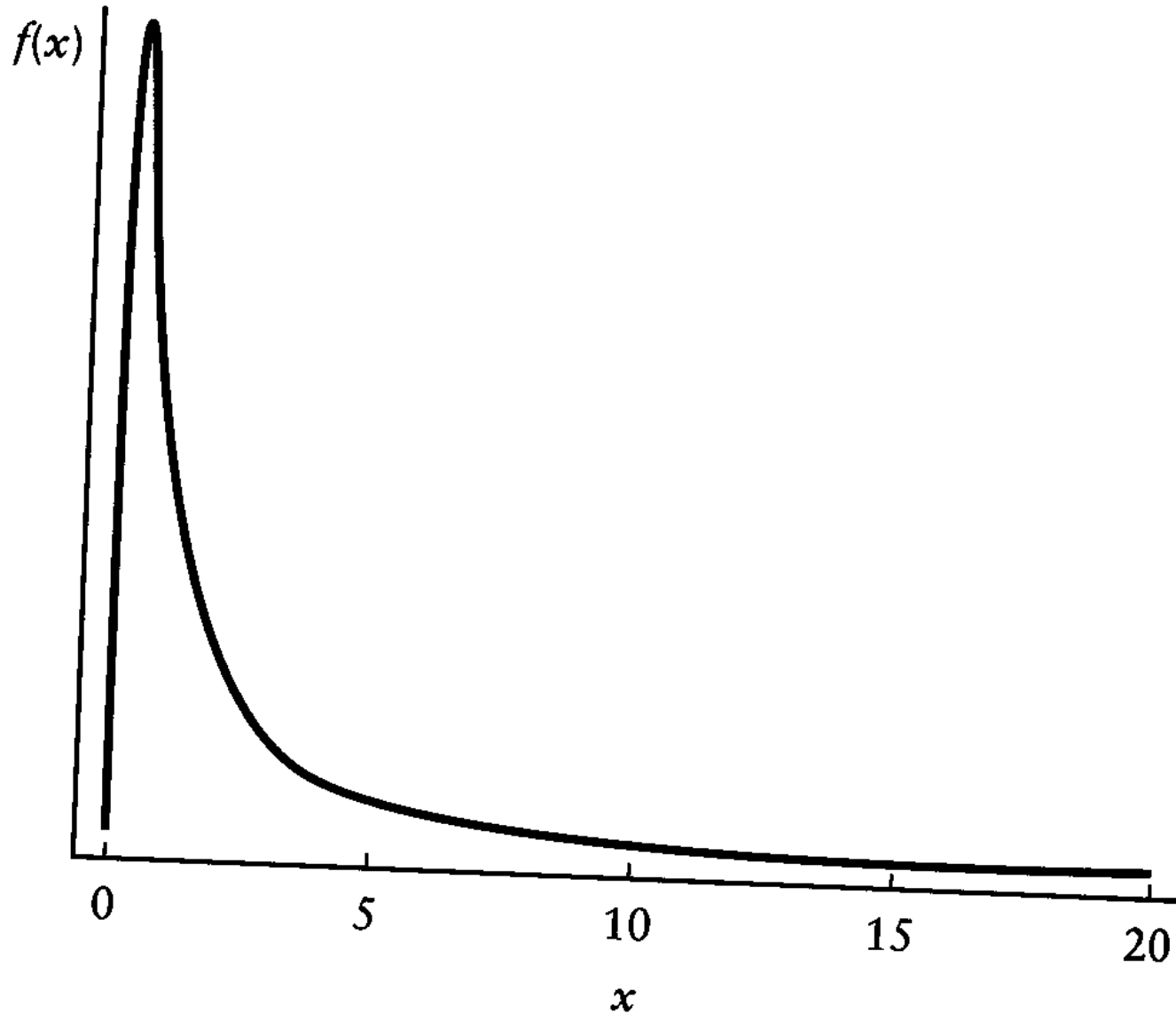
# Review Sums of Squares (SS)

- To calculate the total SS, we take the deviation of each observed data point from the mean of all points, square it, and sum all of these squared deviations.
- To find the F statistic we divide the mean square for the model  $MS_m$  to the mean square error  $MS_e$
- See Textbook page 289

# The F Distribution

- F comes from the ratio of the between- groups variability and the within-groups variability (this is where ANOVA gets its name from, i.e. the statistical analysis of the variability in a study).
- Somewhat like the t distribution, there are many possible F distributions, but it is never negative.
- It is skewed to the right (see next figure).
- It depends on the  $df_m$  of the numerator and the  $df_e$  of the denominator.

# The F Distribution for 4 and 2 df



# ANOVA Assumptions

- Independent observations
- Normal distribution of the continuous dependent variable for each level of the categorical independent variable
- Homogeneity of variance across groups

# One-Way ANOVA Example

- A new drug to treat acne is given to 35 subjects. It is tested against 2 drugs already on the market.
- Each person takes one of the drugs for 4 weeks, and the number of lesions is counted at the end of the study as the dependent variable.
- Model: lesions = type of drug
- Here lesions is the outcome and there is only one predictor variable type of drug (with three groups).

# One-Way ANOVA Results

Source	df	SS	MS	F	P
Model	2	2133.66	1066.83	262.12	0.001
Error	32	130.30	4.07		
Total	34	2263.96			

This test statistic has an F distribution with 2 and 32 degrees of freedom. We obtain the p-value from this F distrib to be 0.001 and it means the probability of obtaining values **over** 262.12 assuming that  $H_0$  true.

P-val is signif and we can conclude that not all means are eq, or at least one differs from the other two.

# 2-Way and 3-Way ANOVA

- In a 2-way ANOVA model, there are two independent variables, and in a 3-way ANOVA model there are three independent variables
- 2-way ANOVA example:  
Blood pressure = agegroup + gender
- 3-way ANOVA example:  
Blood pressure = agegroup + gender + diabetes



# Bonferroni Adjustment Method

- This test addresses the problem of multiple comparisons among the groups.
- If the overall model F test is statistically significant (e.g.  $P < 0.05$ ), then we reject the null hypothesis that all the groups' means are equal.
- However, it is often of interest to determine just which groups are different from the others.
- We don't want to inflate the alpha level, so we need a way to adjust the “pair-wise alpha” so that it keeps the “experiment-wide alpha” at 0.05

# Bonferroni's Method

- First decide on the experiment-wide alpha, which is usually  $\alpha = 0.05$ .
- Calculate the pair-wise alpha ( $\alpha'$ ) that will be used for each comparison of means.
- $\alpha' = 0.05 / c$  where  $0.05$  is the experiment-wide alpha, and  $c$  is the number of pairs of means that will be compared.

# Bonferonni Example

- Suppose we are evaluating a study of mice where 3 different diets (A, B, and C) were given to independent groups of animals, and their weight was measured at the end of the experiment.
- The ANOVA model:  $\text{weight} = \text{diet}$
- There are three possible diet comparisons  
( Number of pairs  $c$  is  $n(n-1)/2=3$  )
- Hence for each pair the new alpha is  **$\alpha' = 0.05/3 = 0.016$**
- We'll reject the null hypothesis only if  
the  $p\text{-value} < 0.016$

# Another Bonferroni Example

- In this example, we will see how to apply the t -test to each pair of compared means, and using the Bonferroni Method to decide which actual mean differences are significant.
- A study was performed to determine which of 5 water temperature settings for the machinery would be the best for removing impurities from water at a waste water treatment plant.
- ANOVA Model: % impurities removed = temp. setting
- Alpha was set at 0.10 for this experiment.

# Water Data

Temperature level				
I (low)	II (l-m)	III (med)	IV (m-h)	V (high)
40%	36%	49%	47%	55%
45%	42%	51%	49%	60%
42%	38%	53%	51%	62%
48%	39%	53%	52%	63%
50%	37%	52%	50%	59%
51%	40%	50%	51%	61%

Mean:    **46.0**                      **38.7**                      **51.3**                      **50.0**                      **60.0**

# ANOVA Table for this Example

Source	df	SS	MS	F	P
Model	4	1458	365.00	47.8	<0.01
Error	25	191	7.63		
Total	29	1649			

Overall reject  $H_0$  so at least one mean % impurities removed is diff from the others

# Multiple Comparisons

- Note that there are 5 temperature settings, which will yield 10 comparisons.
- Bonferroni  $\alpha' = \alpha/c = 0.10/10 = 0.01$
- What we do now is to calculate the p-values for all ten tests and use 0.01 to decide if they are statistically significant or not
- For example say that we obtain the following p-values: 0.001, 0.002, 0.005, 0.02, 0.05, ..., 0.99
- Then only for the first three we can claim that the results are statistically significant and reject the null hypothesis.

# One Last ANOVA Concept

- The **coefficient of variability**  $R^2$  describes how much of the variability in the continuous dependent variable was accounted for by the categorical independent variable.
- $R^2$  ranges from 0 (no variability explained) to 1 (100% of the variability explained).
- The more variables in the model, larger the  $R^2$
- We will not review how it is calculated; it is available in the software output.



# Example for STATA Demo

- Dataset includes pre-test scores and final exam scores for sophomores and juniors in a college course.
- The variables used for ANOVA are final exam score (continuous dependent variable) and year (categorical independent variable with two levels sophomores and juniors).

# STATA Program Code

- Generate descriptive statistics for final grade:
  - `sum final`

```
. sum final
```

Variable	Obs	Mean	Std. Dev.	Min	Max
final	19	85.73684	8.818176	72	99

# Program Code cont.

- One-way ANOVA for effect of year on final grade:
  - anova final year

Number of obs = 19      R-squared = 0.1344  
Root MSE = 8.44193      Adj R-squared = 0.0835

Source	Partial SS	df	MS	F	Prob > F
Model	188.160401	1	188.160401	2.64	0.1226
year	188.160401	1	188.160401	2.64	0.1226
Residual	1211.52381	17	71.2661064		
Total	1399.68421	18	77.7602339		

Note that One-Way Anova with two groups is the same as the t-test (same p-value).