

1/29/14 Class Notes

In this lecture, we focused on prediction uncertainty associated with a linear model. Data and R scripts used were milk prices and pickup truck sales on Craigslist. We worked individually in class, and then reviewed the examples during the lecture.

Optimize Profit

a. EXAMPLE: (milk.csv)

- i. How much should we charge for gallons of milk?
- ii. Start from the end; we want to maximize profit

Let

Y = profit

X = price per unit \rightarrow choice variable

C = cost per unit (\$1.50) \rightarrow known

$F(x)$ = units sold \rightarrow unknown: it is a function of x

Profit = revenue – cost

$Y = f(x) * (\text{price/unit} - \text{cost/unit})$

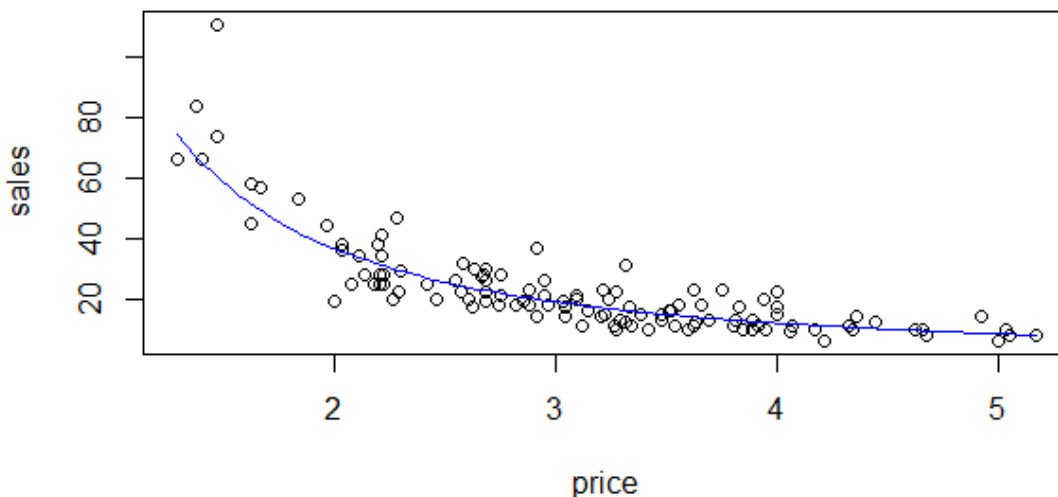
$Y = f(x) * (x - c)$

Profit = $x * f(x) - c * f(x)$

iii. Quantity sold VS. Price charged per unit

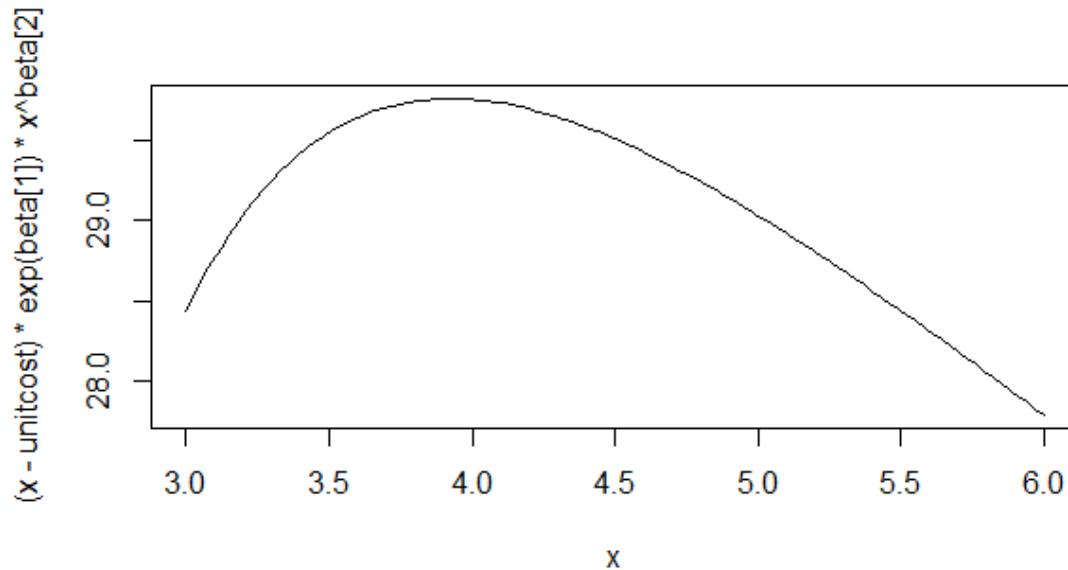
1. Data suggests that as price increases, sales decrease
2. DEMAND CURVE

Demand Curve of Milk Prices



b. Optimization strategies

- i. Take derivative and set to 0
- ii. Plot and Point Method → Plot points and take maximum



Reducing Uncertainty

a. EXAMPLE: 4th Story (pickup.csv, pickup.r)

- i. Remember that,

$$Y = \underbrace{B_0 + B_1(X_i)}_{\text{Systematic Model Fitted}} + \underbrace{E_i}_{\text{Residual}}$$

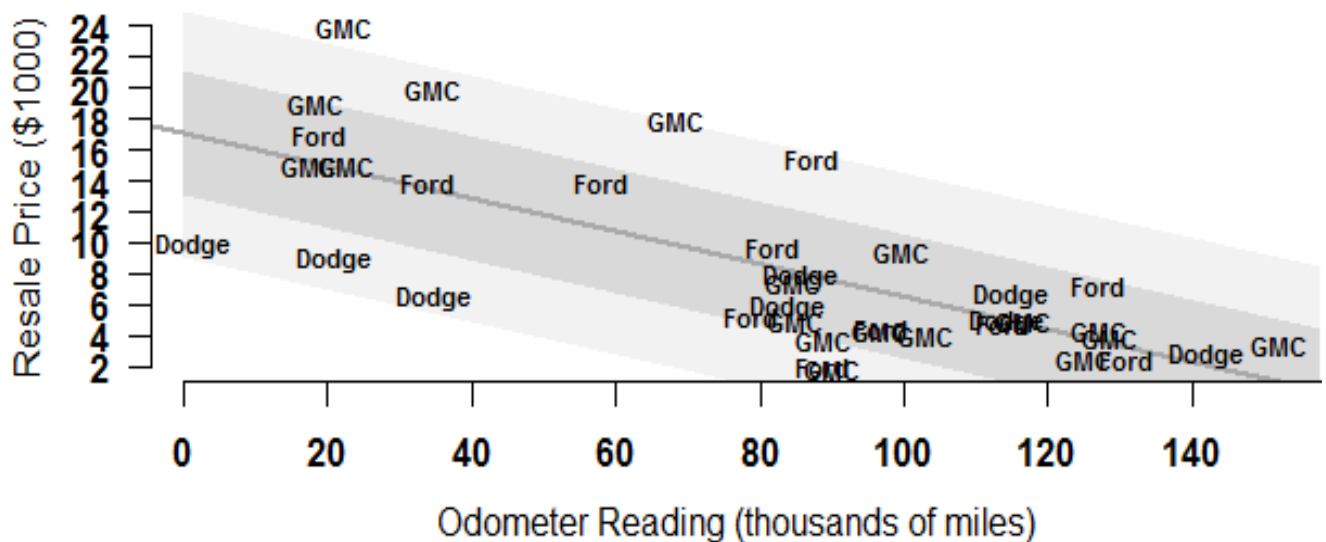
- ii. Point Estimate = $y^* = B_0 + Bx^* + E^*$

- a. Plug in 60,000 for X.
- b. The purpose of doing this is to take the past variability as measure for likely size of e^* . You can use the standard deviation of the residuals.

$$y^* = \hat{\beta}_0 + \hat{\beta}_1 x^* \pm s_e$$

- iii. Interval Estimate =

- a. We widen the interval in order to reduce the likelihood of error
- b. There are no predictions without error bars



How many data points fell outside the given window?

This gives us a measure of empirical coverage.

R² (Coefficient of Determination):

- a. A measure of the information content on a predictor
- b. How to construct:
 - i. Se = Standard deviation of residuals
 - ii. Sy = std dev of original y variable
 - iii. Ratio of these quantities (Se/Sy) tells about predictor:
 - The higher that ratio, the lower the information
 - The lower the ration, the higher the information
 - $1 - (Se/Sy)^2$ (puts on a more intuitive scale)
- c. Relates to Pythagoreans Theorem:
 - Some squares in statistics sum up
 - $Var(resid) + var(fitted\ Values) = var(data)$

Will not dwell on this in class but spend some time understanding this in the notes.

Extra remarks:

- a. `Pdata()` returns fraction of data that falls below given values
- b. Construct intervals on transformed scales and undo predictions at the end
- c. USING R AS A GRAPHING CALCULATOR
 - a. `curve((x-unitcost)*exp(beta[1])*x^beta[2], from=1, to=10)`
 - b. “from = 1, to = 10” set the boundaries of the equation