



Two Way ANOVA

Evaluation Methods & Statistics- Lecture 11

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Research Example (Last Time)

- Consequences of a secondary task on driving
 - Texting
 - Talking on phone
 - Control (no secondary task)



Research Example (This week)

Consequences of a
secondary task on driving

- Texting
- Talking on phone
- Control (no secondary task)

Gender effect

- Male
- Female



What are our DV and IVs?



What are our DV and IVs?

IV 1- Secondary Driving Task

- Level 1- Control Group (No secondary task)
- Level 2- Texting
- Level 3- Talking

IV 2- Participant Gender

- Male
- Female

DV-Driving score



How would we analyse the data?



- We could do lots of t-tests
 - Control to Texting for Males
 - Control to Talking for Males
 - Talking to Texting for Males
 - Control to Texting for Females
 - Control to Talking for Females
 - Talking to Texting for Females
 - Control (Males) to Control (Females) etc.....

- This would inflate our *Type I error rate*

Factorial ANOVA- The Idea

- One Way ANOVA cannot deal with two factors (i.e. two Independent variables)
- We need to use a different type of ANOVA if we have two independent variables



Types of Factorial ANOVA

- Independent factorial design
 - All independent variables are between subjects
- Repeated Measures
 - All independent variables are within subjects
 - E.g. Measuring satisfaction pre and post (IV 1) 3 different interfaces (IV 2) all experienced one after another
- Mixed Design
 - Some independent variables are within subjects and some are between subjects
 - Gender (IV1) on pre and post scores (IV2)



Types of Factorial ANOVA

- What type of ANOVA are we going to use in our example?



Types of Factorial ANOVA

What type of ANOVA are we going to use in our example?

- Independent factorial design
 - All independent variables are between subjects



ANOVA Names

- Number of IV's
- Number of levels in IV's
- Experiment design used to gather data
 - Independent (Between subjects)
 - Repeated Measures (Within subjects)



Example

If we had:

- IV 1- Gender (2 levels- Between)
- IV2- Secondary Task (2 levels- Between)

2x2 Independent ANOVA

– or-

Two Way Independent ANOVA



Example

If we had:

- IV 1- Gender (2 levels- Between Subjects)
- IV2- Secondary Task (3 levels- Between Subjects)

2x3 Independent ANOVA

– or-

Two Way Independent ANOVA

Example

If we had:

- IV 1- Gender (2 levels) (Between Subject)
- IV2- Secondary Task (3 levels) (Within Subjects)

2x3 Mixed Design ANOVA

-or-

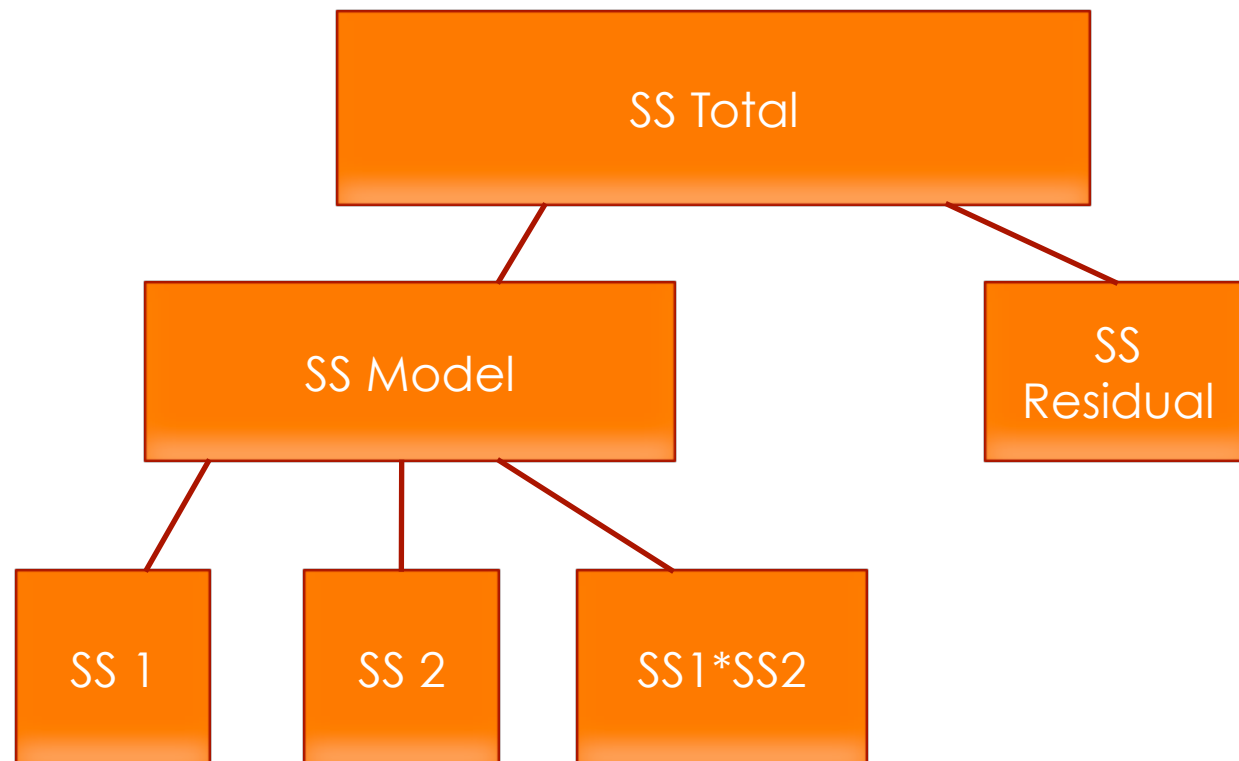
Two Way Mixed Design ANOVA

The Key (again): ANOVA & F Ratio



- F ratio is the ratio of **explained (that accounted for by the model we are proposing)** to **unexplained** variation
- This is calculated using the **Mean Squares**

What ANOVA is doing



Step 1- Total Sum of Squares

- The total amount of variation in our data
- This should look familiar (see Lecture 6)

$$SS_T = \sum \left(x_i - \bar{x}_{grand} \right)^2$$

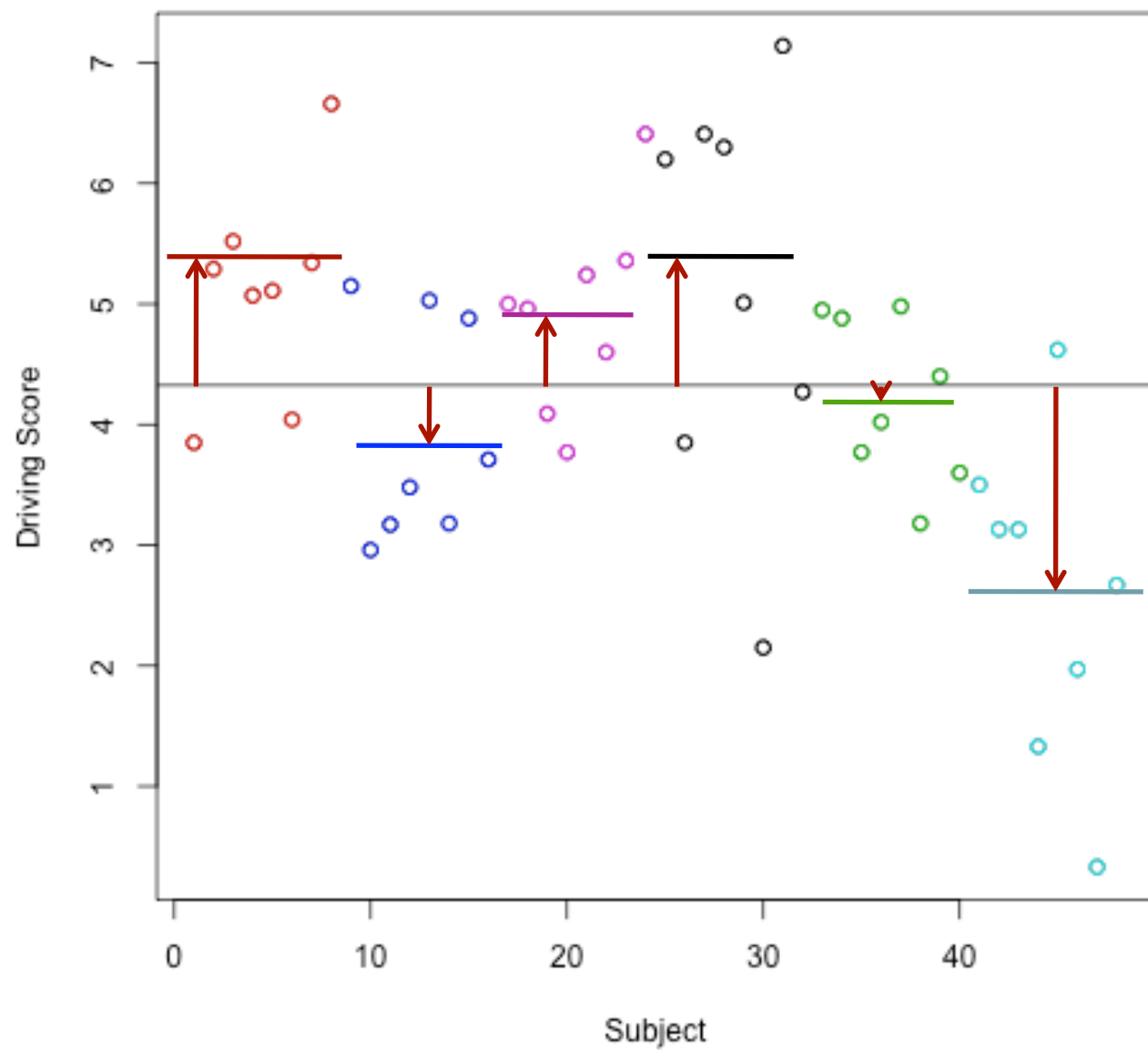
Step 2- Model Sum of Squares



- We now need to know how much variation our model can explain
- How much the total variation can be explained due to data points coming from different groups in “the perfect model”
- n_k is the number of people in that condition
- Sum all levels of IVs together

$$SS_M = \sum n_k (\bar{x}_k - \bar{x}_{grand})^2$$

Scatterplot of driving scores



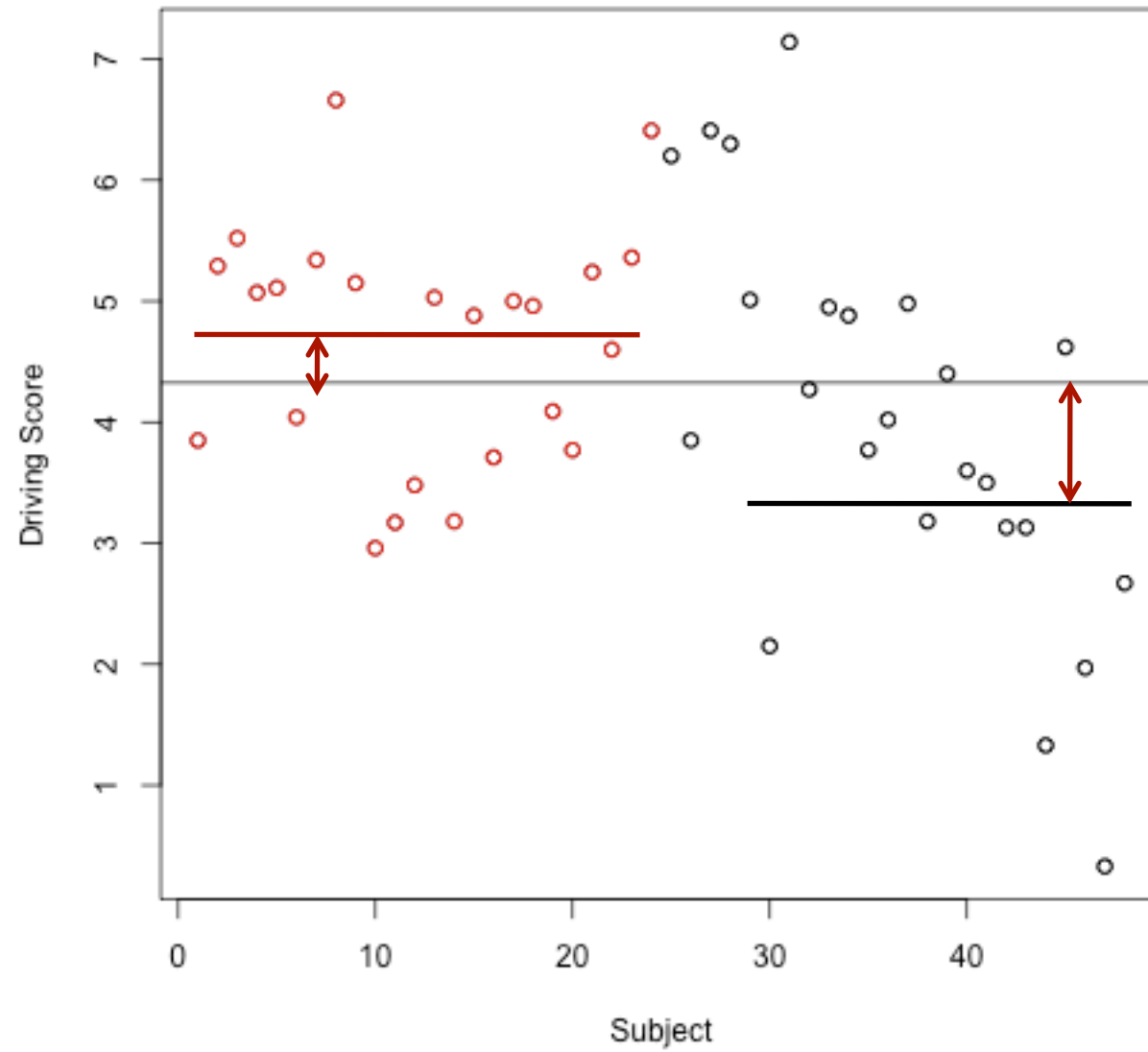
Step 3- Main Effect of Gender



- We group data by levels of Gender alone
- How much the total variation can be explained due to data points coming from different gender groups only in “the perfect model”
- n_k is the number of people in that condition
- Sum both levels of IV together

$$SS_1 = \sum n_k (\bar{x}_k - \bar{x}_{grand})^2$$

Scatterplot of driving scores

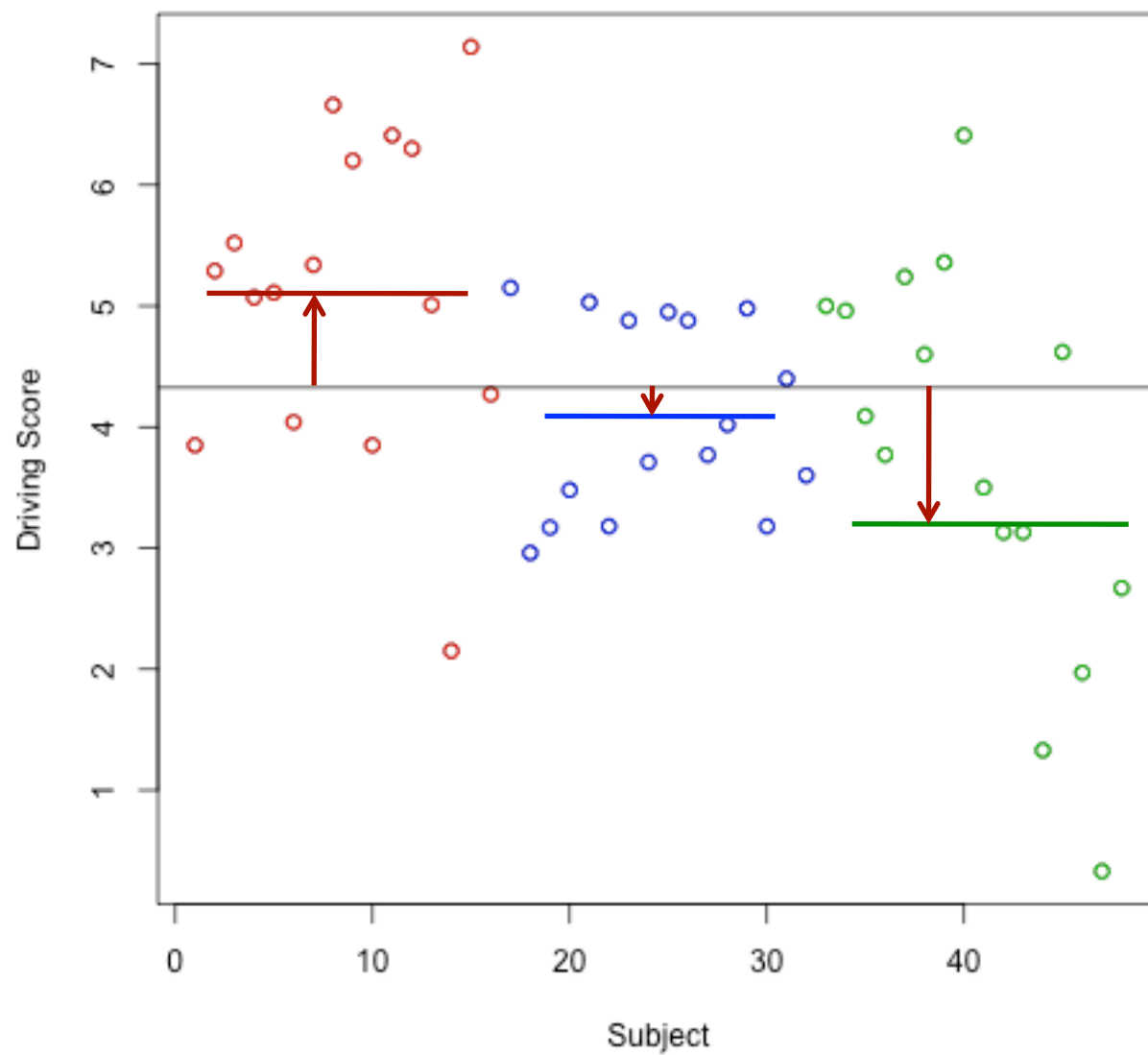


Step 4- Main Effect of Task

- We group data by levels of Task alone
- How much the total variation can be explained due to data points coming from different Task conditions alone in “the perfect model”
- n_k is the number of people in that condition
- Sum levels of IV together

$$SS_2 = \sum n_k (\bar{x}_k - \bar{x}_{grand})^2$$

Scatterplot of driving scores for control (red), talking (blue) and text (green) conditions



Step 5- Interaction (Gender *Task)



- SS_M made up of 3 components:
 - SS₁, SS₂, SS_{1*2}
 - Therefore easiest way is to take away SS₁ and SS₂ from SS_M

$$SS_{1*2} = SS_M - SS_1 - SS_2$$

Step 6- Residual Sum of Squares

- How much of the variation cannot be explained by the model i.e. what error is there in the model prediction?
- Easy way to calculate: $SS_R = SS_T - SS_M$



Degrees of Freedom for each SS



- Degrees of Freedom for SS_T (dfT):
 - $N-1 = 48-1 = 47$
- Degrees of Freedom for SS_M (dfM):
 - Gender: Number of Conditions (k) -1 = 1
 - Task: Number of Conditions (k) -1 = 2
 - Gender*Task= df Gender*df Task = 2
- Degrees of Freedom for SS_R (dfR):
 - $(N-1)*\text{Number of groups} = (8-1)*6 = 42$

F Ratio

- F Ratio gained for each effect (IV) and the interaction
- Mean Squares model (MS_M):
 - SS_M/df_M
 - This is done for main effects SS and interaction SS
- Mean Squares residual (error) (MS_R):
 - SS_R/df_R



F Ratio



Mean Square Model (MS_M)

Mean Square Residual (MS_R)

F Ratio



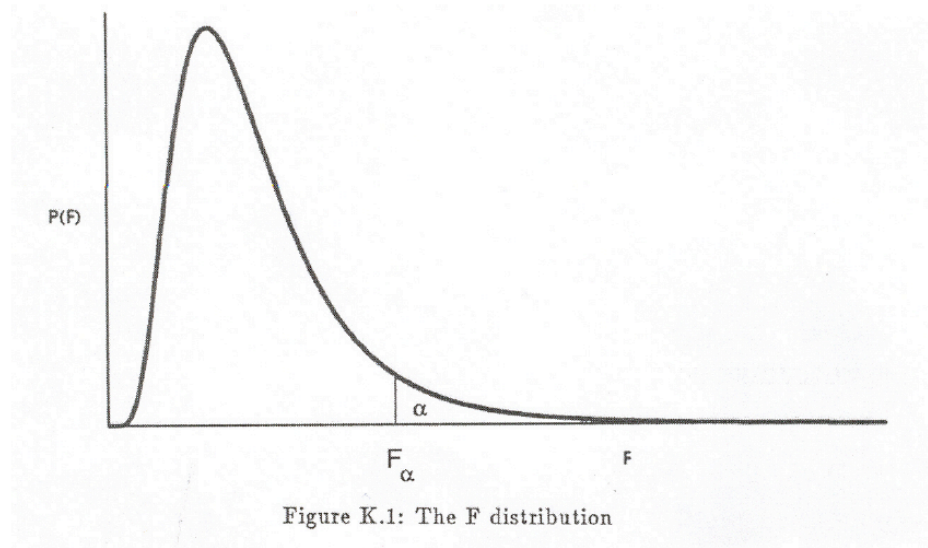
Variation explained by our model

Variation unexplained by our model

F Distribution

- F Distribution for specific pair of degrees of freedom

- Table of Critical Values



Critical values of F for the 0.05 significance level:

	1	2	3	4	5	6	7
1	161.45	199.50	215.71	224.58	230.16	233.99	236.77
2	18.51	19.00	19.16	19.25	19.30	19.33	19.35
3	10.13	9.55	9.28	9.12	9.01	8.94	8.89
4	7.71	6.94	6.59	6.39	6.26	6.16	6.09
5	6.61	5.79	5.41	5.19	5.05	4.95	4.88
6	5.99	5.14	4.76	4.53	4.39	4.28	4.21
7	5.59	4.74	4.35	4.12	3.97	3.87	3.79
8	5.32	4.46	4.07	3.84	3.69	3.58	3.50
9	5.12	4.26	3.86	3.63	3.48	3.37	3.29
10	4.97	4.10	3.71	3.48	3.33	3.22	3.14
11	4.84	3.98	3.59	3.36	3.20	3.10	3.01
12	4.75	3.89	3.49	3.26	3.11	3.00	2.91

Output in R

```
> summary(analysis)
              Df Sum Sq Mean Sq F value    Pr(>F)
gender           1    5.39    5.387    4.409 0.04180 *
condition         2   16.67    8.337    6.823 0.00272 **
gender:condition  2   16.91    8.453    6.918 0.00253 **
Residuals       42   51.32    1.222
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
~
```

Main Effect

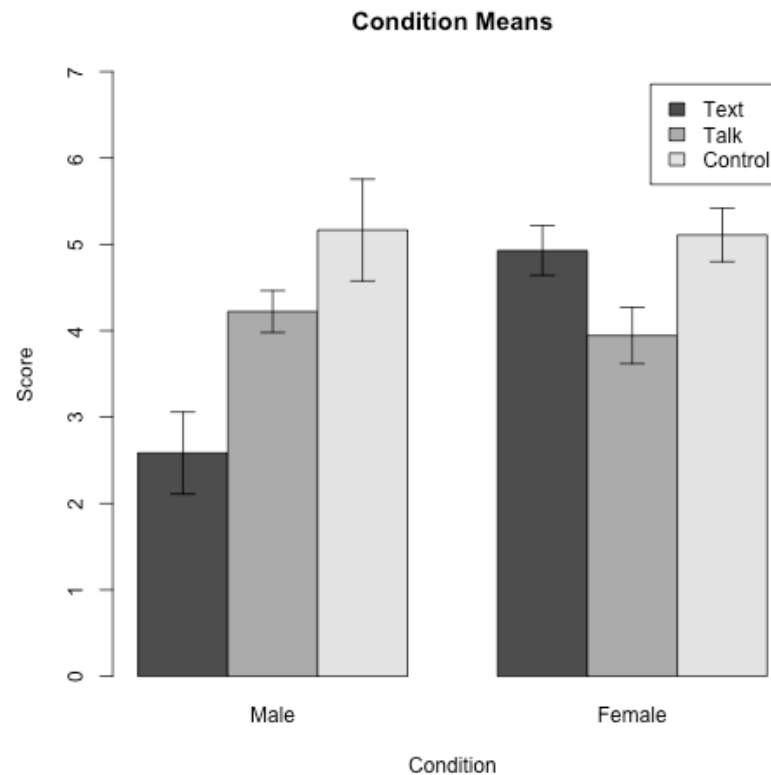
- Significant effect of an IV irrespective of the other IV
- Significant effect of gender on driving score irrespective of the task completed



Output in R

```
> summary(analysis)
              Df Sum Sq Mean Sq F value    Pr(>F)
gender           1    5.39    5.387    4.409 0.04180 *
condition         2   16.67    8.337    6.823 0.00272 **
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Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
~
```

Interactions



- The effect of secondary task on score is not that same for male and females
- interaction effect
- Females have a higher score in the texting condition compared to males
- Interactions supersedes main effects

Omnibus test & Post Hoc

- ANOVA is omnibus test
- How do they break down?
 - Male vs Female ? (Main Effect)
 - Control vs Text, Control vs Talk, Talk vs Text ? (Main effect of Task)
 - Male:Control vs Female Control etc (interaction)
- Again we need ***post hoc tests (See last lecture)***

Reporting ANOVA

- F ratio
- Degrees of Freedom (dof_M , dof_R)
- P value
 - $F(2, 42) = 5.097, p < 0.05$



Output in R

```
> summary(analysis)
              Df Sum Sq Mean Sq F value    Pr(>F)
gender           1    5.39    5.387    4.409 0.04180 *
condition         2   16.67    8.337    6.823 0.00272 **
gender:condition  2   16.91    8.453    6.918 0.00253 **
Residuals        42   51.32    1.222
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
~
```

Reporting ANOVA

e.g. Main effect of Gender

- $F(1, 42) = 4.409, p < 0.05$

Reading

- Field (2012) Discovering Statistics Using R, Chapters 12 -14

