

Regression Analysis

Multiple Linear Regression

Nicoleta Serban, Ph.D.

Professor

School of Industrial and Systems Engineering

Estimating Regression Line &
Predicting a New Response:
Data Example

About This Lesson



Linear Regression: Example 1

Quantitative Predicting Variables:

X_1 = amount (in hundreds of dollars) spent on advertising in 1999

X_2 = total amount of bonuses paid in 1999

X_3 = market share in each territory

X_4 = largest competitor's sales (thousands)

Qualitative Predicting Variable:

X_5 = indicates region where office is located
(1 = south, 2 = west, 3 = midwest)

Response Variable:

Y = yearly sales (in thousands of dollars)



Example 1: Mean Response & Prediction

- a. For all offices with the characteristics such as those of the first office:
 - What is the average estimated sales?
 - What is the standard deviation?
 - What is the 95% confidence interval for this mean response?

- b. If the first office's largest competitor's sales increase to \$303,000 (assuming everything else fixed):
 - What sales would you predict for the first office?
 - What is its standard deviation?
 - What is the 95% prediction interval for this prediction?

Example 1: Mean Response Estimation

```
s2 = summary(model)$sigma^2 # Variance estimate  
X = model.matrix(model) # Design Matrix  
xstar = X[1,] # First office data for formula  
resp.var = s2 *(xstar%*%solve(t(X)%*%X)%*%xstar) # Variance formula  
sqrt(resp.var)
```

```
[,1]  
[1,] 33.19118
```

```
newdata = meddcor[1,-1] # First office data for confidence interval  
predict(model, newdata, interval="confidence") # Confidence Interval
```

	fit	lwr	upr
1	934.7767	865.0446	1004.509

Example 1: Mean Response Estimation

```
s2 = summary(model)$sigma^2 # Variance estimate  
X = model.matrix(model) # Design Matrix  
xstar = X[1,] # First office data for formula  
resp.var = s2 * (xstar %*% solve(t(X) %*% X) %*% xstar) # Variance formula  
sqrt(resp.var)
```

```
[1]  
[1,] 33.19118
```

```
newdata = meddcor[1,-1] # First office data for confidence interval  
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fit	lwr	upr
1 934.7767	865.0446	1004.509

a. Average estimated sales (mean response for sales):

$$\hat{y} = 934.777$$

Estimated standard deviation:

$$se(\hat{y}) = 33.191$$

95% Confidence Interval:

$$(865.045, 1004.509)$$

Interpretation: For offices with the same characteristics as the first, the average estimated sales are \$934,777, with a lower bound of \$865,045 and an upper bound of \$1,004,509.

Example 1: Mean Response Estimation

```
## Change the competitor's sales for prediction of future observation
```

```
xstar.new = xstar  
xstar.new[5] = 303
```

```
# Variance formula
```

```
pred.var = s2*(1+xstar.new%*%solve(t(X)%*%X)%*%xstar.new)  
sqrt(pred.var)
```

```
[,1]  
[1,] 64.31099
```

```
# Prediction Interval
```

```
predict(model, xstar.new[-1], interval="prediction")
```

	fit	lwr	upr
1	911.0569	775.9446	1046.169

Example 1: Mean Response Estimation

```
## Change the competitor's sales for prediction of future observation
```

```
xstar.new = xstar
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xstar.new[5] = 303
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# Variance formula
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pred.var = s2*(1+xstar.new%*%solve(t(X)%*%X)%*%xstar.new)  
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```

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[,1]  
[1,] 64.31099
```

```
# Prediction Interval
```

```
predict(model, xstar.new[-1], interval="prediction")
```

fit	lwr	upr
1 911.0569	775.9446	1046.169

b. Predicted sales of the first office given the higher competitor's sales:

$$\hat{y} = 911.057$$

Estimated standard deviation:

$$se(\hat{y}) = 64.311$$

95% Confidence Interval:

$$(775.945, 1046.169)$$

Interpretation: If the competitor's sales increase to \$303,000 (from \$202,220), the predicted sales reduce by \$23,720 (from \$934,777 to \$911,057). Since this is prediction, the standard deviation increases.

Summary

