

STAT 500

Simple Linear Regression: ANOVA Table and R^2

Regression Analysis: ANOVA

- Write the deviation from the overall sample mean as $Y_i - \bar{Y} = (Y_i - \hat{Y}_i) + (\hat{Y}_i - \bar{Y})$ where $\hat{Y}_i = b_0 + b_1 X_i$
- Partition the corrected total sums of squares

$$\begin{aligned} SS_{corrected\ total} &= \sum_i (Y_i - \bar{Y})^2 = \sum_i (Y_i - \hat{Y}_i + \hat{Y}_i - \bar{Y})^2 \\ &= \sum_i (Y_i - \hat{Y}_i)^2 + \sum_i (\hat{Y}_i - \bar{Y})^2 \\ &\quad + 2 \sum_i (Y_i - \hat{Y}_i)(\hat{Y}_i - \bar{Y}) \\ &= \sum_i (Y_i - \hat{Y}_i)^2 + \sum_i (\hat{Y}_i - \bar{Y})^2 \\ &= SS_{residuals} + SS_{model} \end{aligned}$$

Regression Analysis: ANOVA

- Cross product term is

$$\begin{aligned}2 \sum_i (Y_i - \hat{Y}_i)(\hat{Y}_i - \bar{Y}) &= 2 \sum_i e_i(b_o + b_1 x_i - \bar{Y}) \\&= 2(b_o - \bar{Y}) \sum_i e_i + 2b_1 \sum_i e_i x_i \\&= 0 \quad \text{because } \sum_i e_i = \sum_i e_i x_i = 0\end{aligned}$$

Note that

$$\begin{aligned}SS_{model} &= \sum_i (\hat{Y}_i - \bar{Y})^2 = \sum_i (b_o + b_1 x_i - \bar{Y})^2 \\&= b_1^2 \sum_{i=1}^n (x_i - \bar{x})^2\end{aligned}$$

Regression Analysis: ANOVA

- $SS_{model} = SS_{total} - SS_{error}$
$$= \sum_i (\hat{Y}_i - \bar{Y})^2$$
$$= \sum_i (b_0 + b_1 x_i - \bar{Y})^2$$
$$= b_1^2 \sum_{i=1}^n (x_i - \bar{x})^2$$
- SS_{model} is also denoted by $SS_{regression}$
- SS_{error} is also denoted by $SS_{residuals}$ or SSE

Regression Analysis: ANOVA

SS_{error} has $n - 2$ degrees of freedom because

- Two parameters must be estimated to calculate \hat{Y}_i
- The residuals satisfy two constraints

$$\sum e_i = 0 \quad \text{and} \quad \sum e_i x_i = 0$$

ANOVA Table

Source	df	Sums of Squares
Model	1	$SS_{\text{model}} = \sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2$
Error	$n - 2$	$SS_{\text{error}} = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$
Total	$n - 1$	$SS_{\text{Total}} = \sum_{i=1}^n (Y_i - \bar{Y})^2$

ANOVA Table: Example

Source	df	SS	MS	F	p-value
Model	1	0.22573	0.22573	2961.55	< 0.0001
Error	15	0.00114	0.00007622		
Total	16	0.22688			

Mean Squares

- MS_{error}

- $\hat{\sigma}^2 = MS_{\text{error}} = SS_{\text{error}}/(n - 2)$
- $\hat{\sigma}^2$ is an unbiased estimate of σ^2

$$E(MS_{\text{error}}) = \sigma^2$$

- MS_{model}

- $E(MS_{\text{model}}) = \sigma^2 + \beta_1^2 \sum_{i=1}^n (x_i - \bar{x})^2$
- When $\beta_1=0$, $E(MS_{\text{model}}) = \sigma^2$.
Otherwise, $E(MS_{\text{model}}) > \sigma^2$.

F-test for Significance of Model

- $H_0 : \beta_1 = 0 \rightarrow Y_i = \beta_0 + \epsilon_i$
- $H_a : \beta_1 \neq 0 \rightarrow Y_i = \beta_0 + \beta_1 x_i + \epsilon_i$
- Test Statistic:

$$F = \frac{MS_{\text{model}}}{MS_{\text{error}}}$$

- Reject H_0 if

$$F = \frac{MS_{\text{model}}}{MS_{\text{error}}} > F_{1, n-2, 1-\alpha}$$

F-test: Example

- $H_0 : \beta_1 = 0$
- $H_a : \beta_1 \neq 0$
- $F = 2961.55$ with p-value < 0.0001 .
- Reject H_0 and conclude there is a significant linear relationship between boiling point of water and log of barometric pressure.

Coefficient of Determination (R^2)

$$R^2 = \frac{SS_{\text{model}}}{SS_{\text{Total}}}$$

- Fraction of variation in the response variable that can be explained by the linear regression model with the explanatory variable x .
- Expressed as percentage: $0\% \leq R^2 \leq 100\%$
- Large values of R^2 indicate better model fit.

R^2 : Example

$$R^2 = \frac{SS_{\text{model}}}{SS_{\text{Total}}} = \frac{0.22573}{0.22688} = 0.9950$$

99.50% of the variation in log(barometric pressure) can be explained by the linear regression model with boiling point of water.

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Simple Linear Regression: Inference for Parameters
and Prediction Intervals

Inference for Model Parameters

- Population Slope - β_1
- Population Intercept - β_0
- Conditional Mean - $\mu_{Y|x}$

Inference for the Slope (β_1)

- Discuss inference for β_1 in detail (then summarize the rest)

$$b_1 = \frac{\sum_{i=1}^n (x_i - \bar{x})(Y_i - \bar{Y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})Y_i}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

- b_1 is a linear combination of normal random variables (the Y_i 's) so b_1 is normally distributed with

$$E(b_1) = \beta_1 \quad \text{Var}(b_1) = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

- $b_1 \sim N\left(\beta_1, \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}\right)$

Inference for the Slope (β_1)

Examine

$$\text{Var}(b_1) = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

A more precise estimate of β_1 can be obtained by :

- Spreading out the X values
- Getting a larger sample i.e. more (X, Y) pairs
- Making the error variance smaller

Inference for the Slope (β_1)

- Use MS_{error} to estimate σ^2
(Note that $MS_{error} \sim \frac{\sigma^2 \chi_{n-2}^2}{n-2}$)
- Standard error of b_1 is $S_{b_1} = \sqrt{MS_{error} / \sum_{i=1}^n (x_i - \bar{x})^2}$
- $(b_1 - \beta_1) / S_{b_1}$ has a t -distribution with $n - 2$ d.f.

Hypothesis Test for β_1

- Null and Alternative Hypotheses

$$H_0 : \beta_1 = 0 \quad H_a : \beta_1 \neq 0$$

- Test Statistic

$$T = \frac{b_1 - 0}{S_{b_1}}$$

- Reject H_0 if $|T| > t_{n-2, 1-\alpha/2}$
- Note that $T^2 = F$, this t -test for β_1 is the same as the F-test for significance of model from ANOVA Table.
- One-sided alternative hypothesis is possible for the t -test:
 $H_a : \beta_1 > 0$ or $H_a : \beta_1 < 0$

CI for β_1

- $100(1 - \alpha)\%$ confidence interval for β_1 :

$$b_1 \pm t_{n-2, 1-\alpha/2} S_{b_1}$$

Forbes Data

Weisberg, Sanford, *Applied Linear Regression*, Wiley, 1980.

- James D. Forbes collected data in the mountains of Scotland
- $n=17$ locations (at different altitudes)
- Objective: Predict barometric pressure (in inches of mercury) from boiling point of water (X) in $^{\circ}\text{F}$.
- Use $Y=\log(\text{barometric pressure})$
- Motivation: Fragile barometers of the 1840's were difficult to transport

Analysis of the Forbes Data

- Test $H_o : \beta_1 = 0$ ($Y_i = \beta_0 + \epsilon_i$)
versus $H_a : \beta_1 \neq 0$ ($Y_i = \beta_0 + \beta_1 x_i + \epsilon_i$)

- Evaluate

$$t = \frac{b_1 - 0}{s_{b_1}} = \frac{.020623 - 0}{0.000379} = 54.42$$

- The least squares estimate of the slope is 54 standard errors away from zero (p-value $\ll .0001$).
It is extremely unlikely that an estimate that far from zero could occur simply because of random errors when β_1 is actually zero.
Consequently, reject the null hypothesis and conclude that the slope is positive.

Analysis of the Forbes Data

- A 95% confidence interval for the slope indicates that the slope is “very well” estimated from these data

$$b_1 \pm t_{15,.975} S_{b_1}$$

$$\Rightarrow 0.020623 \pm (2.131)(0.00037895)$$

$$\Rightarrow (0.0198, 0.0214)$$

Inference for the Intercept (β_0)

- $b_o = \bar{Y} - b_1 \bar{x} \sim N(\beta_o, \sigma^2(\frac{1}{n} + \frac{\bar{x}^2}{\sum_i (x_i - \bar{x})^2}))$

- b_o has standard error $S_{b_o} = \sqrt{MS_{error} \left(\frac{1}{n} + \frac{\bar{x}^2}{\sum_{i=1}^n (x_i - \bar{x})^2} \right)}$

- Reject $H_o : \beta_o = 0$ if $|t| = \left| \frac{b_o - 0}{S_{b_o}} \right| > t_{n-2, 1-\alpha/2}$

- $100(1 - \alpha)\%$ confidence interval for β_o is

$$b_o \pm t_{n-2, 1-\alpha/2} S_{b_o}$$

Inference for the Intercept (β_0)

- Rarely considered
- Values of x must be near 0 for meaningful interpretations
- Would be most likely to use confidence interval

Analysis of the Forbes Data

- Test $H_o : \beta_0 = 0$ ($Y_i = \beta_1 x_i + \epsilon_i$)
versus $H_a : \beta_0 \neq 0$ ($Y_i = \beta_0 + \beta_1 x_i + \epsilon_i$)
- Evaluate $t = \frac{b_0 - 0}{S_{b_0}} = \frac{-0.971 - 0}{0.0769} = -12.6$
- The least squares estimate of the intercept is 12.6 standard errors away from zero (p-value $\ll .0001$).
Reject the null hypothesis and conclude that the intercept is negative. (No practical motivation)
- A 95% confidence interval for the intercept is
 $b_0 \pm t_{15,.975} S_{b_0} \Rightarrow -0.971 \pm (2.131)(0.0769) \Rightarrow (-1.135, -0.807)$

Inference for Conditional Means

Inference for $\mu_{Y|x} = E(Y|X = x) = \beta_o + \beta_1 x$

- Estimate is $\hat{\mu}_{Y|x} = b_o + b_1 x$
- $\hat{\mu}_{Y|x}$ is a linear function of two normally distributed random variables (b_o and b_1 , not independent)
- $\hat{\mu}_{Y|x}$ is $N\left(\beta_o + \beta_1 x, \sigma^2 \left(\frac{1}{n} + \frac{(x - \bar{x})^2}{\sum_{i=1}^n (x_i - \bar{x})^2}\right)\right)$
- Note: value of x does not need to be present in sample.

Inference for Conditional Means

- standard error is

$$S_{\hat{\mu}_{Y|x}} = S_{b_o + b_1 x} = \sqrt{MS_{error} \left(\frac{1}{n} + \frac{(x - \bar{x})^2}{\sum_{i=1}^n (x_i - \bar{x})^2} \right)}$$

- $100(1 - \alpha)\%$ confidence interval for $\beta_o + \beta_1 x$ is

$$(b_o + b_1 x) \pm t_{n-2, 1-\alpha/2} S_{\hat{\mu}_{Y|x}}$$

Confidence Region for a Line Segment

Use the Scheffe' procedure to get simultaneous confidence intervals for every x in an entire line segment:

$$(b_0 + b_1x) \pm \sqrt{2F_{2,n-2,1-\alpha}} S_{b_0+b_1x}$$

for $a \leq x \leq b$

Prediction

Predict the value for Y at given x :

$$Y_{new} = \beta_0 + \beta_1 x + \epsilon$$

- Estimate is still $\hat{Y} = b_0 + b_1 x$
- Standard error is

$$S_{pred} = \sqrt{MS_{error} \left(1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{\sum_{i=1}^n (x_i - \bar{x})^2} \right)}$$

- $100(1 - \alpha)\%$ prediction interval:

$$(b_0 + b_1 x) \pm t_{n-2, 1-\alpha/2} S_{pred}$$

Comparison

- Confidence Interval for Condition Mean $\mu_{Y|x}$
 - Inference for a point on the population regression line given value of x
 - Source of inference is estimating regression line
- Prediction Interval for Y
 - Inference for a point in the scatterplot of all population values given value of x .
 - Sources of inference are estimating regression line AND predicting Y given the regression line.

Analysis of the Forbes Data

- Construct a 95% confidence interval for the mean of possible log-pressure measurements when the boiling point of water is $x=209$ °F

- Estimated mean is

$$\hat{\mu}_{Y|x} = b_0 + b_1x = -0.9710 + (.0206)(209) = 3.339$$

- Evaluate the standard error of this estimate

$$S_{\hat{\mu}_{Y|x}} = \sqrt{.0000762 \left(\frac{1}{17} + \frac{(209 - 202.953)^2}{530.78} \right)} = 0.00312$$

- A 95% confidence interval is

$$\hat{\mu}_{Y|x} \pm t_{15,.975} S_{\hat{\mu}_{Y|x}} \Rightarrow 3.339 \pm (2.131)(0.00312) \Rightarrow (3.333, 3.346)$$

Analysis of the Forbes Data

- Apply the exponential function to the end points to get an *approximate* confidence interval for the mean pressure

(28.02, 28.39) inches of Hg

- This could be computed with either the REG procedure or the GLM procedure in SAS by adding an additional line to the data file with $X=209$ and a missing value for Y

Analysis of the Forbes Data

Scheffe procedure for constructing a 95% confidence region for a segment of the true regression line

$$\text{Evaluate } (b_0 + b_1x) \pm \sqrt{2F_{(2,n-2),1-\alpha}} S_{b_0+b_1x}$$

$$\Rightarrow (b_0 + b_1x) \pm \sqrt{2F_{(2,15),0.95}} S_{b_0+b_1x}$$

$$\Rightarrow (b_0 + b_1x) \pm (2.713) \sqrt{.0000762 \left(\frac{1}{17} + \frac{(x-202.953)^2}{530.78} \right)}$$

Analysis of the Forbes Data

- Construct a 95% prediction interval for a log-pressure value when the boiling point of water is $x=209$ °F
- Prediction is the estimated mean

$$\hat{Y} = b_0 + b_1x + error = -0.9710 + (.0206)(209) + 0 = 3.339$$

- Evaluate the standard error of the prediction (include the variation of the associated random error, estimated as $MS_{error} = .0000762$)

$$S_{pred} = \sqrt{.0000762 \left(1 + \frac{1}{17} + \frac{(209 - 202.953)^2}{530.78} \right)} = 0.00927$$

Analysis of the Forbes Data

- A 95% prediction interval is

$$\hat{y} \pm t_{15,.975} S_{pred} \quad \Rightarrow \quad 3.339 \pm (2.131)(0.00927)$$

$$\Rightarrow \quad (3.319, 3.359)$$

- Apply the exponential function to the end points to get an *approximate* prediction interval for barometric pressure:
(27.63, 28.76) inches of Hg
- This could be computed with either the REG or GLM procedure in SAS by adding an additional line to the data file with X=209 and a missing value for Y