

# INTRODUCTION TO ANOVA

Examples

# Example

Caffeine (Michael T. Brannick

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# Computational Example: Caffeine

G1: Control	G2: Mild	G3: Jolt
	Test Scores	
75	80	70
77	82	72
79	84	74
81	86	76
83	88	78

# Computational Example: Caffeine

G1: Control	G2: Mild	G3: Jolt
	Test Scores	
$75=79-4$	$80=84-4$	$70=74-4$
$77=79-2$	$82=84-2$	$72=74-2$
$79=79+0$	$84=84+0$	$74=74+0$
$81=79+2$	$86=84+2$	$76=74+2$
$83=79+4$	$88=84+4$	$78=74+4$
	Means	
79	84	74
Mean of means	79	

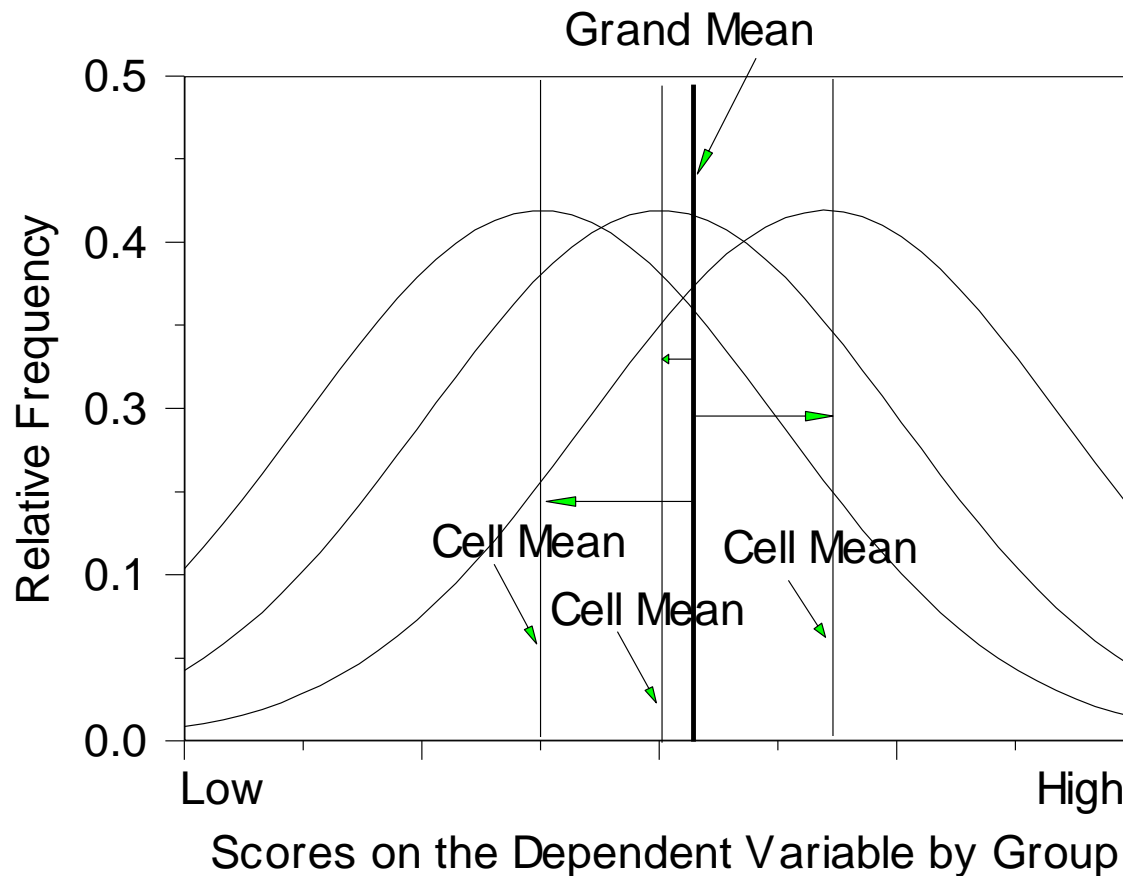
Between  
Sum of  
Squares

$$SS_B = \sum N_A (\bar{X}_A - \bar{X}_G)^2$$

	$\bar{X}_A$	$\bar{X}_G$	$(\bar{X}_A - \bar{X}_G)^2$
G1	79	79	0
Control	79	79	0
M=79	79	79	0
SD=3.16	79	79	0
	79	79	0
G2	84	79	25
M=84	84	79	25
SD=3.16	84	79	25
	84	79	25
	84	79	25
G3	74	79	25
M=74	74	79	25
SD=3.16	74	79	25
	74	79	25
	74	79	25
Sum			250

The between sum of squares relates the Cell Means to the Grand Mean. This is related to the variance of the means.

$$SS_B = \sum N_A (\bar{X}_A - \bar{X}_G)^2$$



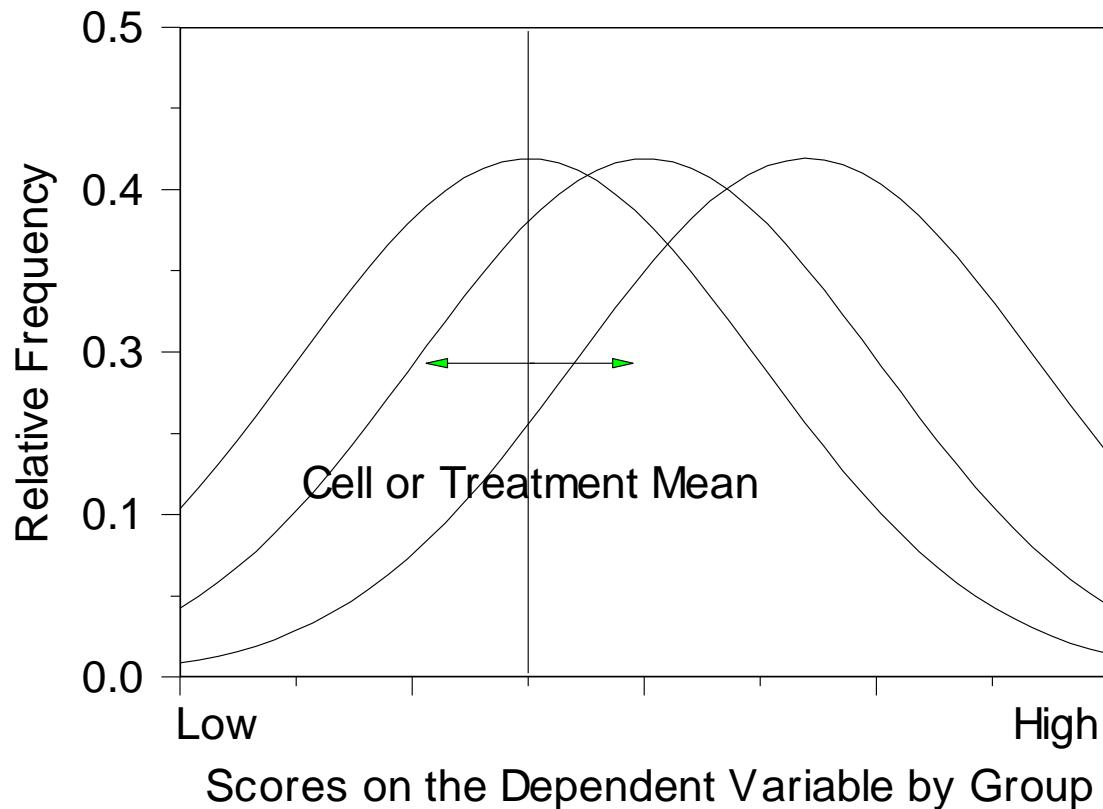
Within  
Sum of  
Squares

$$SS_W = \sum (X_i - \bar{X}_A)^2$$

	$X_i$	$\bar{X}_A$	$(X_i - \bar{X}_A)^2$
G1	75	79	16
Control	77	79	4
M=79	79	79	0
SD=3.16	81	79	4
	83	79	16
G2	80	84	16
M=84	82	84	4
SD=3.16	84	84	0
	86	84	4
	88	84	16
G3	70	74	16
M=74	72	74	4
SD=3.16	74	74	0
	76	74	4
	78	74	16
Sum			120

Within sum of squares refers to the variance within cells. That is, the difference between scores and their cell means.  $SS_W$  estimates error.

$$SS_W = \sum (X_i - \bar{X}_A)^2$$





- Calculate the MSE

$$MSE = \frac{1}{N - K} \sum_j \sum_i (x_{ij} - \bar{X}_j)^2 = SS_w / N - K$$

- $120 / (15 - 3) = 10$

$$F = \frac{SSB / (K - 1)}{MSE}$$

- $F = 250 / (3 - 1) / 10 = 12.5$

# ANOVA Source Table (1)

Source	SS	df	MS	F
Between Groups	250	$k-1=2$	$SS/df$ $250/2=$ $125$ $=MS_B$	$F =$ $MS_B/MS_W$ $= 125/10$ $=12.5$
Within Groups	120	$N-k=$ $15-3=12$	$120/12 =$ $10 =$ $MS_W$	
Total	370	$N-1=14$		

# ANOVA Source Table (2)

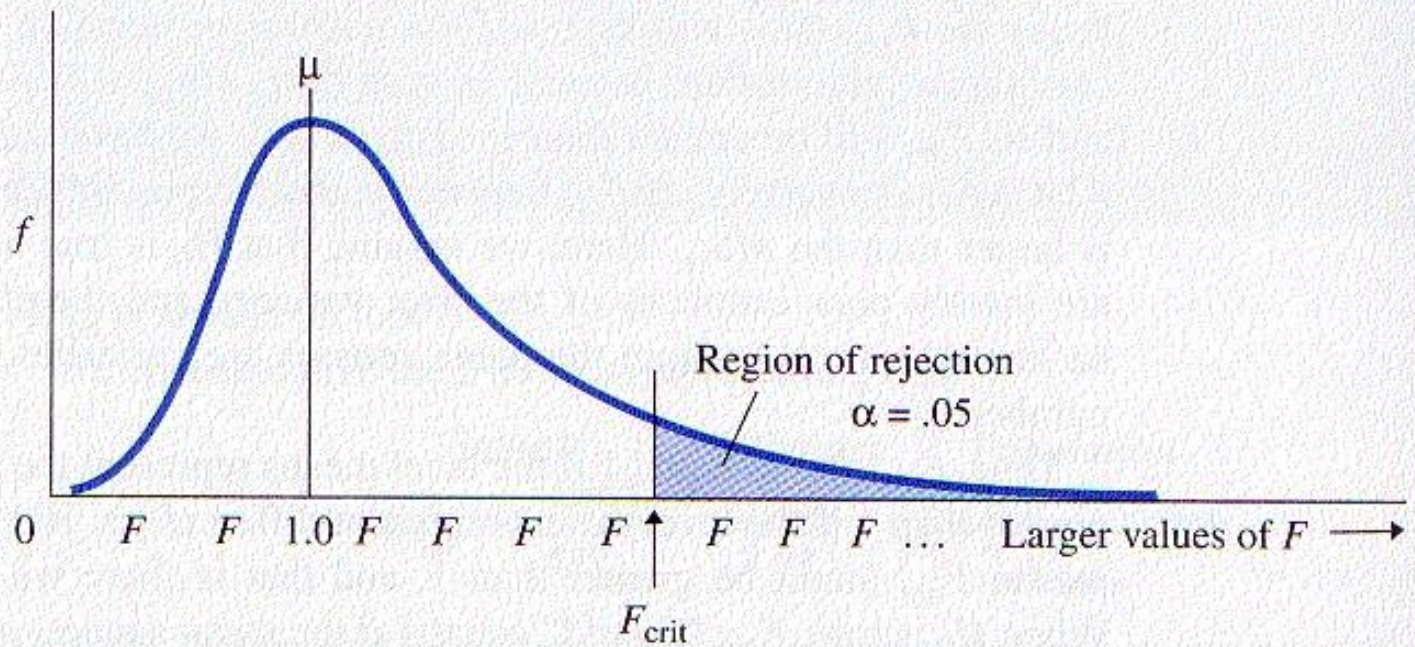
- $df$  – Degrees of freedom. Divide the sum of squares by degrees of freedom to get
- $MS$ , Mean Squares, which are population variance estimates.
- $F$  is the ratio of two mean squares.  $F$  is another distribution like  $z$  and  $t$ . There are tables of  $F$  used for significance testing.

# The F Distribution

- F" distribution, with  $K-1$  and  $N-K$  degrees of freedom (where  $K$  is the number of groups and  $N$  is the total number of observations)
- $K=3, N=15 \rightarrow F(2,12)$  (working at 0.05)

# The F Distribution

**FIGURE 17.2** Sampling Distribution of  $F$  When  $H_0$  Is True



# F Table – Critical Values

	Numerator df: $df_B$				
$df_W$	1	2	3	4	5
5 5%	6.61	5.79	5.41	5.19	5.05
1%	16.3	13.3	12.1	11.4	11.0
10 5%	4.96	4.10	3.71	3.48	3.33
1%	10.0	7.56	6.55	5.99	5.64
12 5%	4.75	3.89	3.49	3.26	3.11
1%	9.33	6.94	5.95	5.41	5.06
14 5%	4.60	3.74	3.34	3.11	2.96
1%	8.86	6.51	5.56	5.04	4.70

# Review 6 Steps

- Set alpha (.05).
- State Null & Alternative
- $H_0: \mu_1 = \mu_2 = \mu_3$
- $H_1$ : not all  $\mu$  are =.
- Calculate test statistic:  
 $F=12.5$
- Determine critical value  $F_{.05}(2,12) = 3.89$
- Decision rule: If test statistic  $>$  critical value, reject  $H_0$ .
- Decision: Test is significant ( $12.5 > 3.89$ ).  
**population are different.**