

# Two-way anova

# Aims

- Rationale of factorial ANOVA
- Partitioning variance
- Interaction effects
  - ▶ Interaction graphs
  - ▶ Interpretation

# What is Two-Way Independent ANOVA?

- Two independent variables
  - ▶ Two-way = 2 Independent variables
  - ▶ Three-way = 3 Independent variables
- Different participants in *all* conditions
  - ▶ Independent = 'different participants'
- Several independent variables is known as a factorial design.

# Benefit of Factorial Designs

- We can look at how variables *interact*.
- Interactions
  - ▶ Show how the effects of one var might depend on the effects of another
  - ▶ Are often more interesting than main effects.
- Examples
  - ▶ Interaction between stress and well-watered conditions on the growth of different plants.

# An Example

- Field (2009): Testing the effects of alcohol and gender on 'the beer-goggles effect':
  - ▶ V 1 (**Alcohol**): none, 2 pints, 4 pints
  - ▶ V 2 (**Gender**): male, female
- Dependent variable (DV) was an objective measure of the attractiveness of the partner selected at the end of the evening.

# Factorial anova as regression

- Subset the data
  - ▶ Gender: “Male” and “Female”, ref level=“Male”
  - ▶ For alcohol: keep the levels “None” and “4 pints”, ref level =“None”
  - ▶ We obtain 2 factors, each with two levels

# Factorial anova as regression

$$\text{Attractiveness}_i = b_0 + b_1 * \text{gender}_i + b_2 * \text{alcohol}_i + b_3 * \text{gender}_i * \text{alcohol}_i + \varepsilon_i$$

$$E(A|M, \text{none}) = b_0$$

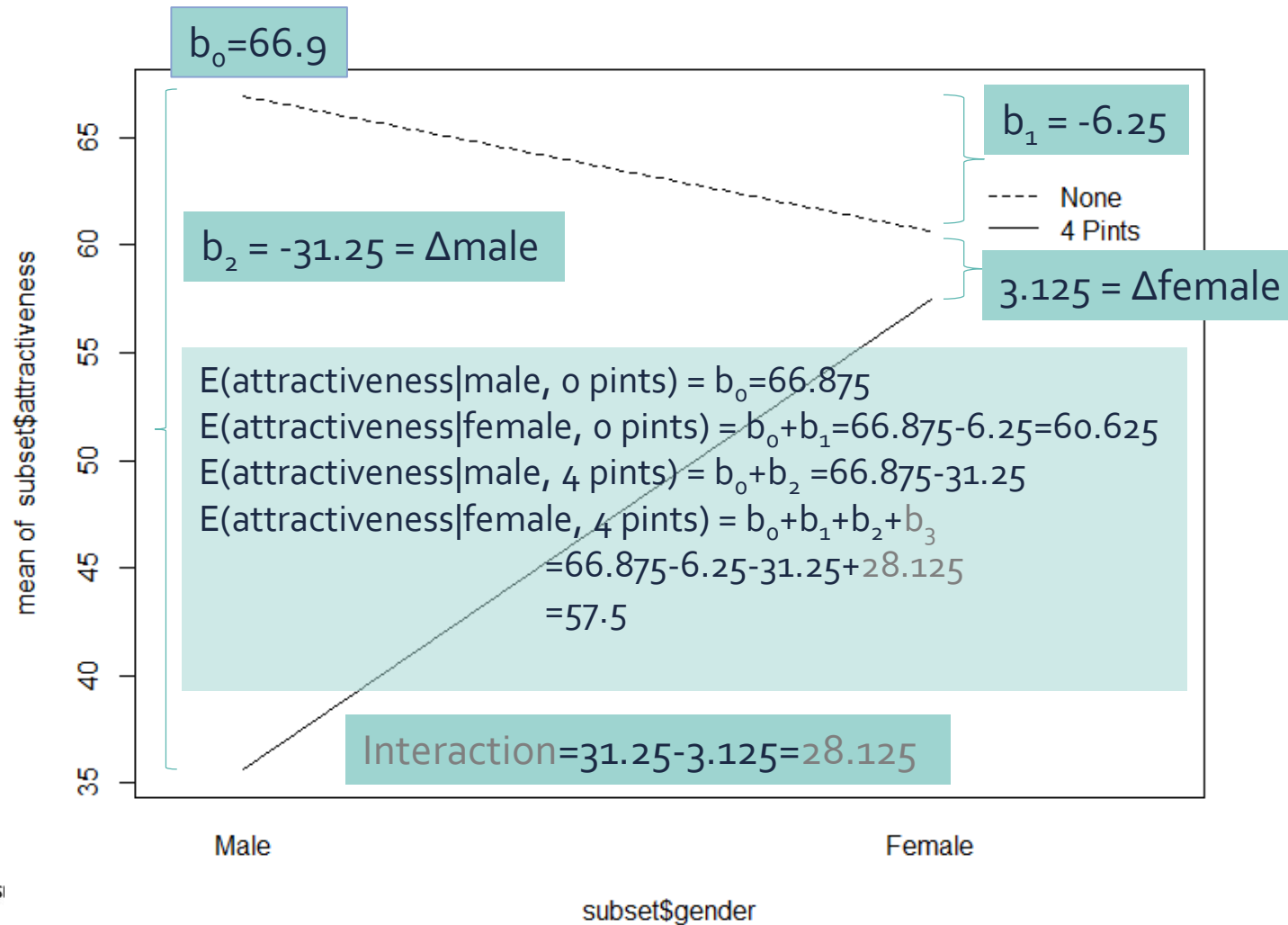
$$E(A|F, \text{none}) = b_0 + b_1$$

$$E(A|M, 4\text{pints}) = b_0 + b_2$$

$$E(A|F, 4\text{pints}) = b_0 + b_1 + b_2 + b_3$$

Parameter		Estimate		Standard Error	t Value	Pr >  t
Intercept	$b_0$	66.87500000	B	3.05502002	21.89	<.0001
gender Female	$b_1$	-6.25000000	B	4.32045075	-1.45	0.1591
gender Male		0.00000000	B	-	-	-
alcohol 4 Pints	$b_2$	-31.25000000	B	4.32045075	-7.23	<.0001
alcohol None		0.00000000	B	-	-	-
gender*alcohol Female 4 P	$b_3$	28.12500000	B	6.11004004	4.60	<.0001
gender*alcohol Female None		0.00000000	B	-	-	-
gender*alcohol Male 4 Pints		0.00000000	B	-	-	-
gender*alcohol Male None		0.00000000	B	-	-	-

# What Is an Interaction?





# Factorial anova as regression

$$E(A|M, \text{none}) = b_0$$

$$E(A|F, \text{none}) = b_0 + b_1$$

$$E(A|M, 4\text{pints}) = b_0 + b_2$$

$$E(A|F, 4\text{pints}) = b_0 + b_1 + b_2 + b_3$$

$b_0$  is the mean attractiveness for men that drunk no alcohol  
(ref=men, no alcohol)

$b_1$  is the difference in mean attractiveness between women that drunk no alcohol and men that drunk no alcohol

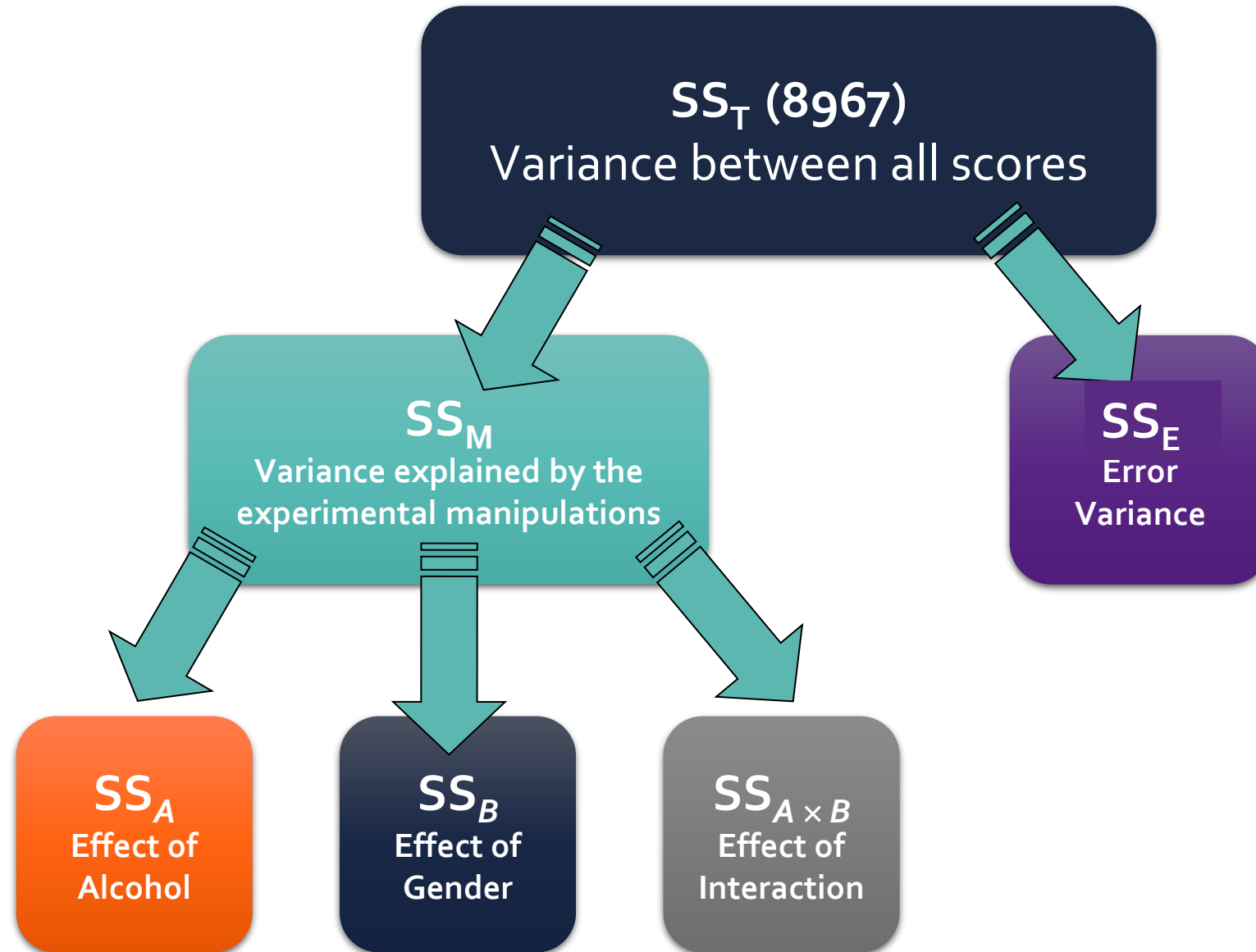
$b_2$  is the difference in mean attractiveness between men that drunk 4 pints vs none

$b_3$  compares the difference between men and women in the no alcohol condition to the difference between men and women in the 4 pints condition

# Variance partitioning balanced data

**Table 12.1:** Data for the beer-goggles effect

Alcohol	None	j=1	2 Pints	j=2	4 Pints	j=3
Gender	Female k=1	Male k=2	Female	Male	Female	Male
R=8	65	50	70	45	55	30
	70	55	65	60	65	30
	60	80	60	85	70	30
	60	65	70	65	55	55
	60	70	65	70	55	35
	55	75	60	70	60	20
	60	75	60	80	50	45
	55	65	50	60	50	40
Total	485	535	500	535	460	285
Mean $\bar{x}_{.jk}$	60.625	66.875	62.50	66.875	57.50	35.625
Variance $s^2_{jk}$	24.55	106.70	42.86	156.70	50.00	117.41



# Step 1: Calculate $SS_T$

65	50	70	45	55	30
50	55	65	60	65	30
70	80	60	85	70	30
45	65	70	65	55	55
55	70	65	70	55	35
30	75	60	70	60	20
70	75	60	80	50	45
55	65	50	60	50	40

$\bar{x}_{...} = \text{Grand Mean} = 58.33$

$$SS_T = s^2(N - 1)$$

$$SS_T = 190.78(48 - 1)$$

$$SS_T = 8966.67$$

## Step 2: Calculate $SS_M$

$$SS_M = \sum_{j=1}^k n_j (\bar{x}_{.j} - \bar{x}_{..})^2$$

$$\begin{aligned} SS_M &= 8(60.625 - 58.33)^2 + 8(66.875 - 58.33)^2 + 8(62.5 - 58.33)^2 \\ &\quad + 8(66.875 - 58.33)^2 + 8(57.5 - 58.33)^2 + 8(35.625 - 58.33)^2 \\ SS_M &= 5479.167 \end{aligned}$$

# Step 2a: Calculate $SS_G$

Female		
65	70	55
70	65	65
60	60	70
60	70	55
60	65	55
55	60	60
60	60	50
55	50	50

Mean Female = 60.21

$\bar{x}_{..k}$

Male		
50	45	30
55	60	30
80	85	30
65	65	55
70	70	35
75	70	20
75	80	45
65	60	40

Mean Male = 56.46

$$SS_{Gender} = \sum_{k=1}^K n_k (\bar{x}_{..k} - \bar{x}_{...})^2$$

$$SS_{Gender} = 24(60.21 - 58.33)^2 + 24(56.46 - 58.33)^2 = 168.75$$

# Step 2b: Calculate $SS_A$

None	
65	50
70	55
60	80
60	65
60	70
55	75
60	75
55	65

Mean None = 63.75

$\bar{x}_{.j.}$

2 Pints	
70	45
65	60
60	85
70	65
65	70
60	70
60	80
50	60

Mean 2 Pints =  
64.6875

4 Pints	
55	30
65	30
70	30
55	55
55	35
60	20
50	45
50	40

Mean 4 Pints =  
46.5625

$$SS_{Alcohol} = \sum_{j=1}^J n_j (\bar{x}_{.j.} - \bar{x}_{...})^2$$

$$SS_{Alcohol} = 16(63.75 - 58.33)^2 + 16(64.6875 - 58.33)^2 + 16(46.5625 - 58.33)^2$$

$$SS_{Alcohol} = 3332.292$$

## Step 2c: Calculate $SS_{A*G}$

$$\begin{aligned}SS_{A*G} &= SS_M - SS_{Alcohol} - SS_{Gender} \\SS_{A*G} &= 5479.167 - 168.75 - 3332.292 \\SS_{A*G} &= 1978.125\end{aligned}$$

Note: True in balanced designs



## Step 3: Calculate $SS_E$

$$SS_E = \sum_{j=1}^J \sum_{k=1}^K s_{jk}(n_{jk} - 1)$$

$$\begin{aligned} SS_E &= s_{11}^2(n_{11} - 1) + s_{12}^2(n_{12} - 1) \\ &+ s_{21}^2(n_{21} - 1) + s_{22}^2(n_{22} - 1) \\ &+ s_{31}^2(n_{31} - 1) + s_{32}^2(n_{32} - 1) \\ SS_E &= (24.55 * 7) + (106.7 * 7) \\ &+ (42.86 * 7) + (156.7 * 7) \\ &+ (50 * 7) + (117.41 * 7) \\ &= 3487.52 \end{aligned}$$

# Two-way anova table

Source of variation	Degrees of freedom	Sum of squares	Mean square	F-ratio
Factor A	J-1	$SS_A$	$MS_A = SS_A / (J-1)$	$F_A = MS_A / MS_E$
Factor G	K-1	$SS_G$	$MS_G = SS_G / (K-1)$	$F_G = MS_G / MS_E$
Interaction	$(J-1)(K-1)$	$SS_{AG}$	$MS_{AG} = SS_{AG} / (J-1)(K-1)$	$F_{AG} = MS_{AG} / MS_E$
Error	$JK(R-1) = N - JK$	$SS_E$	$MS_E = SS_E / JK(R-1)$	

# Sums of squares

- When data is unbalanced, there are different ways to calculate the sums of squares. Assume the model  $A + B + A*B$ 
  - ▶ **Type I SS:** Tests for the presence of an effect given that the previous one stated is already in the model
    - $SS(A)$ : reduction in residual SS attributable to A
    - $SS(B|A)$ : reduction in residual SS attributable to B when A is already in the model
    - $SS(A*B|A,B)$ : reduction in residual SS attributable to  $A*B$  when A and B are already in the model
  - ▶ **Type II SS:** Tests for the presence of an effect, given that the others not containing this term are already in the model
    - $SS(A|B)$ ,  $SS(B|A)$ ,  $SS(A*B|A,B)$
  - ▶ **Type III SS:** Tests for the presence of an effect, given that the others are in the model
    - $SS(A|B, A*B)$ ,  $SS(B|A, A*B)$ ,  $SS(A*B|A, B)$
- Do not interpret a main effect if interactions are present (generally speaking, if a significant interaction is present, the main effects should not be further analysed).
- When data is balanced, types I, II and III all give the same results.

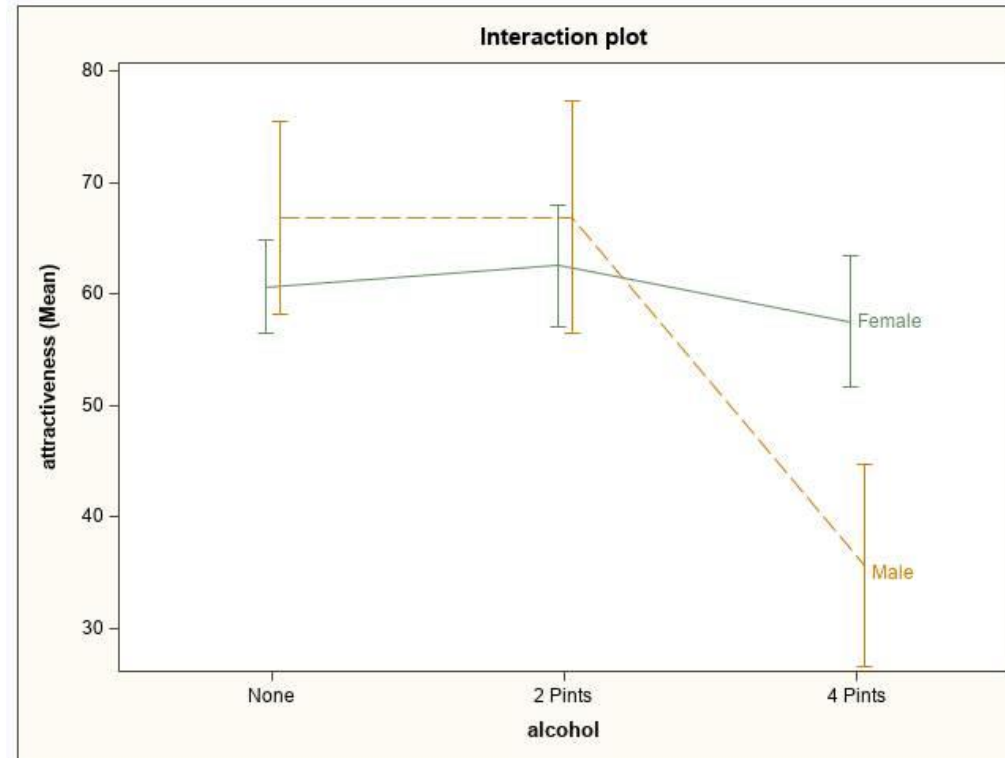
# Interpreting Factorial ANOVA

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	4433.593750	1477.864583	19.79	<.0001
Error	28	2090.625000	74.665179		
Corrected Total	31	6524.218750			

R-Square	Coeff Var	Root MSE	attractiveness Mean
0.679559	15.66622	8.640901	55.15625

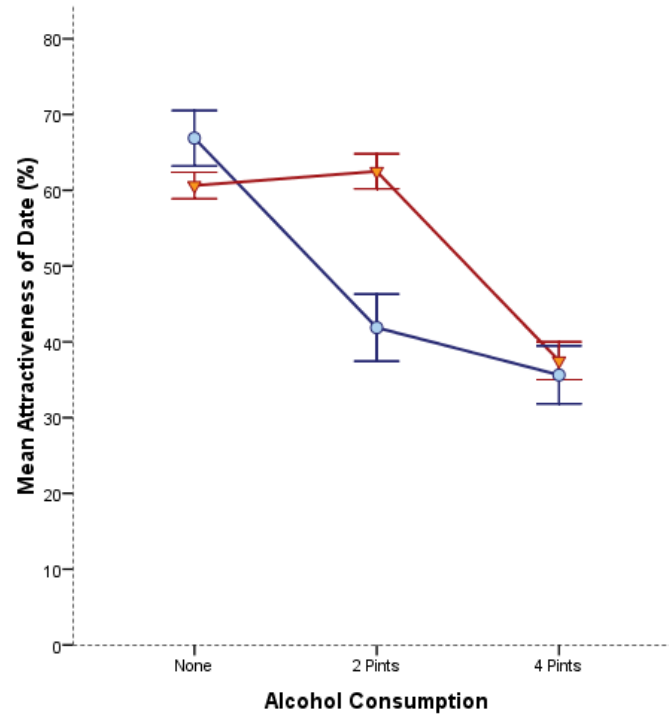
Source	DF	Type III SS	Mean Square	F Value	Pr > F
gender	1	488.281250	488.281250	6.54	0.0163
alcohol	1	2363.281250	2363.281250	31.65	<.0001
gender*alcohol	1	1582.031250	1582.031250	21.19	<.0001

# Interpretation: Interaction

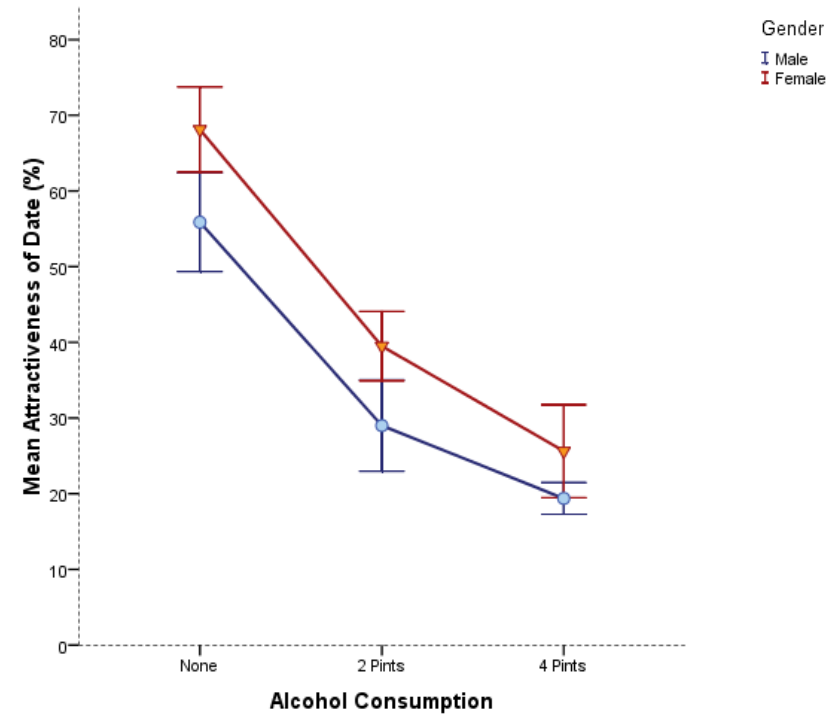


There was a significant interaction between the amount of alcohol consumed and the gender of the person selecting a mate, on the attractiveness of the partner selected ( $p=8e-5$ ).

# Is There Likely to Be a Significant Interaction Effect?

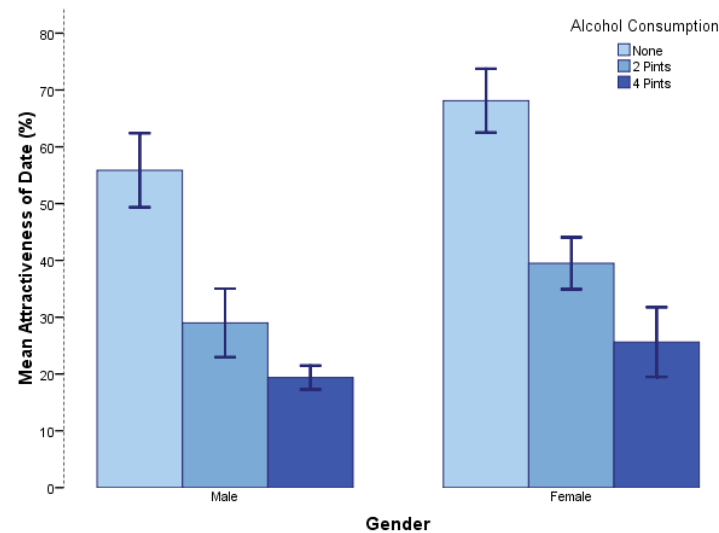


Yes

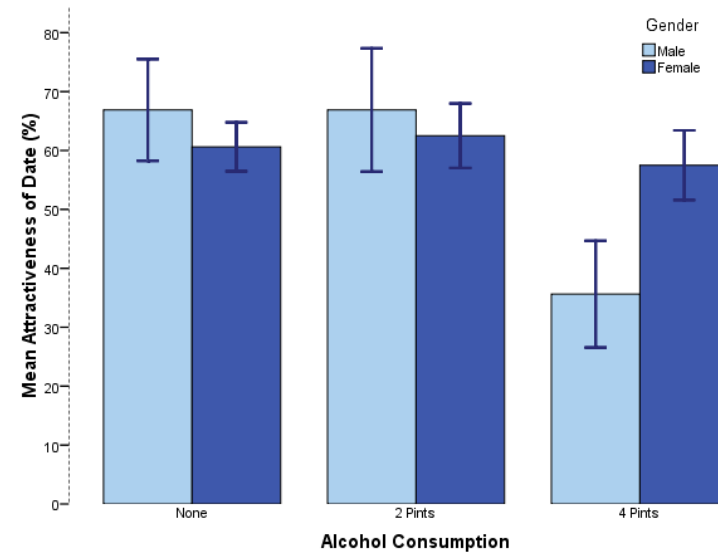


No

# Is There Likely to Be a Significant Interaction Effect?



No



Yes

# Workflow two-way anova

1. Summary statistics
2. Graphs (box plots, bar charts, interaction plot)
3. Anova model
4. Testing assumptions
5. Post-hoc tests provided that a significant interaction was found



# The anova table

## glm model

### The GLM Procedure

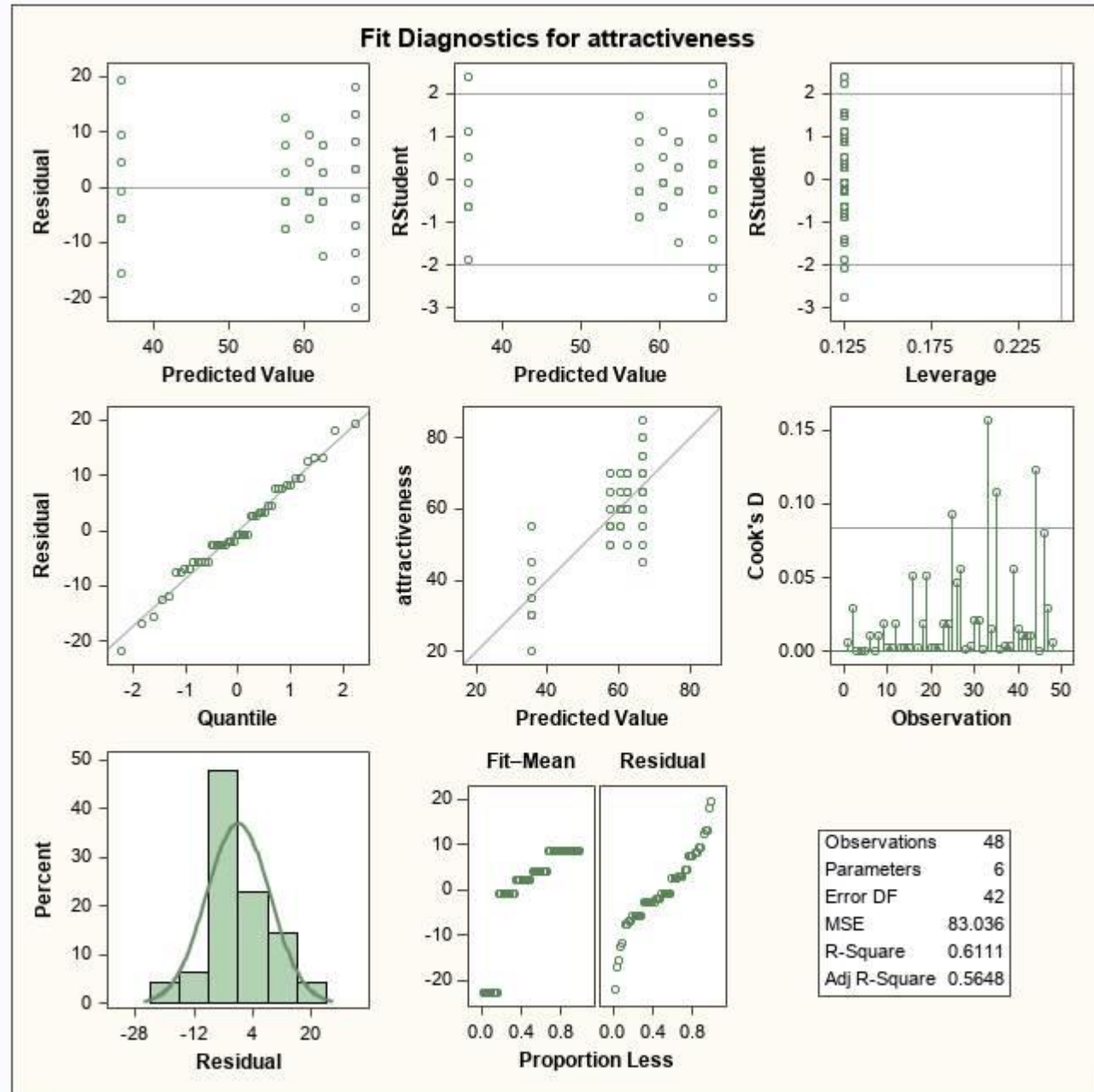
Dependent Variable: attractiveness

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	5	5479.166667	1095.833333	13.20	<.0001
Error	42	3487.500000	83.035714		
Corrected Total	47	8966.666667			

R-Square	Coeff Var	Root MSE	attractiveness Mean
0.611059	15.62125	9.112393	58.33333

Source	DF	Type III SS	Mean Square	F Value	Pr > F
gender	1	168.750000	168.750000	2.03	0.1614
alcohol	2	3332.291667	1666.145833	20.07	<.0001
gender*alcohol	2	1978.125000	989.062500	11.91	<.0001

# Testing the assumptions



# All-pairwise comparisons

Differences of gender*alcohol Least Squares Means Adjustment for Multiple Comparisons: Tukey														
gender	alcohol	_gender	_alcohol	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
Female	2 Pints	Female	4 Pints	5.0000	4.5562	42	1.10	0.2787	0.8796	0.05	-4.1948	14.1948	-8.6013	18.6013
Female	2 Pints	Female	None	1.8750	4.5562	42	0.41	0.6828	0.9984	0.05	-7.3198	11.0698	-11.7263	15.4763
Female	2 Pints	Male	2 Pints	-4.3750	4.5562	42	-0.96	0.3424	0.9278	0.05	-13.5698	4.8198	-17.9763	9.2263
Female	2 Pints	Male	4 Pints	26.8750	4.5562	42	5.90	<.0001	<.0001	0.05	17.6802	36.0698	13.2737	40.4763
Female	2 Pints	Male	None	-4.3750	4.5562	42	-0.96	0.3424	0.9278	0.05	-13.5698	4.8198	-17.9763	9.2263
Female	4 Pints	Female	None	-3.1250	4.5562	42	-0.69	0.4966	0.9826	0.05	-12.3198	6.0698	-16.7263	10.4763
Female	4 Pints	Male	2 Pints	-9.3750	4.5562	42	-2.06	0.0459	0.3287	0.05	-18.5698	-0.1802	-22.9763	4.2263
Female	4 Pints	Male	4 Pints	21.8750	4.5562	42	4.80	<.0001	0.0003	0.05	12.6802	31.0698	8.2737	35.4763
Female	4 Pints	Male	None	-9.3750	4.5562	42	-2.06	0.0459	0.3287	0.05	-18.5698	-0.1802	-22.9763	4.2263
Female	None	Male	2 Pints	-6.2500	4.5562	42	-1.37	0.1774	0.7432	0.05	-15.4448	2.9448	-19.8513	7.3513
Female	None	Male	4 Pints	25.0000	4.5562	42	5.49	<.0001	<.0001	0.05	15.8052	34.1948	11.3987	38.6013
Female	None	Male	None	-6.2500	4.5562	42	-1.37	0.1774	0.7432	0.05	-15.4448	2.9448	-19.8513	7.3513
Male	2 Pints	Male	4 Pints	31.2500	4.5562	42	6.86	<.0001	<.0001	0.05	22.0552	40.4448	17.6487	44.8513
Male	2 Pints	Male	None	4.23E-15	4.5562	42	0.00	1.0000	1.0000	0.05	-9.1948	9.1948	-13.6013	13.6013
Male	4 Pints	Male	None	-31.2500	4.5562	42	-6.86	<.0001	<.0001	0.05	-40.4448	-22.0552	-44.8513	-17.6487



# Simple tests of effect

Simple Differences of gender*alcohol Least Squares Means Adjustment for Multiple Comparisons: Bonferroni													
Slice	alcohol	_alcohol	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
gender Female	2 Pints	4 Pints	5.0000	4.5562	42	1.10	0.2787	0.8361	0.05	-4.1948	14.1948	-6.3616	16.3616
gender Female	2 Pints	None	1.8750	4.5562	42	0.41	0.6828	1.0000	0.05	-7.3198	11.0698	-9.4866	13.2366
gender Female	4 Pints	None	-3.1250	4.5562	42	-0.69	0.4966	1.0000	0.05	-12.3198	6.0698	-14.4866	8.2366

Simple Differences of gender*alcohol Least Squares Means Adjustment for Multiple Comparisons: Bonferroni													
Slice	alcohol	_alcohol	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
gender Male	2 Pints	4 Pints	31.2500	4.5562	42	6.86	<.0001	<.0001	0.05	22.0552	40.4448	19.8884	42.6116
gender Male	2 Pints	None	4.23E-15	4.5562	42	0.00	1.0000	1.0000	0.05	-9.1948	9.1948	-11.3616	11.3616
gender Male	4 Pints	None	-31.2500	4.5562	42	-6.86	<.0001	<.0001	0.05	-40.4448	-22.0552	-42.6116	-19.8884

Simple Differences of gender*alcohol Least Squares Means Adjustment for Multiple Comparisons: Bonferroni													
Slice	gender	_gender	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
alcohol 2 Pints	Female	Male	-4.3750	4.5562	42	-0.96	0.3424	0.3424	0.05	-13.5698	4.8198	-13.5698	4.8198

Simple Differences of gender*alcohol Least Squares Means Adjustment for Multiple Comparisons: Bonferroni													
Slice	gender	_gender	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
alcohol 4 Pints	Female	Male	21.8750	4.5562	42	4.80	<.0001	<.0001	0.05	12.6802	31.0698	12.6802	31.0698

Simple Differences of gender*alcohol Least Squares Means Adjustment for Multiple Comparisons: Bonferroni													
Slice	gender	_gender	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
alcohol None	Female	Male	-6.2500	4.5562	42	-1.37	0.1774	0.1774	0.05	-15.4448	2.9448	-15.4448	2.9448

# User defined contrasts

Least Squares Means Estimates Adjustment for Multiplicity: Sidak												
Effect	Label	Estimate	Standard Error	DF	t Value	Pr >  t	Adj P	Alpha	Lower	Upper	Adj Lower	Adj Upper
gender*alcohol	none female - none male	-6.2500	4.5562	42	-1.37	0.1774	0.7904	0.05	-15.4448	2.9448	-19.3290	6.8290
gender*alcohol	2 pints female - 2 pints male	-4.3750	4.5562	42	-0.96	0.3424	0.9650	0.05	-13.5698	4.8198	-17.4540	8.7040
gender*alcohol	4 pints female - 4 pints male	21.8750	4.5562	42	4.80	<.0001	0.0002	0.05	12.6802	31.0698	8.7960	34.9540
gender*alcohol	d1:female 4 pints - female none	-3.1250	4.5562	42	-0.69	0.4966	0.9959	0.05	-12.3198	6.0698	-16.2040	9.9540
gender*alcohol	female 2 pints - female none	1.8750	4.5562	42	0.41	0.6828	0.9999	0.05	-7.3198	11.0698	-11.2040	14.9540
gender*alcohol	d2:male 4 pints - male none	-31.2500	4.5562	42	-6.86	<.0001	<.0001	0.05	-40.4448	-22.0552	-44.3290	-18.1710
gender*alcohol	male 2 pints - male none	4.23E-15	4.5562	42	0.00	1.0000	1.0000	0.05	-9.1948	9.1948	-13.0790	13.0790
gender*alcohol	d1 vs d2	28.1250	6.4434	42	4.36	<.0001	0.0006	0.05	15.1216	41.1284	9.6284	46.6216



# To Transform ... Or Not

- Transforming the data helps as often as it hinders the accuracy of  $F$  (Games & Lucas, 1966).
- Games (1984):
  - ▶ The central limit theorem: sampling distribution of the mean will be normal in samples  $> 30$  anyway.
  - ▶ Transforming the data changes the hypothesis being tested
    - E.g. when using a log transformation and comparing means, you change from comparing arithmetic means to comparing geometric means

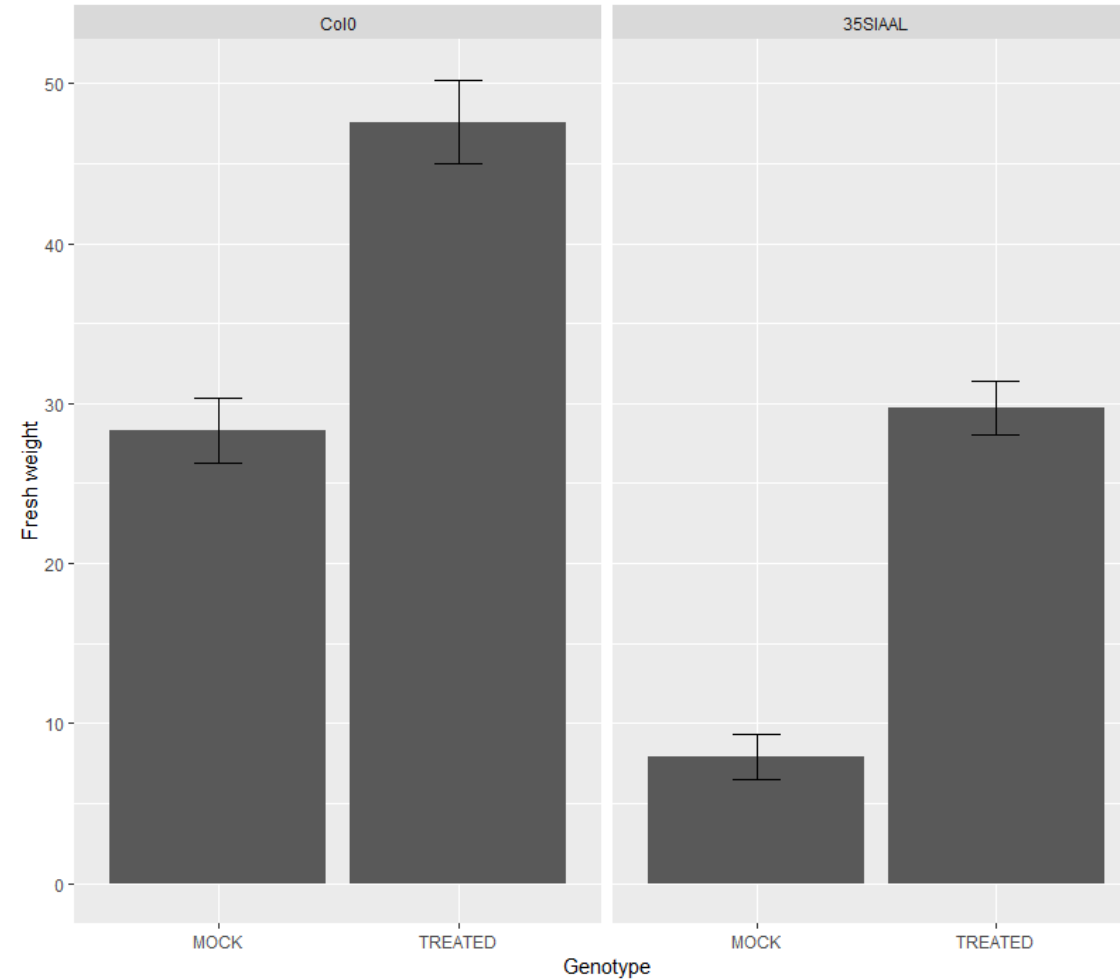
$$\begin{array}{ccc} 2 & \begin{array}{c} \text{18} \\ \text{—————} \\ \text{2x18} \end{array} & = 6 \begin{array}{c} \text{6} \\ \text{□} \\ \text{6x6} \end{array} \end{array}$$

$$\sqrt[n]{\prod_{i=1}^n x_i} = \exp\left(\frac{\sum_{i=1}^n \log_e(x_i)}{n}\right)$$

- ▶ In small samples it is tricky to determine normality one way or another.
- ▶ The consequences for the statistical model of applying the 'wrong' transformation could be worse than the consequences of analysing the untransformed scores.



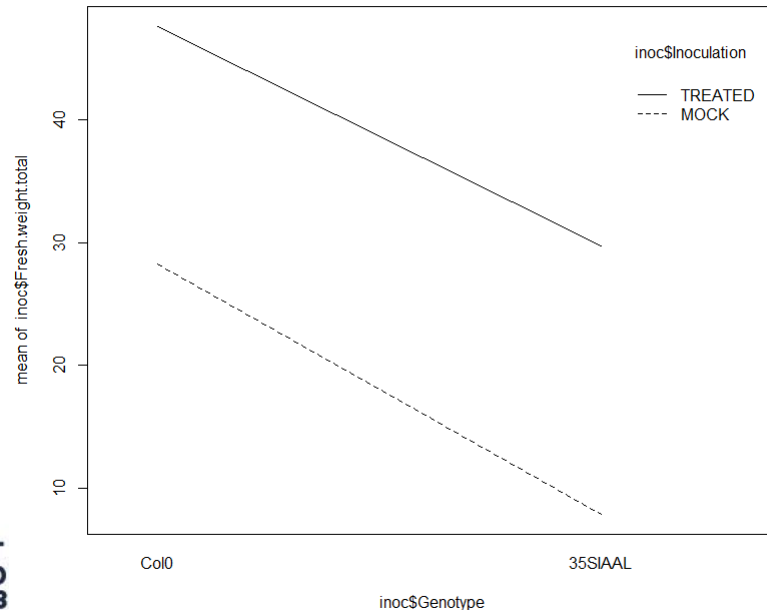
# Effect log transformation



# Effect log transformation

## Original scale

$H_0: \text{Col T} - \text{Col M} = 35\text{S T} - 35\text{S M}$



## Log scale

$H_0: \log(\text{Col T}) - \log(\text{Col M})$

$= \log(35\text{S T}) - \log(35\text{S M})$

After backtransformation:  $\text{Col T}/\text{Col M} = 35\text{S T}/35\text{S M}$

