
Hw3 Part 1 Review

CS100

Linear Regression

Assumed Linear Model: $Y = A + Bx$

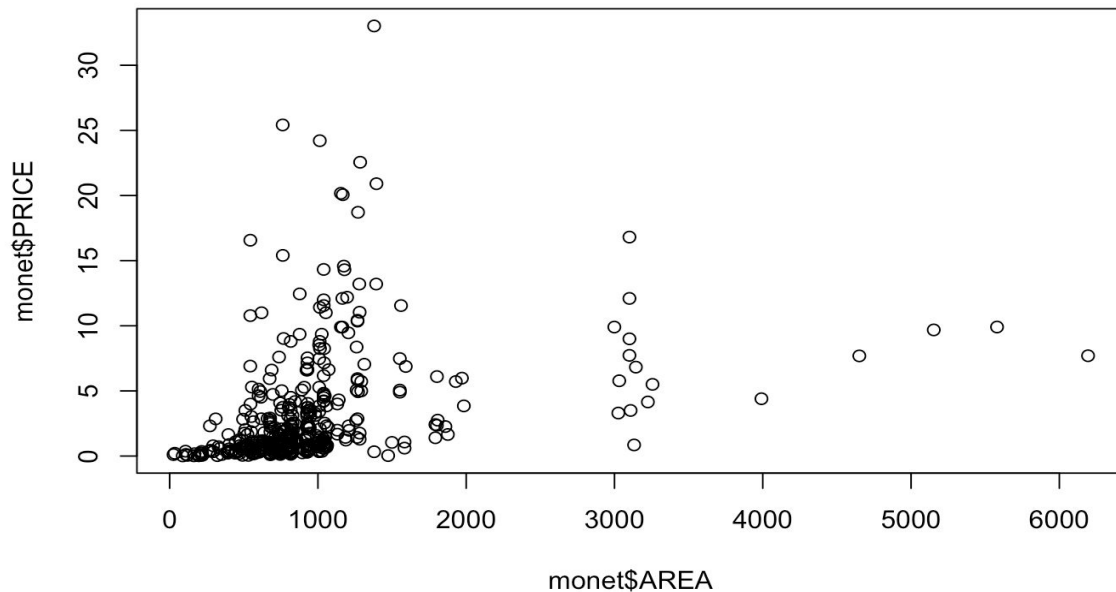
`abline(a = A, b = B)`

Log/Sqrt Transformation

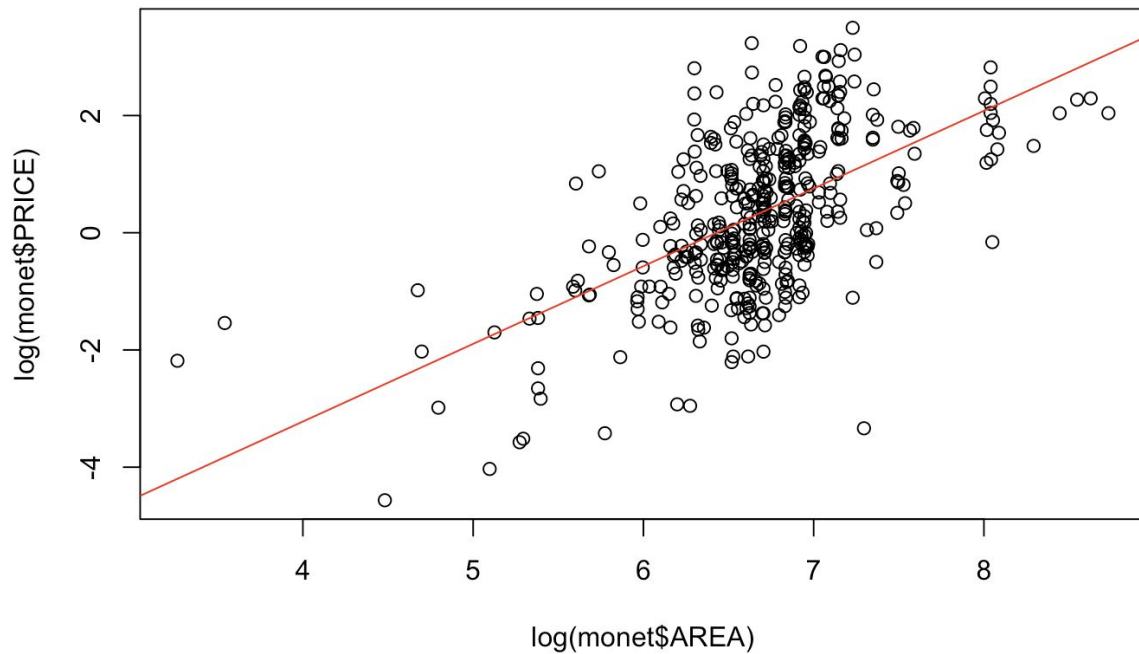
When/Why do we do Log or Sqrt Transformation?

- Not all relationships are Linear (EX) monet.PRICE VS monet.AREA
- Some of the relationships can be **MADE** linear with **APPROPRIATE** transformation.
- If residual plot is random, it is good fit for Linear Model.
- If residual plot is not random (U curve, inverted-U curve), not a good fit for Linear Model → We transform the data to use fit linear model.

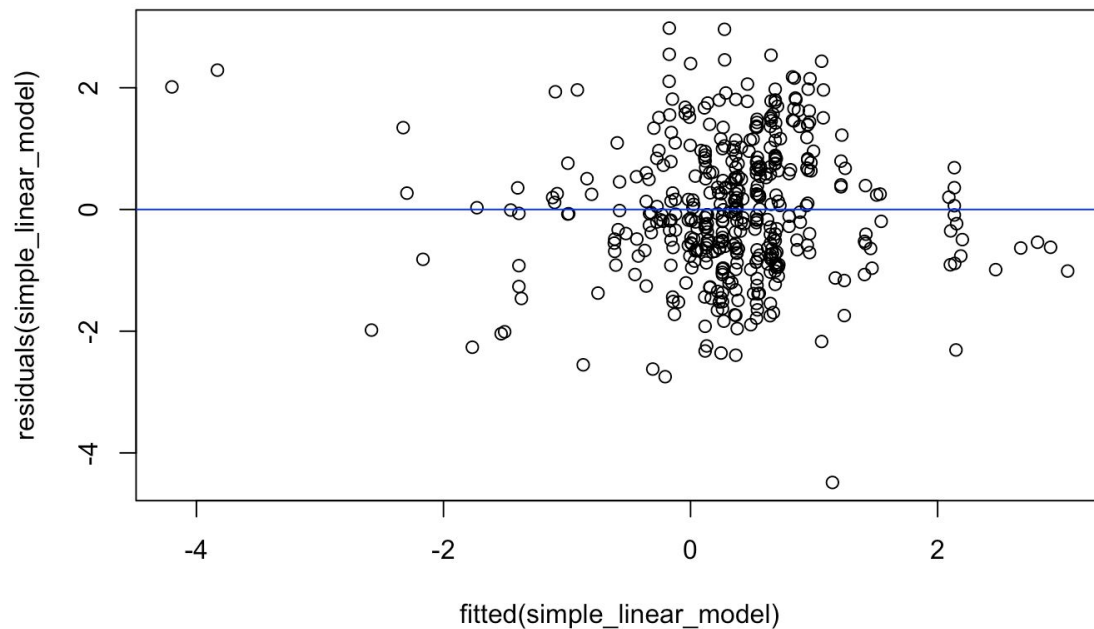
PRICE vs AREA (LINEAR RELATIONSHIP?)



Log(PRICE) vs Log(AREA) (BETTER?)



RESIDUAL (remove the first 2...??)



CORRELATION COEFFICIENT

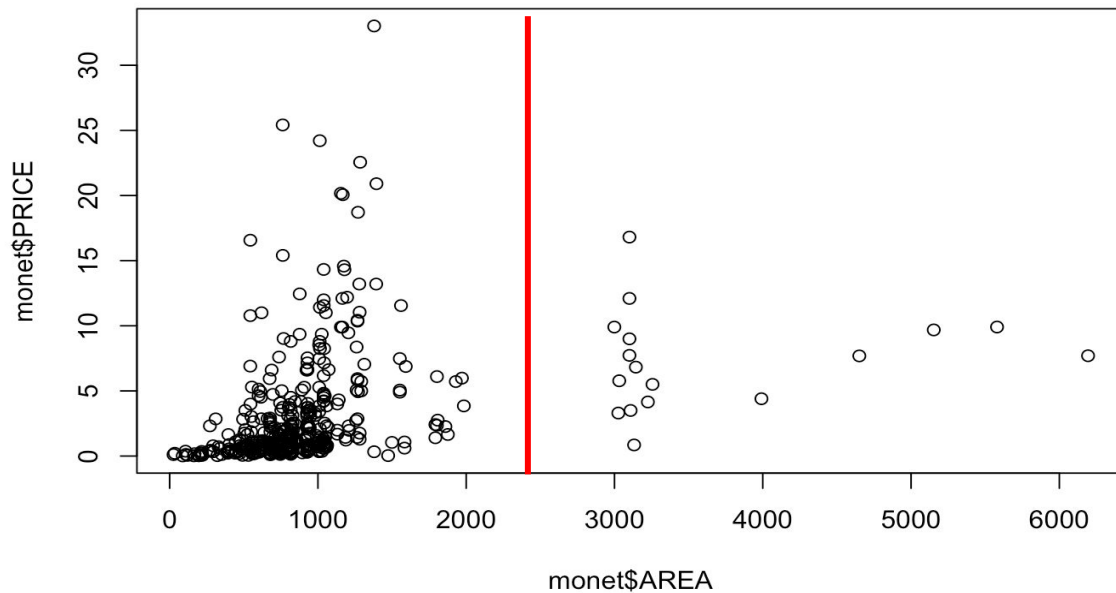
- measures the **STRENGTH** of linear relationship between 2 variables.

REGRESSION SLOPE

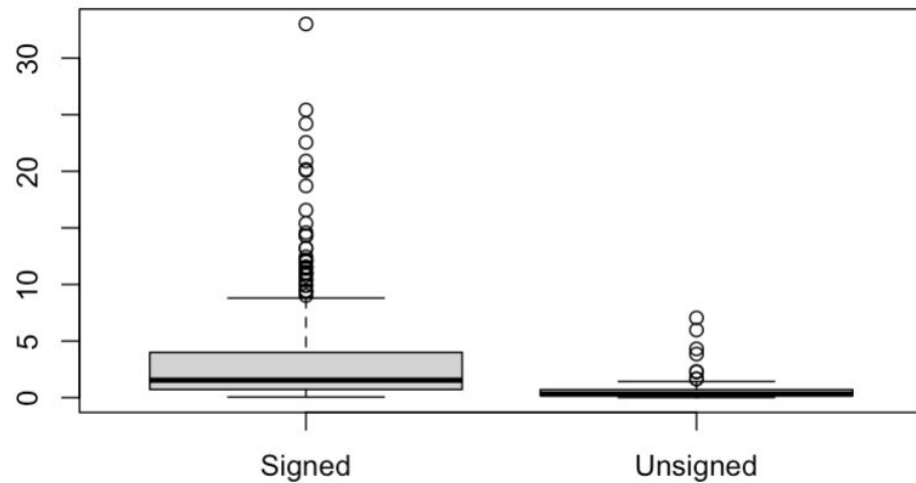
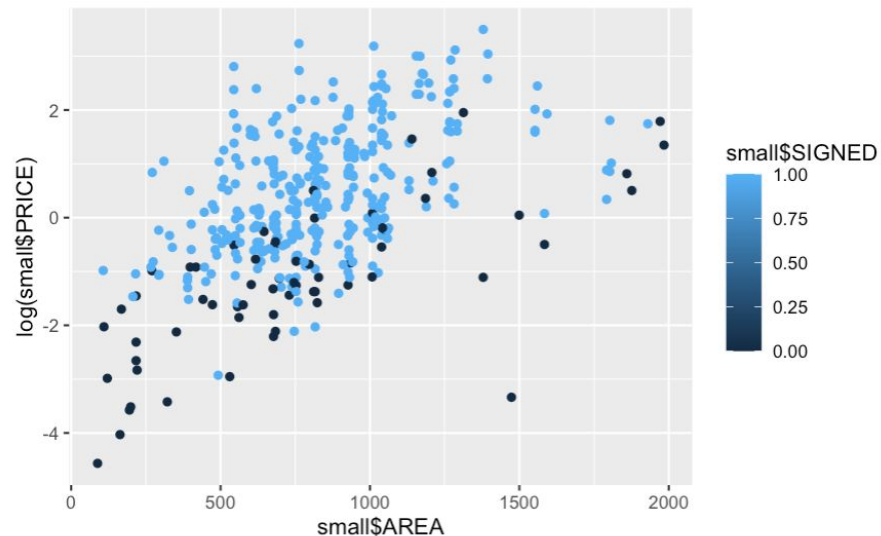
- measures the **STEEPNESS** of the linear relationship between 2 variables.

“Intuitively, the more correlated 2 variables are, the easier it is to draw a line of best linear fit. Hence, we computed the correlation of each transformation.”

PRICE vs AREA

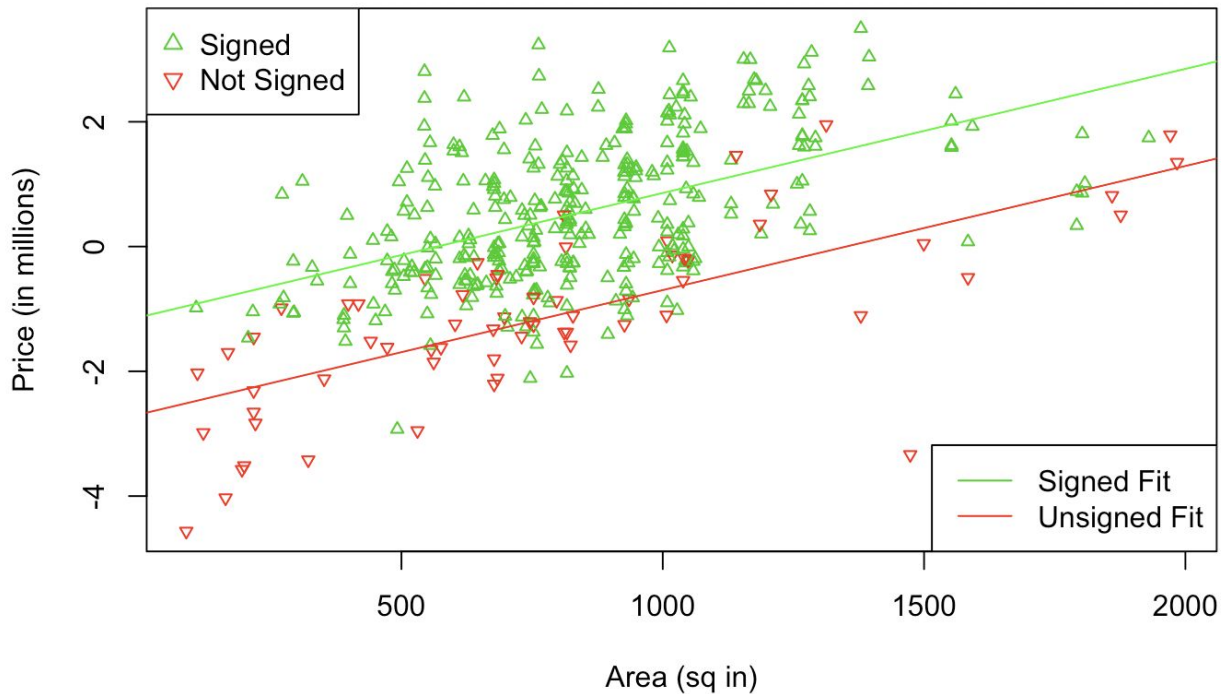


SIGNED AND UNSIGNED PLOT OF SMALL PICTURES



log(PRICE) vs AREA fit curve

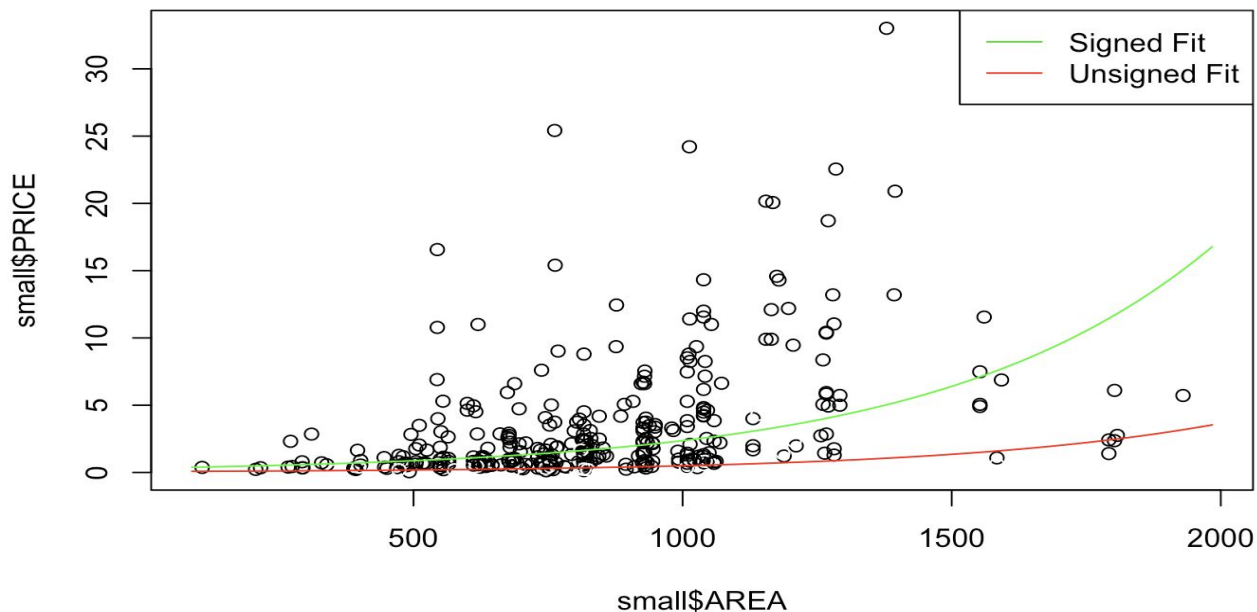
Monet's Small Paintings



```
```{r}
signed_linear_model <- lm(log(small$PRICE) ~ small$AREA + small$SIGNED)
a <- signed_linear_model$coefficients[1]
b <- signed_linear_model$coefficients[2]
dummy <- signed_linear_model$coefficients[3]
plot(log(small$PRICE) ~ small$AREA, col = as.factor(small$SIGNED))
abline(a = a + dummy, b = b, col = "red")
abline(a = a, b = b, col = "black")
```
```

PRICE vs AREA fit curve

```
```\{r\}  
#Plot model in non - log space.
a <- signed_linear_model$coefficients[1]
b <- signed_linear_model$coefficients[2]
dummy <- signed_linear_model$coefficients[3]
plot(small$PRICE~small$AREA)
curve(exp(a + b * x + dummy), col = "green", add = TRUE)
curve(exp(a + b * x), col = "red", add = TRUE)
```\
```



RESIDUAL

