

Single Group Analyses		Single Group Experiments	
T-test	Z-test	Single Sample Design	
	T-test	Pre-Post Designs	
Mixed-effects models		Repeated Measures Designs	
Multi-Group Analyses		Multi-Group Experiments	
	T-test	Two Group Design	
Analysis of Variance		Multi-Group Design	
Regression Analysis		Continuous Predictor Design	
Mixed Analyses		Mixed Experiments	
Mixed-effects models		Group Design with Repeated Measurements	
Analysis of Covariance		Group Design also measured on a continuous variable	

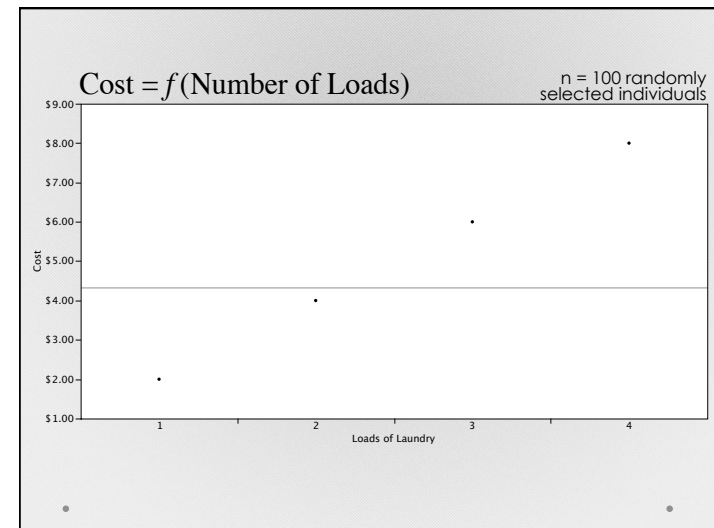
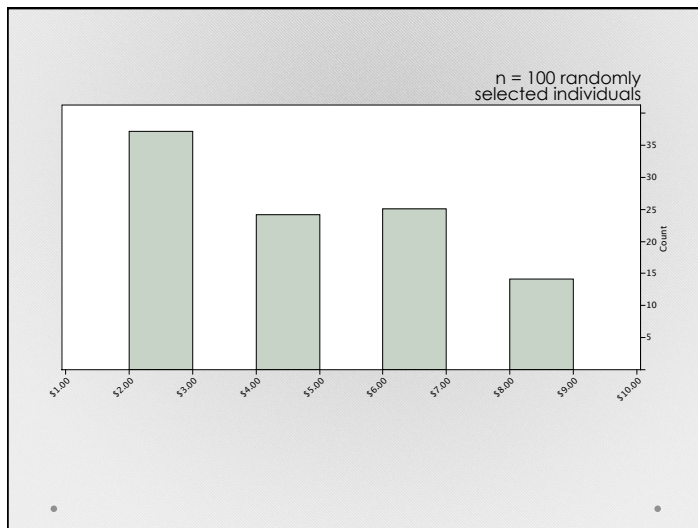
Single Group Analyses		Single Group Experiments	
T-test	Z-test	Single Sample Design	
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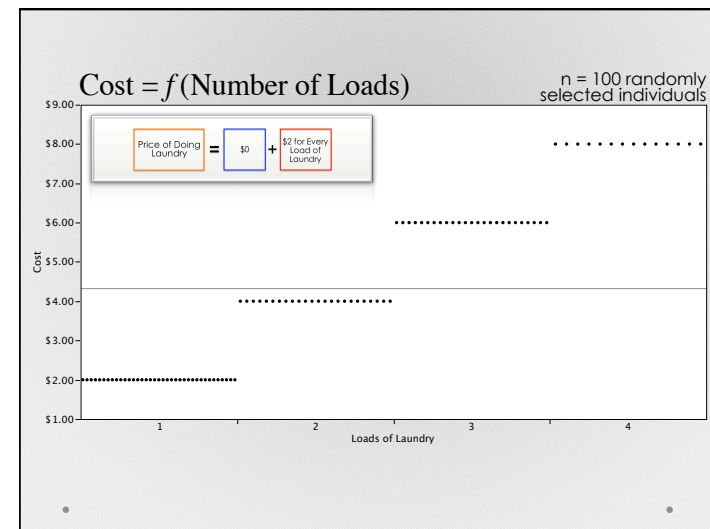
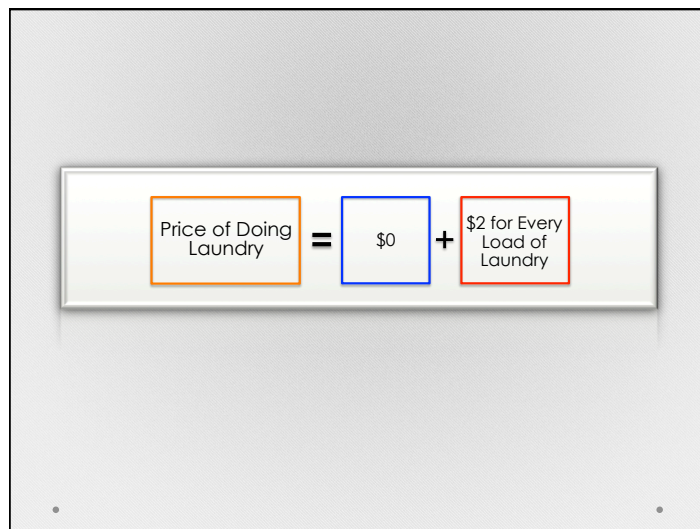
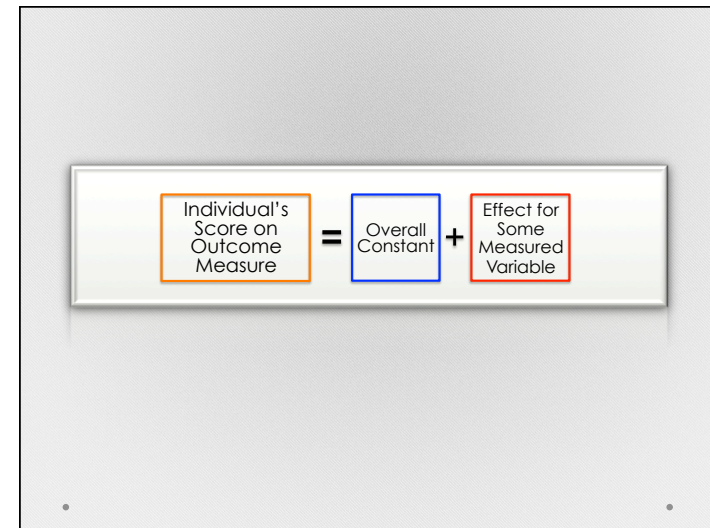
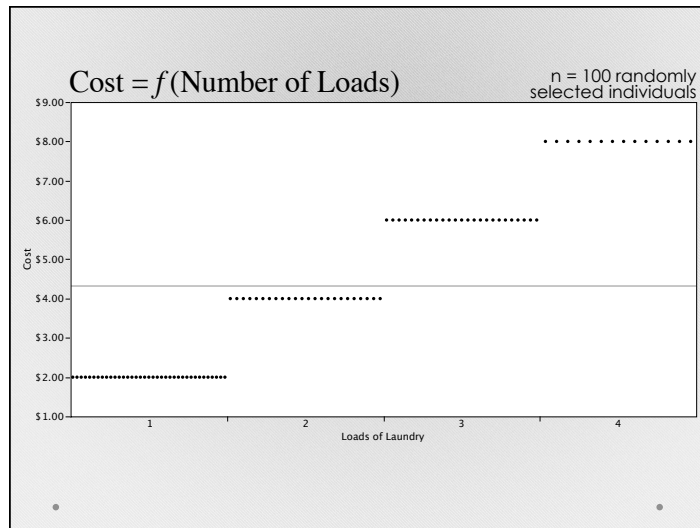
Single Group Analyses		Single Group Experiments	
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Mixed Analyses		Mixed Experiments	
Mixed-effects models		Group Design with Repeated Measurements	
Analysis of Covariance		Group Design also measured on a continuous variable	

Mathematical Models	
<ul style="list-style-type: none"> Extend our descriptions of nature to situations with more than two groups Are intended to be parsimonious descriptions of effects and relationships (no more complicated than necessary) Can describe <i>functional</i> and also <i>statistical</i> relationships between variables 	

Modeling Functional Relationships

Price of Doing Laundry

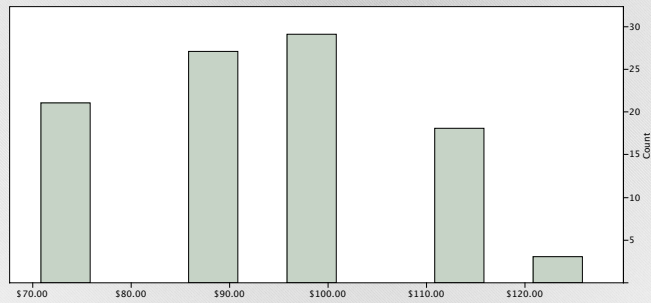




Halloween at a downtown club

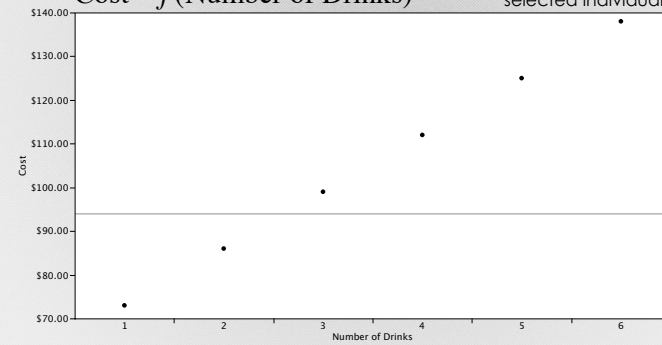


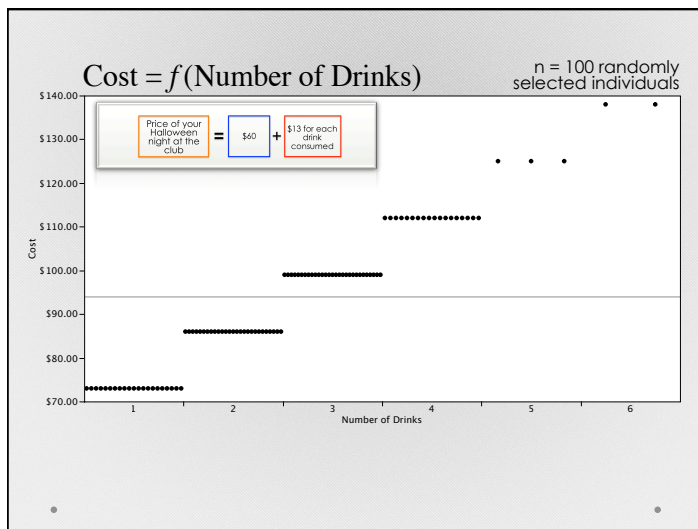
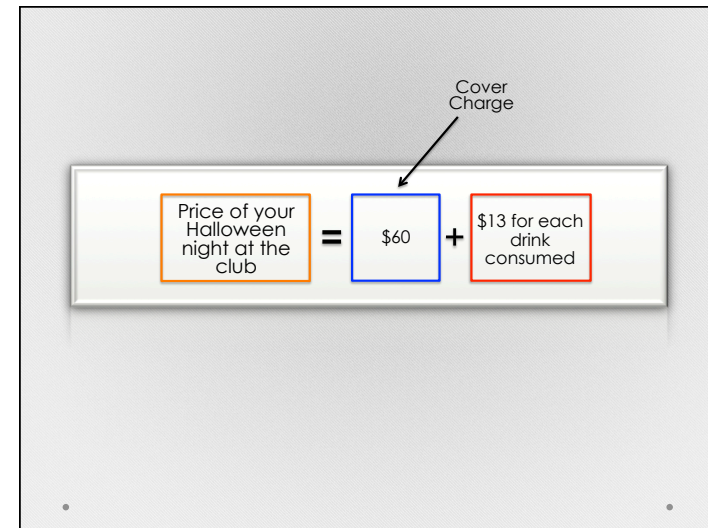
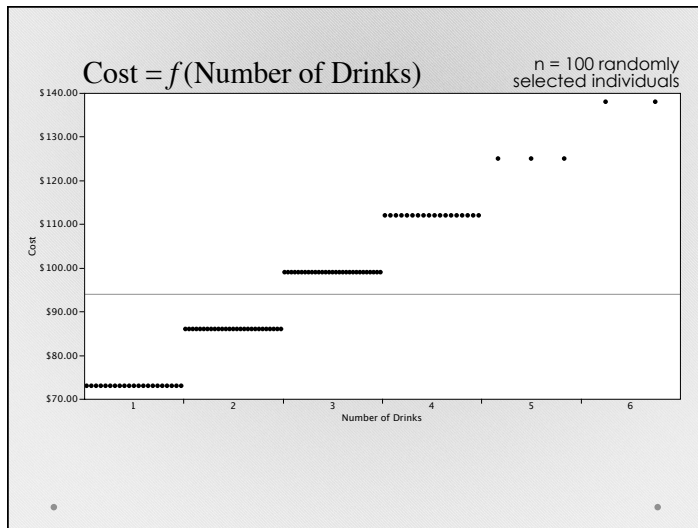
n = 100 randomly
selected individuals



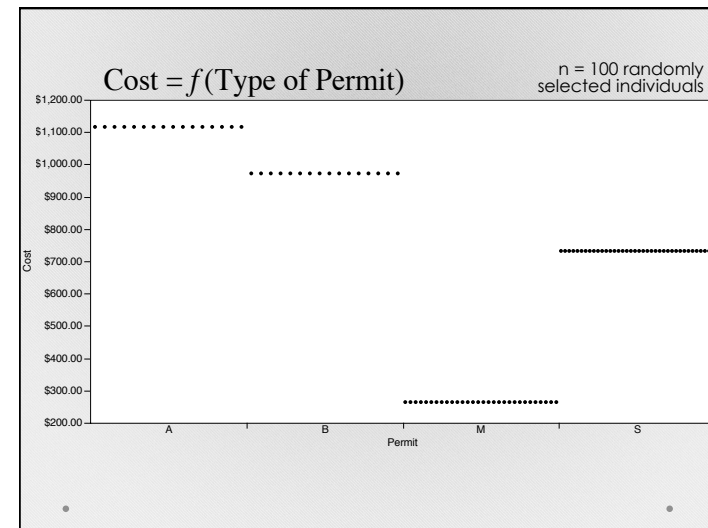
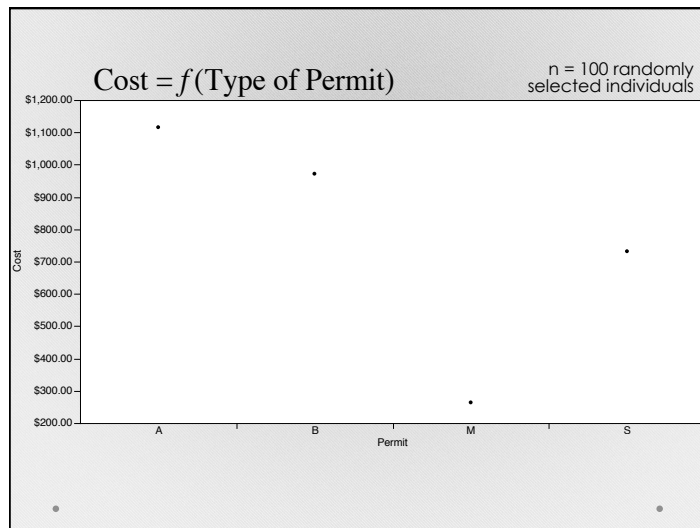
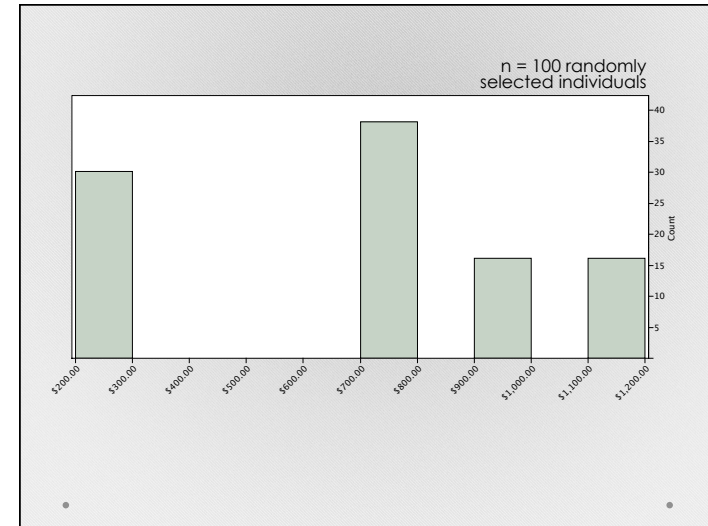
Cost = f (Number of Drinks)

n = 100 randomly
selected individuals

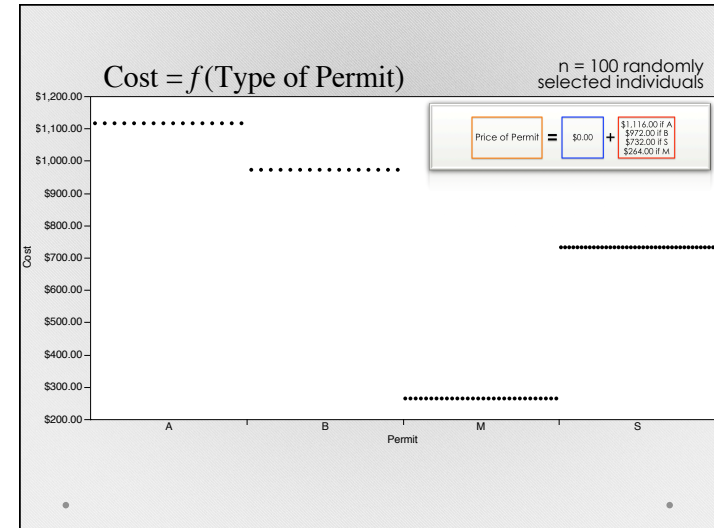




Parking Permit Cost



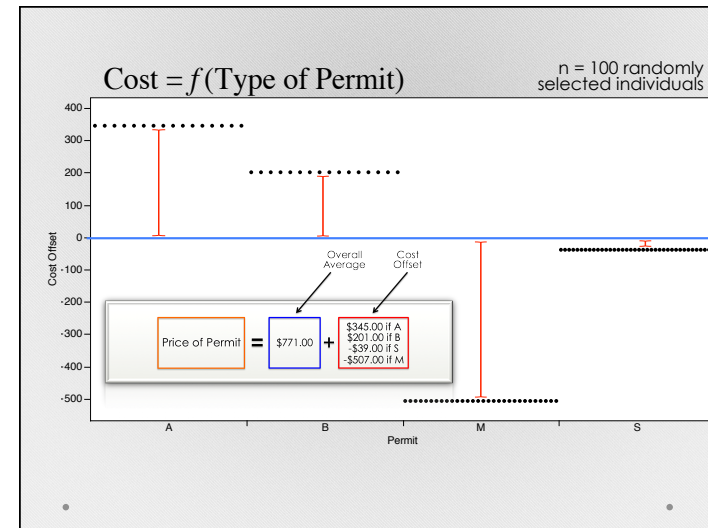
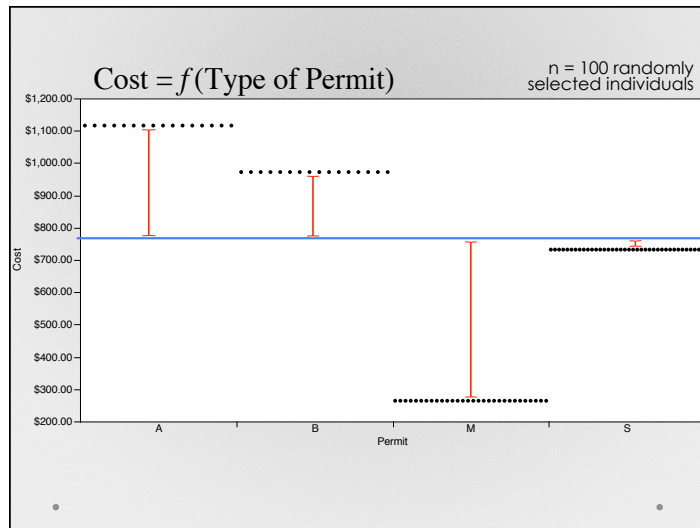
$$\text{Price of Permit} = \$0.00 + \begin{cases} \$1,116.00 \text{ if A} \\ \$972.00 \text{ if B} \\ \$732.00 \text{ if S} \\ \$264.00 \text{ if M} \end{cases}$$



$$\text{Price of Permit} = \$0.00 + \begin{cases} \$1,116.00 \text{ if A} \\ \$972.00 \text{ if B} \\ \$732.00 \text{ if S} \\ \$264.00 \text{ if M} \end{cases}$$

Overall Average Cost Offset

$$\text{Price of Permit} = \$771.00 + \begin{cases} \$345.00 \text{ if A} \\ \$201.00 \text{ if B} \\ -\$39.00 \text{ if S} \\ -\$507.00 \text{ if M} \end{cases}$$

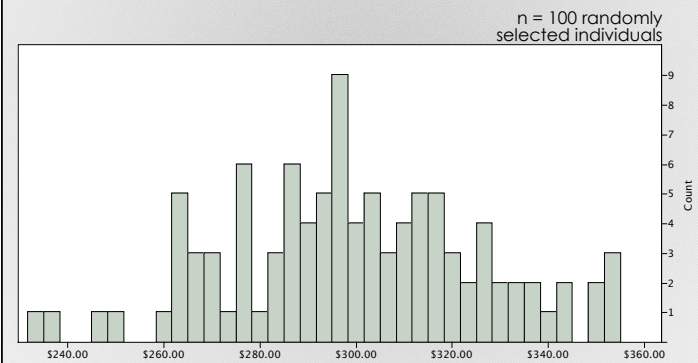


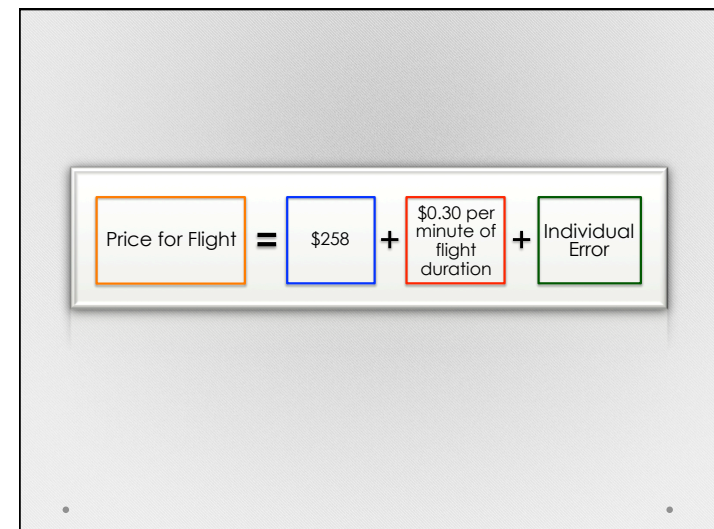
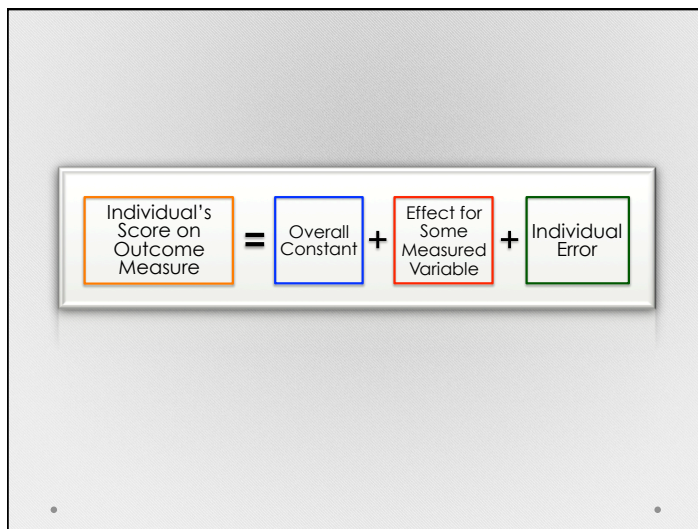
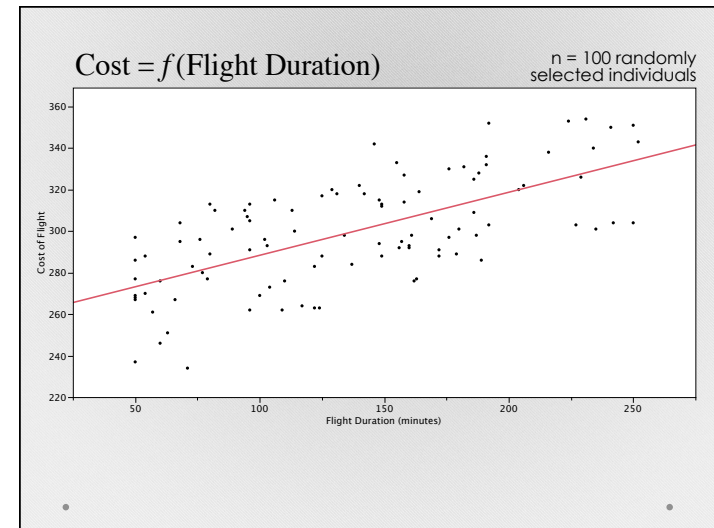
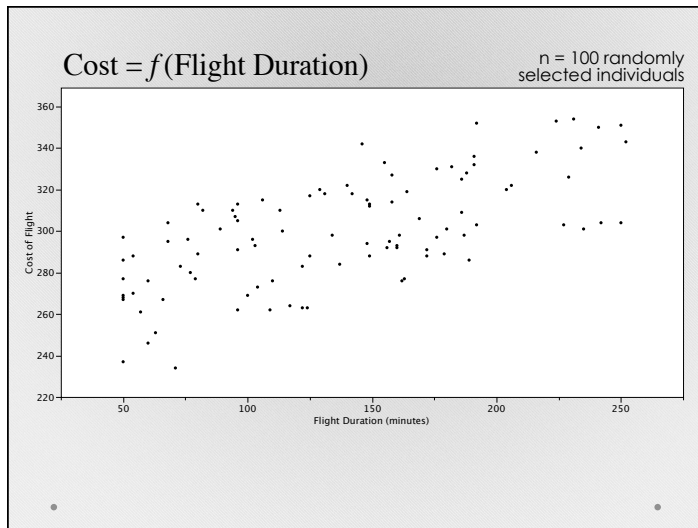
Modeling Statistical Relationships

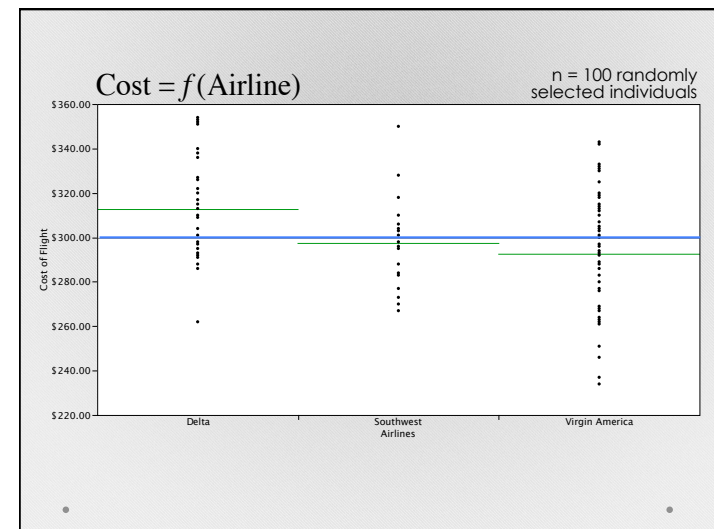
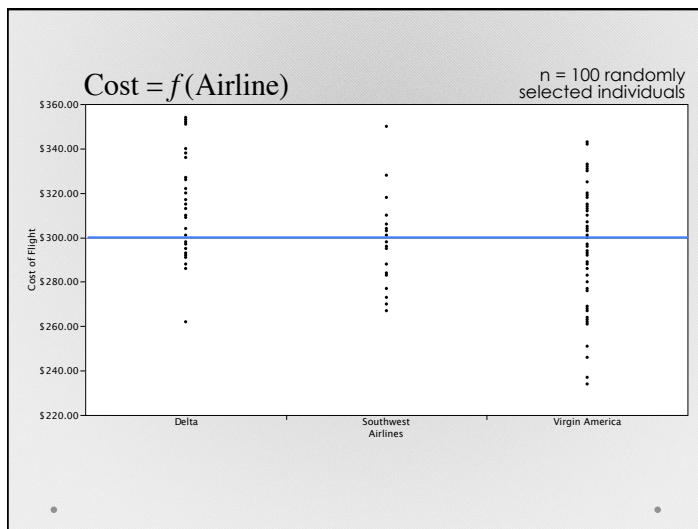
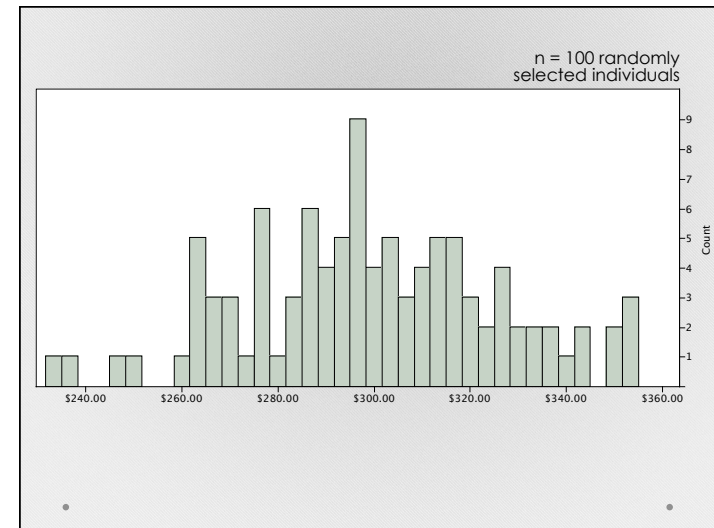
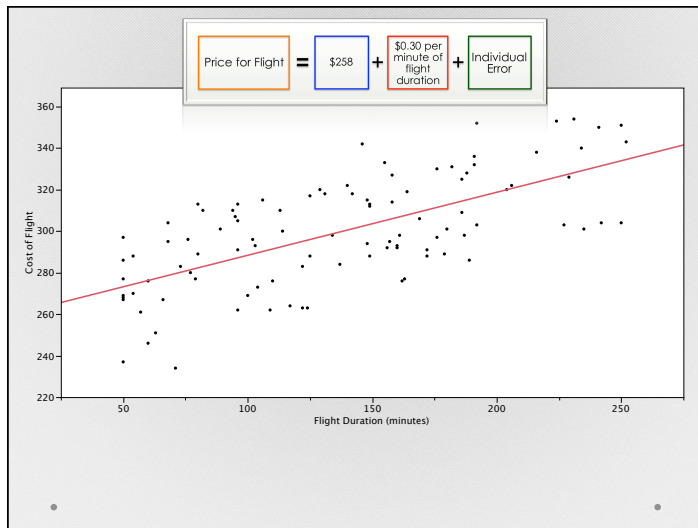
$$\text{Individual's Score on Outcome Measure} = \text{Overall Constant} + \text{Effect for Some Measured Variable}$$

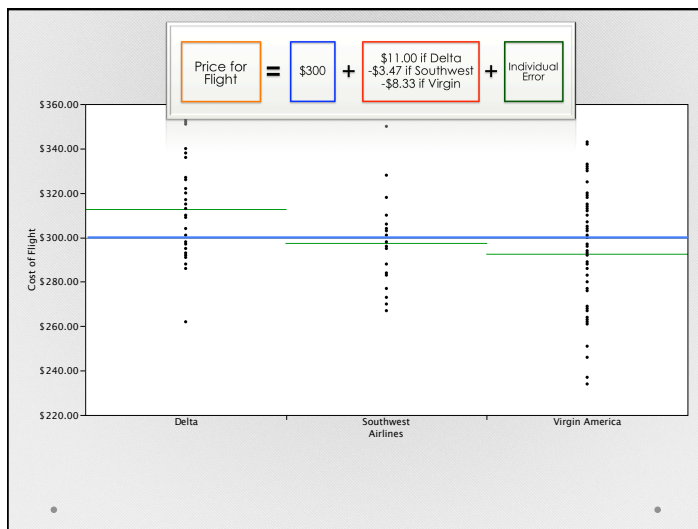
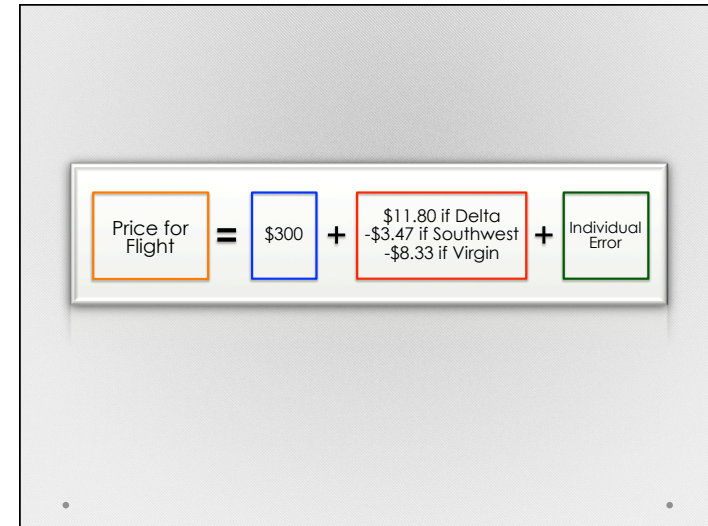
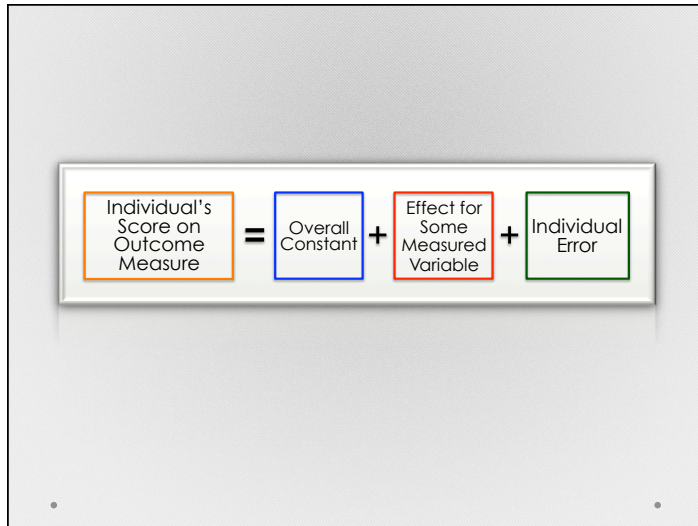
$$\text{Individual's Score on Outcome Measure} = \text{Overall Constant} + \text{Effect for Some Measured Variable} + \text{Individual Error}$$

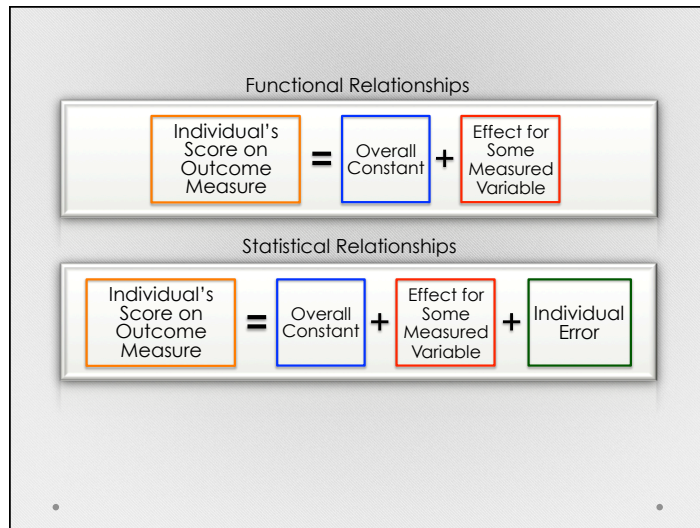
Flight Costs











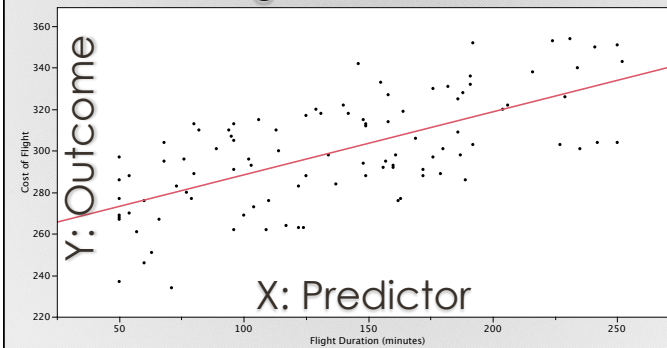
The General Linear Model

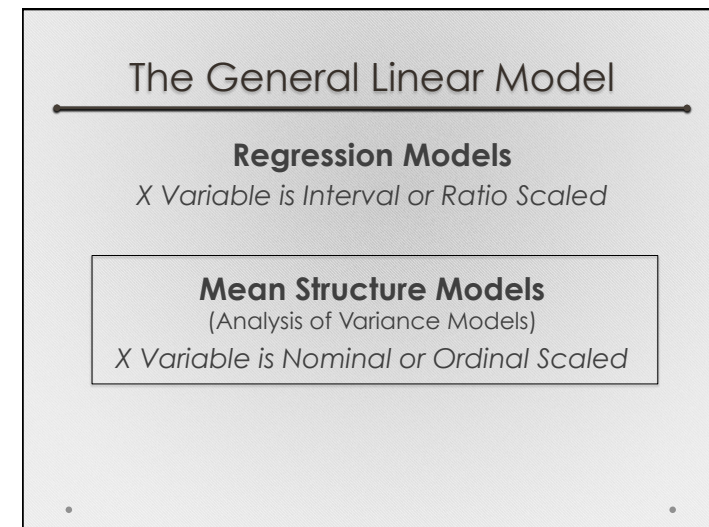
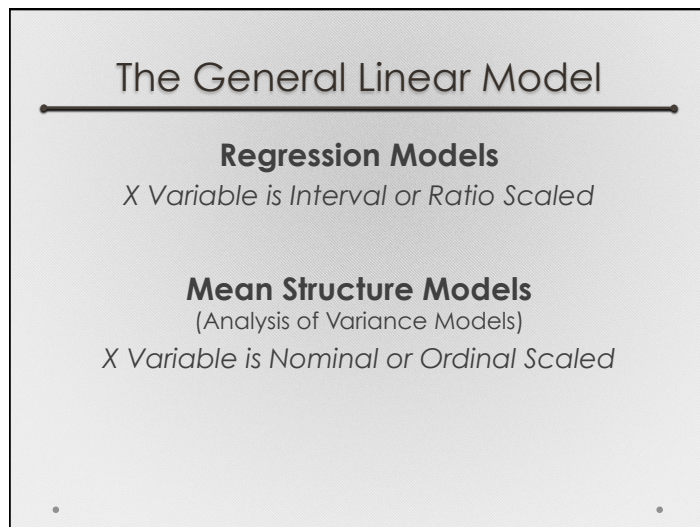
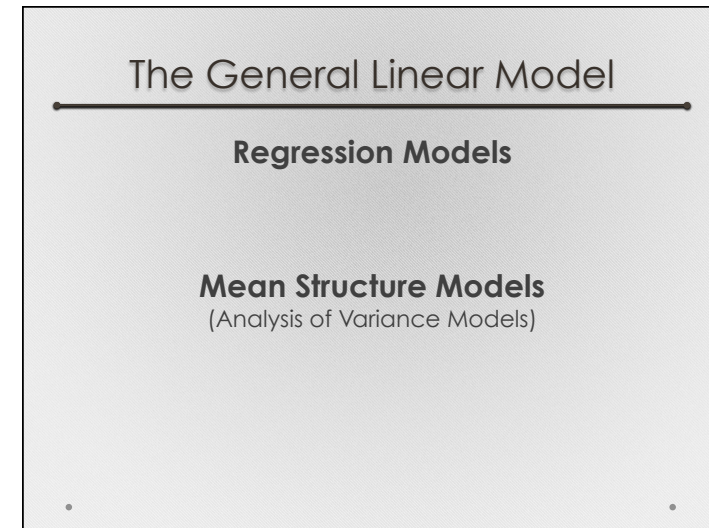
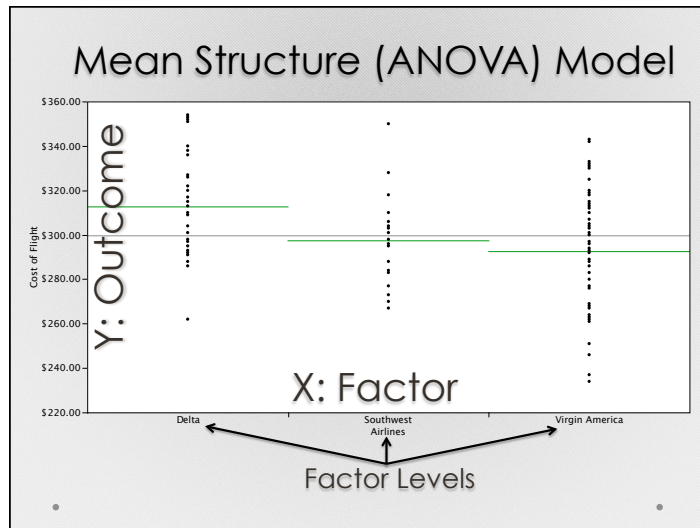
The General Linear Model

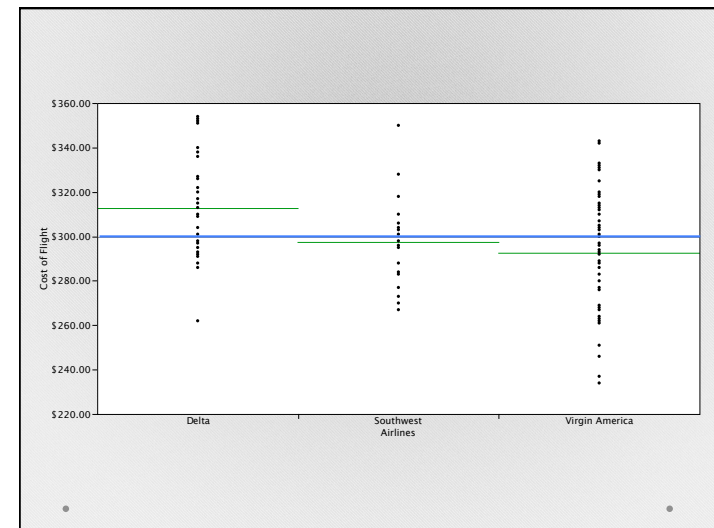
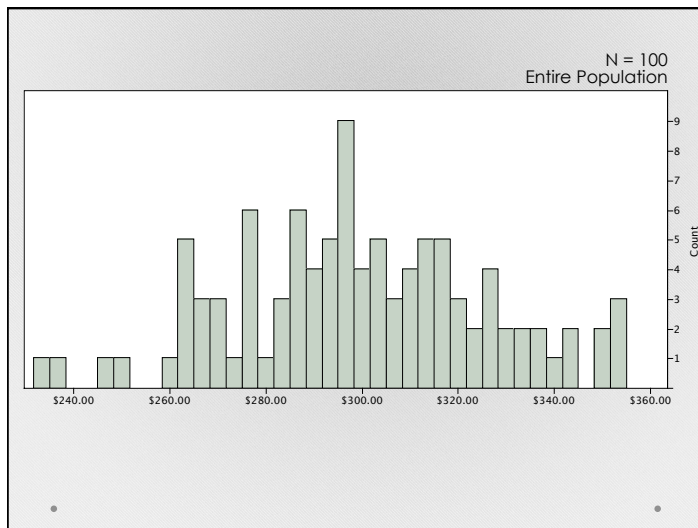
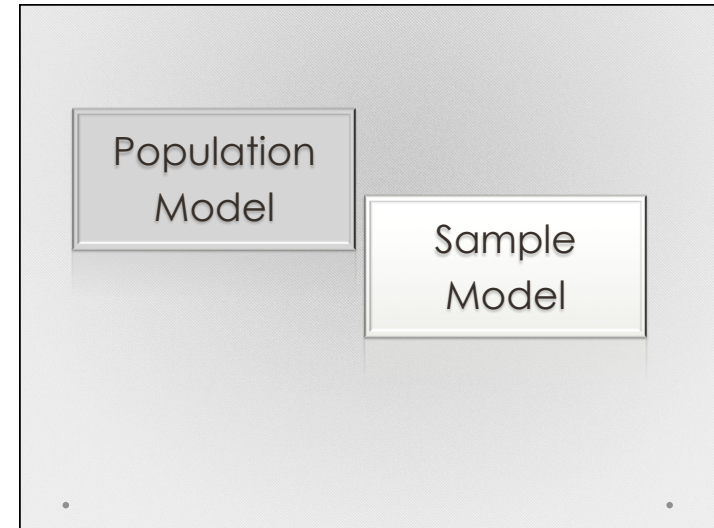
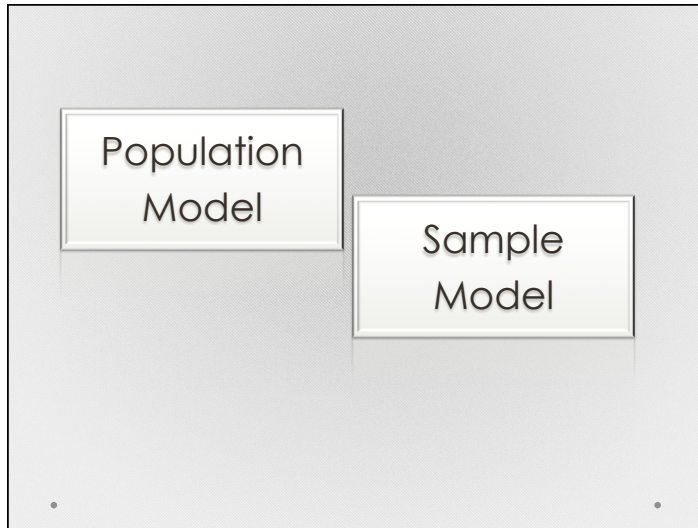
Regression Models

Mean Structure Models (Analysis of Variance Models)

Regression Model







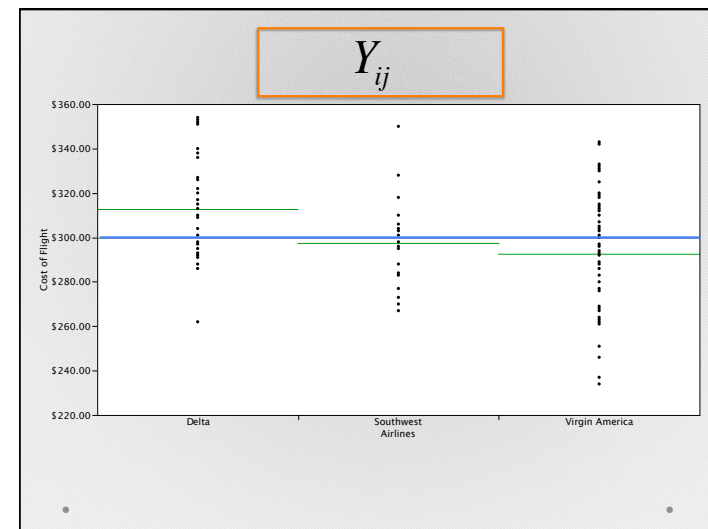
$$\text{Price for Flight} = \$300 + \begin{matrix} \$11.80 \text{ if Delta} \\ -\$3.47 \text{ if Southwest} \\ -\$8.33 \text{ if Virgin} \end{matrix} + \text{Individual Error}$$

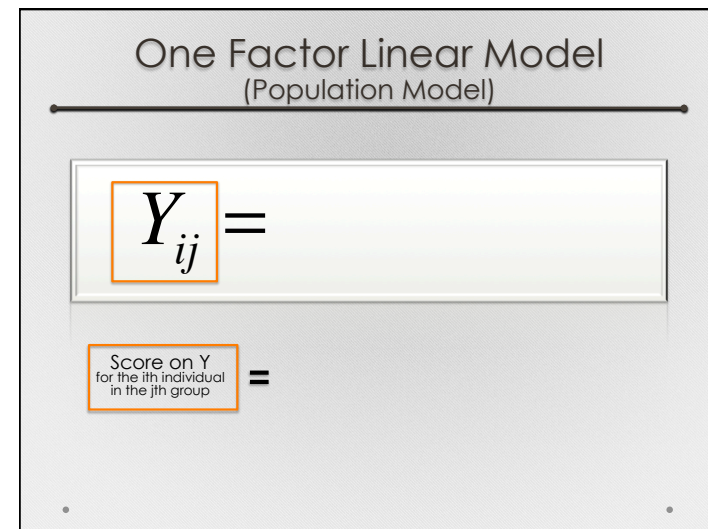
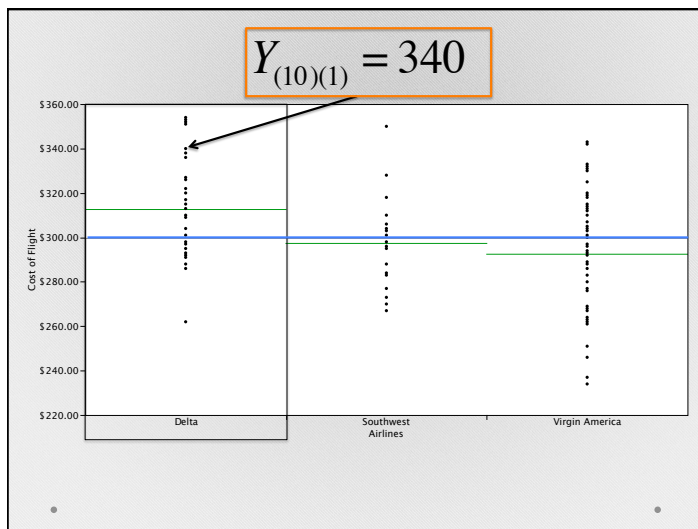
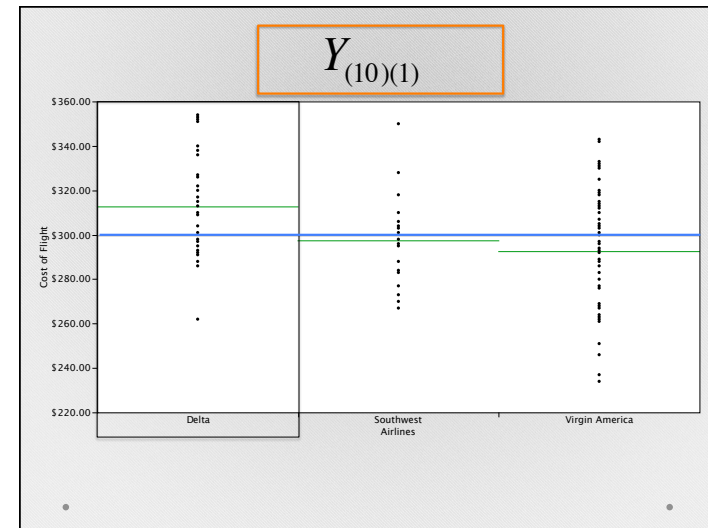
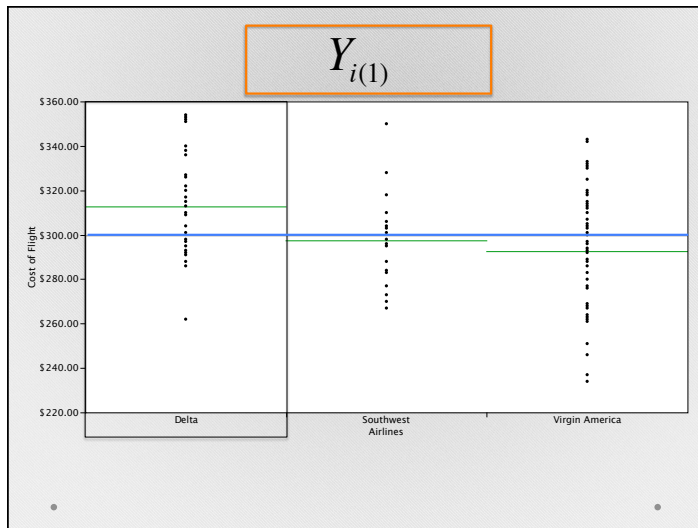
One Factor Linear Model (Population Model)

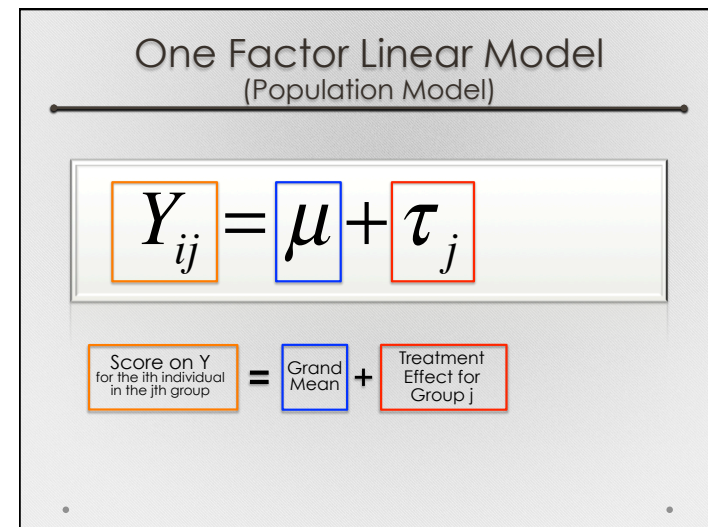
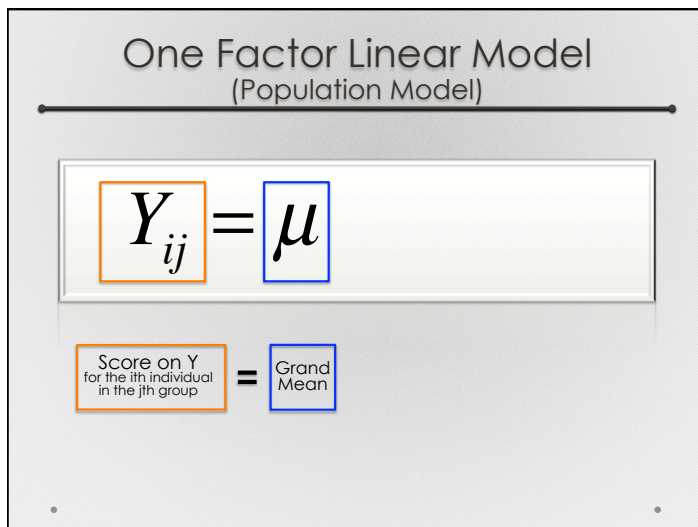
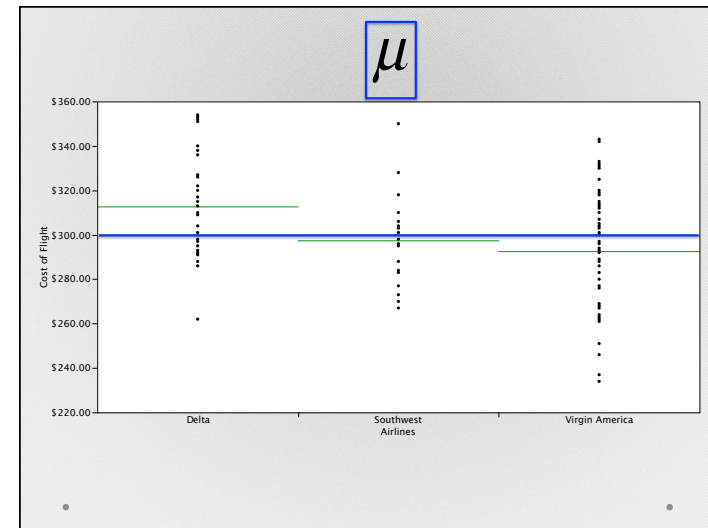
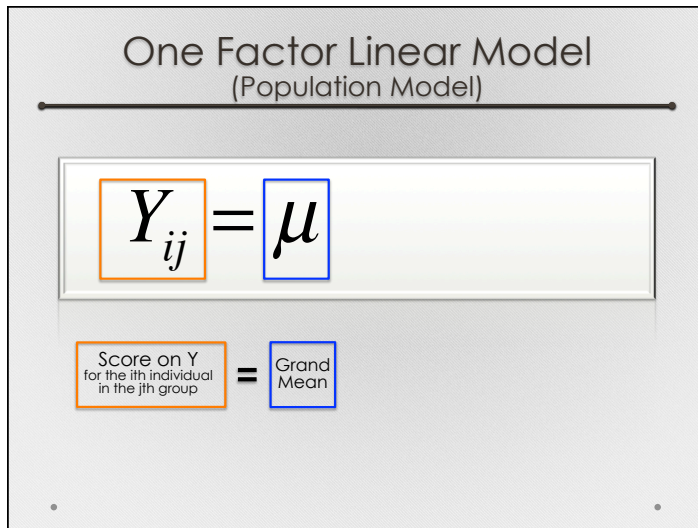
One Factor Linear Model (Population Model)

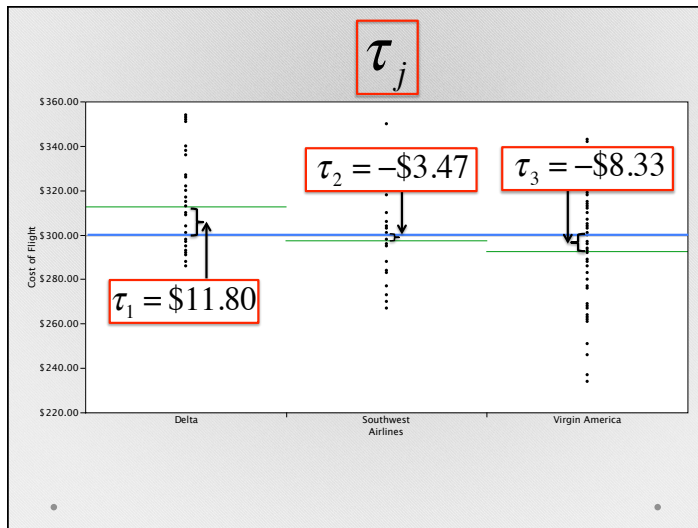
$$Y_{ij} =$$

Score on Y
for the i th individual
in the j th group









Delta
 $\tau_1 = \$11.80$

Southwest
 $\tau_2 = -\$3.47$

Virgin America
 $\tau_3 = -\$8.33$

One Factor Linear Model (Population Model)

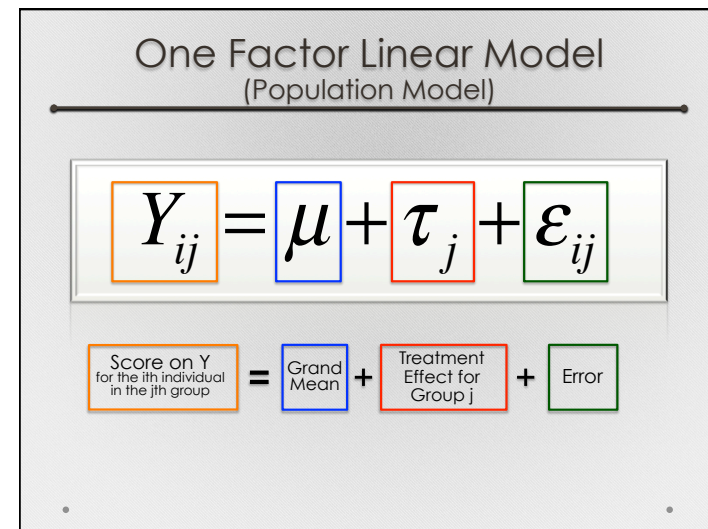
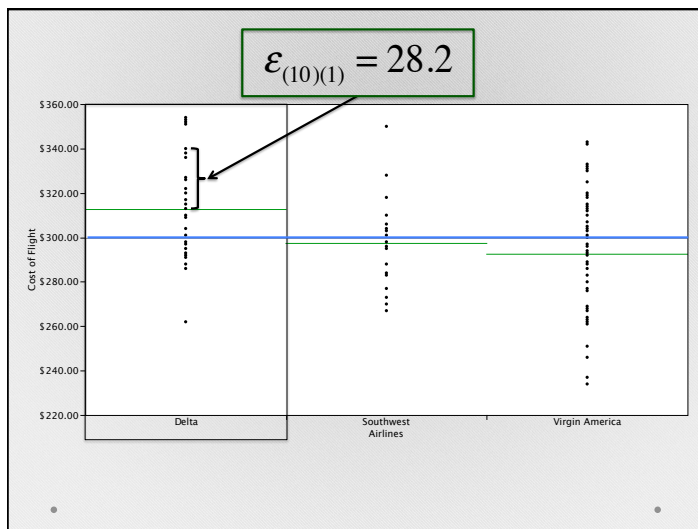
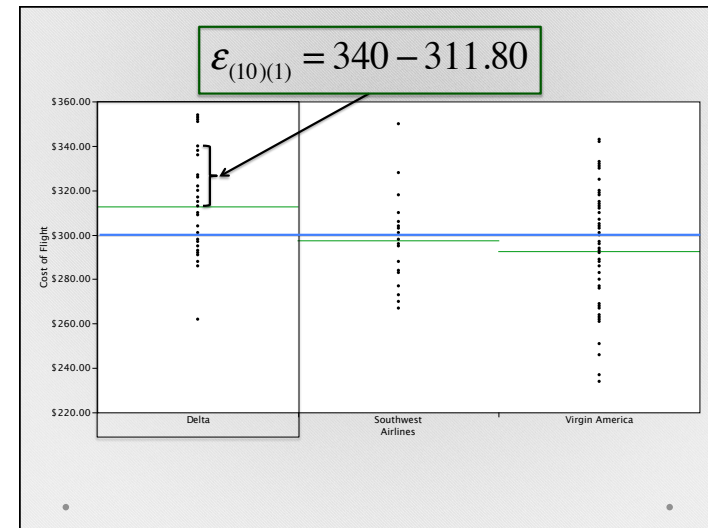
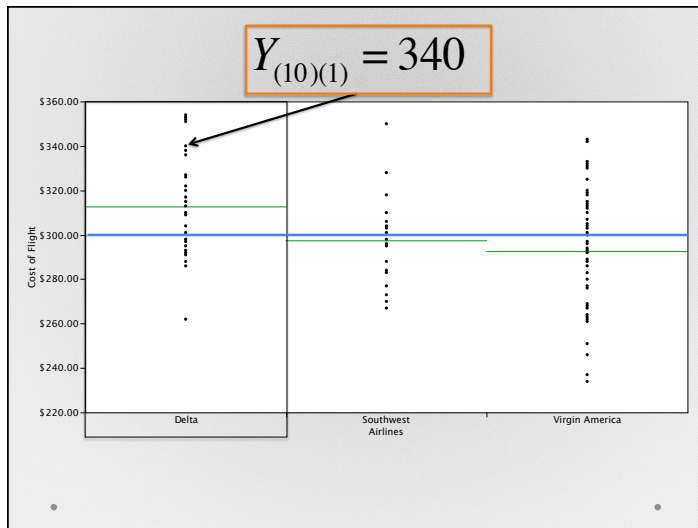
$$Y_{ij} = \mu + \tau_j$$

Score on Y for the i th individual in the j th group = Grand Mean + Treatment Effect for Group j

One Factor Linear Model (Population Model)

$$Y_{ij} = \mu + \tau_j + \epsilon_{ij}$$

Score on Y for the i th individual in the j th group = Grand Mean + Treatment Effect for Group j + Error



For a Person on Delta

$$Y_i = \$300 + \$11.80 + \varepsilon_i$$

For a Person on Southwest

$$Y_i = \$300 - \$3.47 + \varepsilon_i$$

For a Person on Virgin America

$$Y_i = \$300 - \$8.33 + \varepsilon_i$$

One Factor Linear Model (Population Model)

$$Y_{ij} = \mu + \tau_j + \varepsilon_{ij}$$

Score on Y
for the i th individual
in the j th group

$$= \text{Grand Mean} + \text{Treatment Effect for Group } j + \text{Error}$$

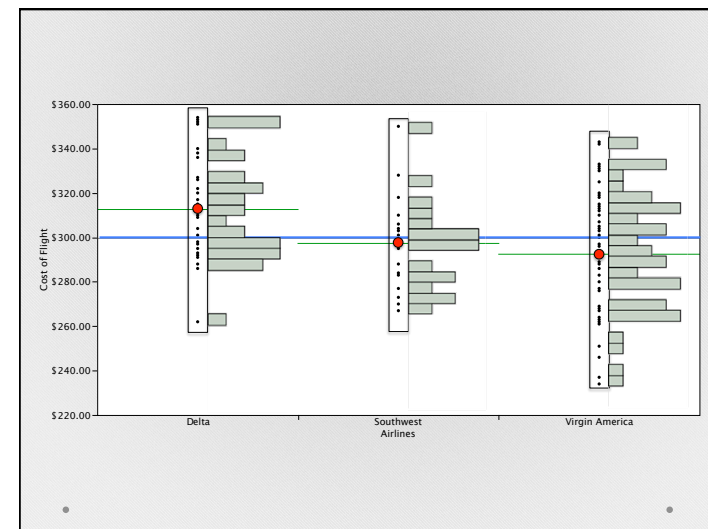
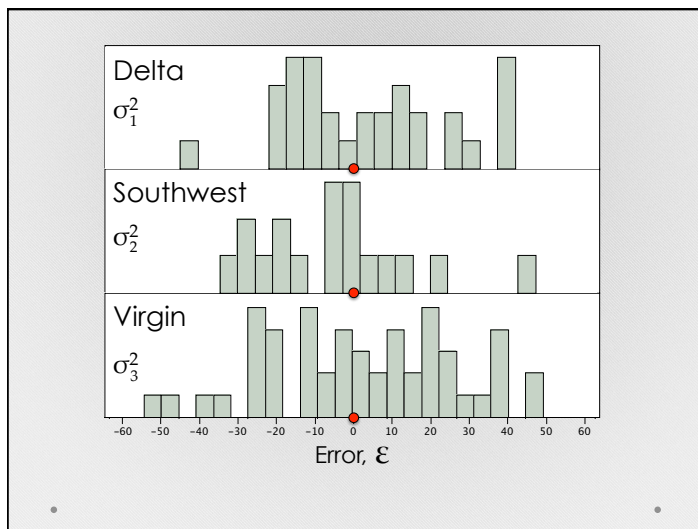
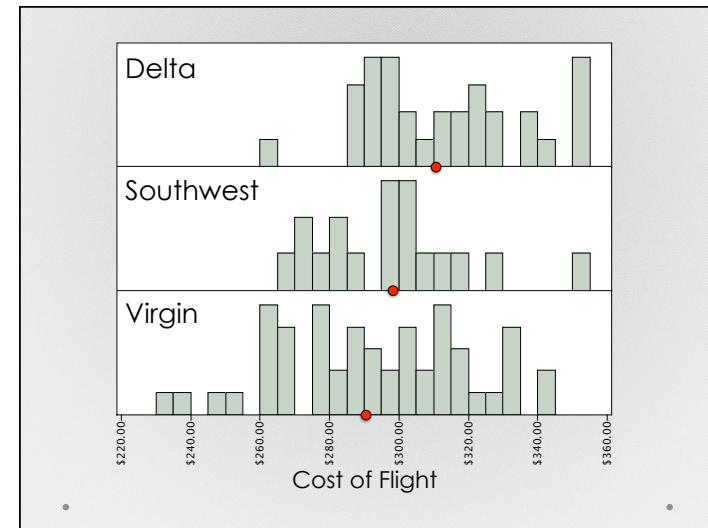
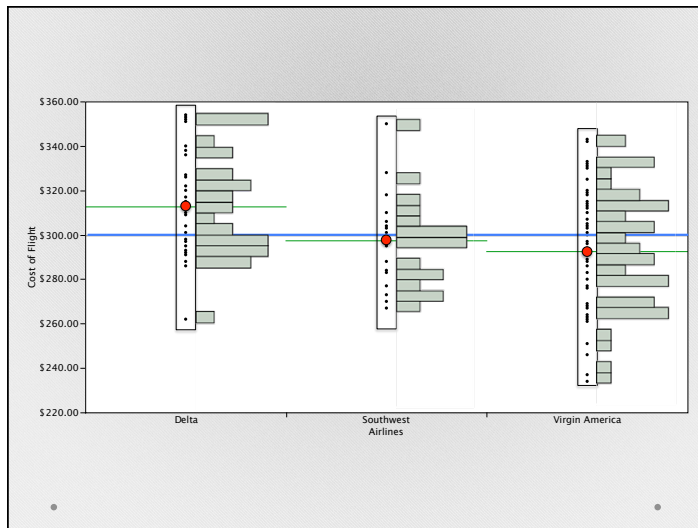
One Factor Linear Model (Population Model)

$$Y_{ij} = \mu + \tau_j + \varepsilon_{ij}$$

Score on Y
for the i th individual
in the j th group

$$= \text{Grand Mean} + \text{Treatment Effect for Group } j + \text{Error}$$

The Distribution of ε



One Factor Linear Model
(Population Model)

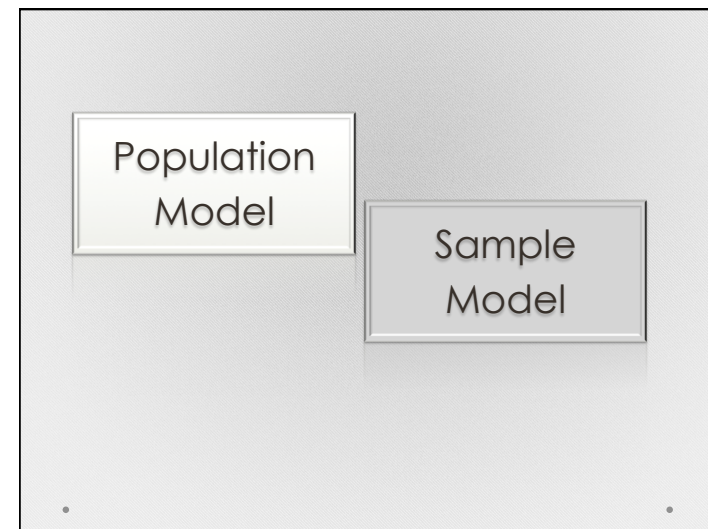
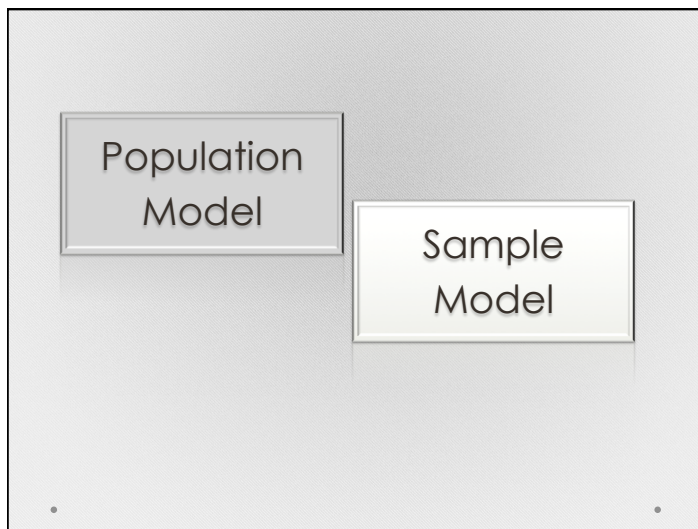
$$Y_{ij} = \mu + \tau_j + \varepsilon_{ij}$$

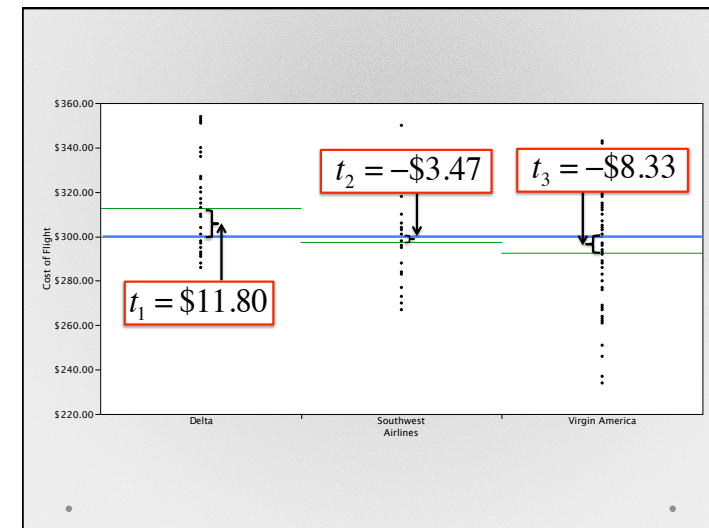
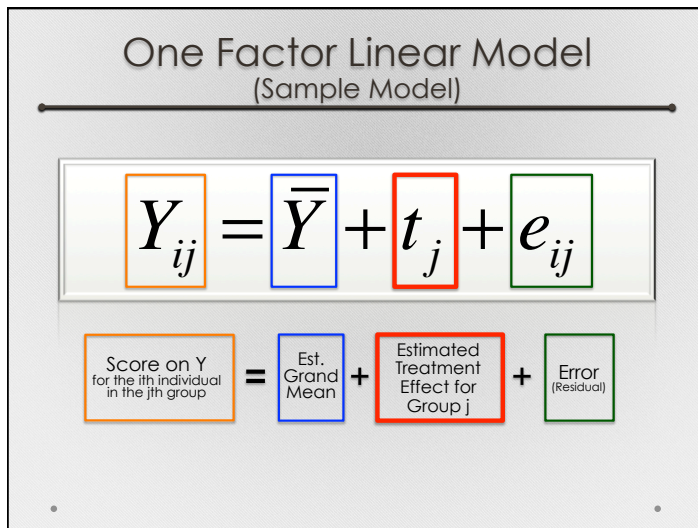
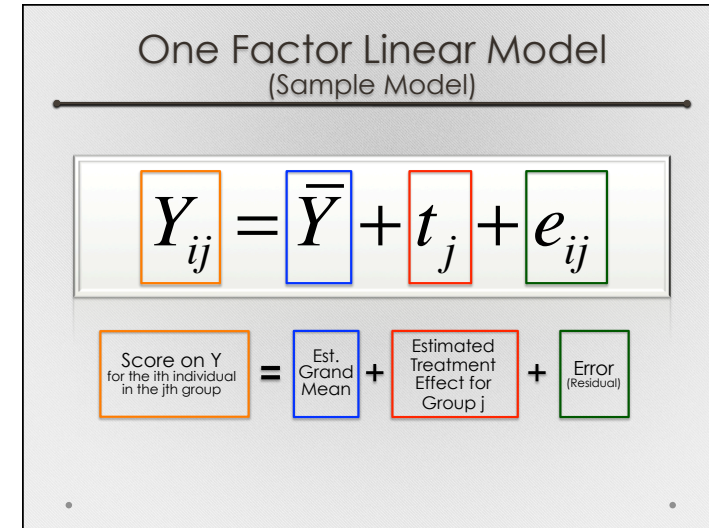
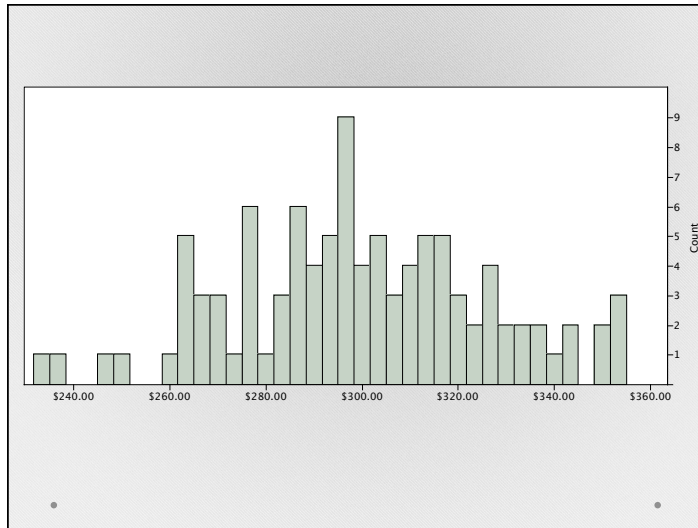
Score on Y for the i th individual in the j th group = Grand Mean + Treatment Effect for Group j + Error

One Factor Linear Model
(Population Model)

$$Y_{ij} = \mu + \tau_j + \varepsilon_{ij}$$

Score on Y for the i th individual in the j th group = Grand Mean + Treatment Effect for Group j + Error





One Factor Linear Model (Sample Model)

$$Y_{ij} = \bar{Y} + t_j + e_{ij}$$

Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j + Error (Residual)

One Factor Linear Model (Sample Model)

$$Y_{ij} = \bar{Y} + t_j + e_{ij}$$

Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j + Error (Residual)

One Factor Linear Model (Sample Model)

$$Y_{ij} = \bar{Y} + t_j + e_{ij}$$

Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j + Error (Residual)

"Residual" Error

$$e_{ij} = Y_{ij} - \hat{Y}_{ij}$$

An individual's residual is the difference between that individual's **ACTUAL** observed value and the value predicted by the model

One Factor Linear Model
(Sample Model – Prediction for an Individual)

$$\hat{Y}_{ij} = \bar{Y} + t_j$$

Predicted Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j

One Factor Linear Model
(Sample Model – Prediction for an Individual)

$$\hat{Y}_{ij} = \bar{Y}_j$$

Predicted Score on Y for the i th individual in the j th group = Sample Mean for Group j

One Factor Linear Model
(Sample Model – Prediction for an Individual)

$$\hat{Y}_{ij} = \bar{Y}_{\cdot j}$$

Predicted Score on Y for the i th individual in the j th group = Sample Mean for Group j

One Factor Linear Model
(Sample Model – Prediction for an Individual)

$$\hat{Y}_{ij} = \bar{Y} + t_j$$

Predicted Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j

"Residual" Error

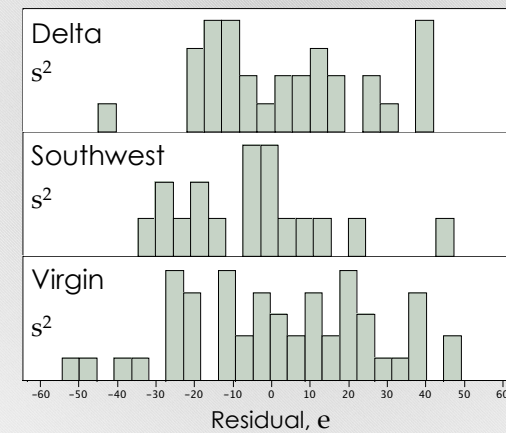
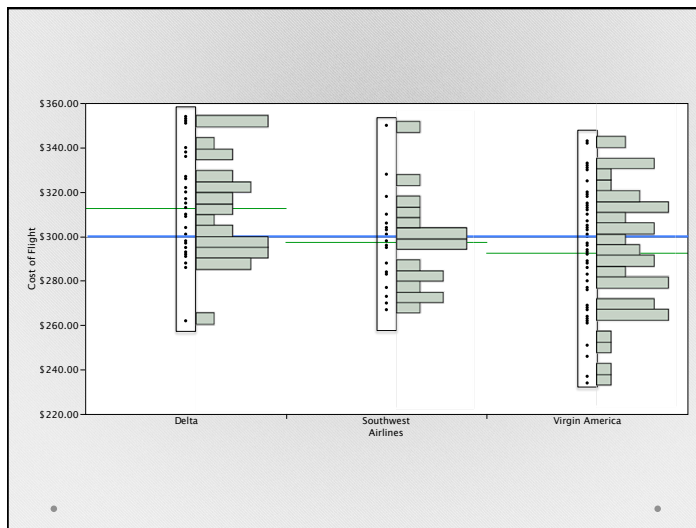
$$e_{ij} = Y_{ij} - \hat{Y}_{ij}$$

An individual's residual is the difference between that individual's **ACTUAL** observed value and the value predicted by the model

"Residual" Error

$$e_{ij} = Y_{ij} - (\bar{Y} + t_j)$$

An individual's residual is the difference between that individual's **ACTUAL** observed value and the value predicted by the model



FROM BEFORE

Calculating the Variance

Population

$$SS = \sum (X_i - \mu)^2 \Rightarrow \sigma^2 = \frac{SS}{N}$$

Sample

$$SS = \sum (X_i - \bar{X})^2 \Rightarrow s^2 = \frac{SS}{n-1}$$

FROM BEFORE

Calculating the Variance

Population

$$SS = \sum (X_i - \mu)^2 \Rightarrow \sigma^2 = \frac{SS}{N}$$

Sample

$$SS = \sum (X_i - \bar{X})^2 \Rightarrow s^2 = \frac{SS}{\boxed{n-1}}$$

Calculating the Variance

Population

$$SS = \sum (X_i - \mu)^2 \Rightarrow \sigma^2 = \frac{SS}{N}$$

Sample

$$SS = \sum (X_i - \bar{X})^2 \Rightarrow s^2 = \frac{SS}{df}$$

Calculating the Variance

Population

$$SS = \sum (X_i - \mu)^2 \Rightarrow \sigma^2 = \frac{SS}{N}$$

Sample

$$SS = \sum (X_i - \bar{X})^2 \Rightarrow s^2 = \frac{SS}{df}$$

Sums of Squares for Error

$$SS_e = \sum_j \sum_i (Y_{ij} - \hat{Y}_{ij})^2$$

The sums of squares for error is the sum of each individual's squared error

Sums of Squares for Error

$$SS_e = \sum_j \sum_i e_{ij}^2$$

The sums of squares for error is the sum of each individual's squared error

Mean Squared Error

$$MS_e = s_p^2 = \frac{SS_e}{df_e}$$

The mean squared error is the mean of the squared error of each individual

Mean Squared Error Estimates σ^2

$$MS_e \hat{=} \sigma^2$$

The mean squared error is an unbiased estimator of the variance in the population

One Factor Linear Model
(Sample Model)

$$Y_{ij} = \bar{Y} + t_j + e_{ij}$$

Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j + Error (Residual)

One Factor Linear Model
(Sample Model)

$$Y_{ij} = \bar{Y} + t_j + e_{ij}$$

Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j + Error (Residual)

One Factor Linear Model
(Sample Model)

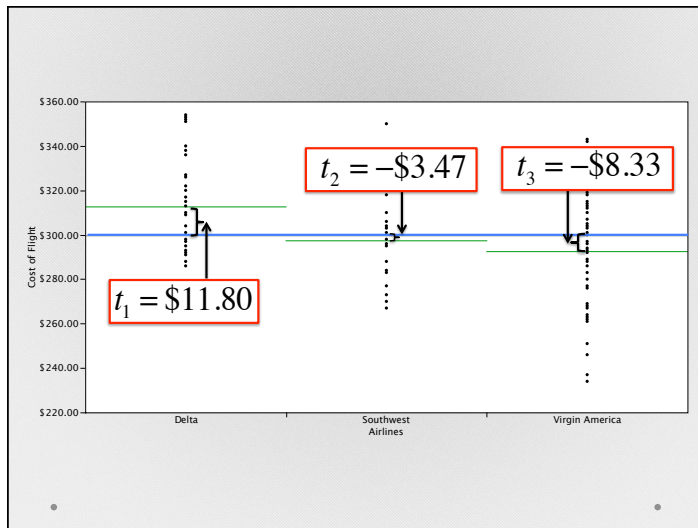
$$Y_{ij} = \bar{Y} + t_j + e_{ij}$$

Score on Y for the i th individual in the j th group = Est. Grand Mean + Estimated Treatment Effect for Group j + Error (Residual)

Sample Treatment Effects Estimate True Treatment Effects

$$t_j \hat{=} \tau_j$$

The treatment effects of a sample model are unbiased estimators of the true treatment effects in the population



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Inferences about Treatment Effects

H_0 : In the population the cost of flying is the same for these airlines

H_1 : In the population the cost of flying is *not* the same for these airlines

Inferences about Treatment Effects

$$H_0 : \tau_1 = \tau_2 = \dots = \tau_j = 0$$

$$H_1 : \text{Not All } \tau_j = 0$$

Statistical Inference with Linear Models

- Analysis of Variance Approach
- General Linear Test Approach
(model comparison)

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