

# Linear Regression Part 2

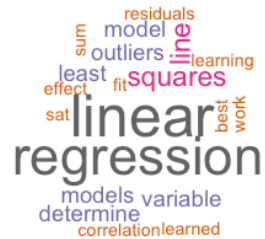
DATA 606 - Statistics & Probability for Data Analytics

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# One Minute Paper Results

**What was the most important thing you learned during this class?**



**What important question remains unanswered for you?**



# NYS Report Card

NYS publishes data for each school in the state. We will look at the grade 8 math scores for 2012 and 2013. 2013 was the first year the tests were aligned with the Common Core Standards. There was a lot of press about how the passing rates for most schools dropped. Two questions we wish to answer:

1. Did the passing rates drop in a predictable manner?
2. Were the drops different for charter and public schools?

```
load('../course_data/NYSReportCard-Grade7Math.Rda')
names(reportCard)
```

```
## [1] "BEDSCODE"      "School"        "NumTested2012" "Mean2012"
## [5] "Pass2012"      "Charter"       "GradeSubject"  "County"
## [9] "BOCES"         "NumTested2013" "Mean2013"      "Pass2013"
```

# reportCard Data Frame

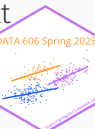
Show  entries

Search:

BEDSCODE	School	NumTested2012	Mean2012	Pass2012	Charter	GradeSubject	County	BOCES	NumTested2013	Mean2013	Pa
010100010020	NORTH ALBANY ACADEMY	47	649	13	false	Grade 7 Math	Albany	BOCES ALBANY-SCHOH-SCHENECTADY-SARAT	45	268	
010100010030	WILLIAM S HACKETT MIDDLE SCHOOL	212	652	30	false	Grade 7 Math	Albany	BOCES ALBANY-SCHOH-SCHENECTADY-SARAT	250	279	
010100010045	STEPHEN AND HARRIET MYERS MIDDLE SCHOOL	262	670	50	false	Grade 7 Math	Albany	BOCES ALBANY-SCHOH-SCHENECTADY-SARAT	256	284	

Showing 1 to 3 of 1,362 entries

Previous  2 3 4 5 ... 454 Next



# Descriptive Statistics

```
summary(reportCard$Pass2012)
```

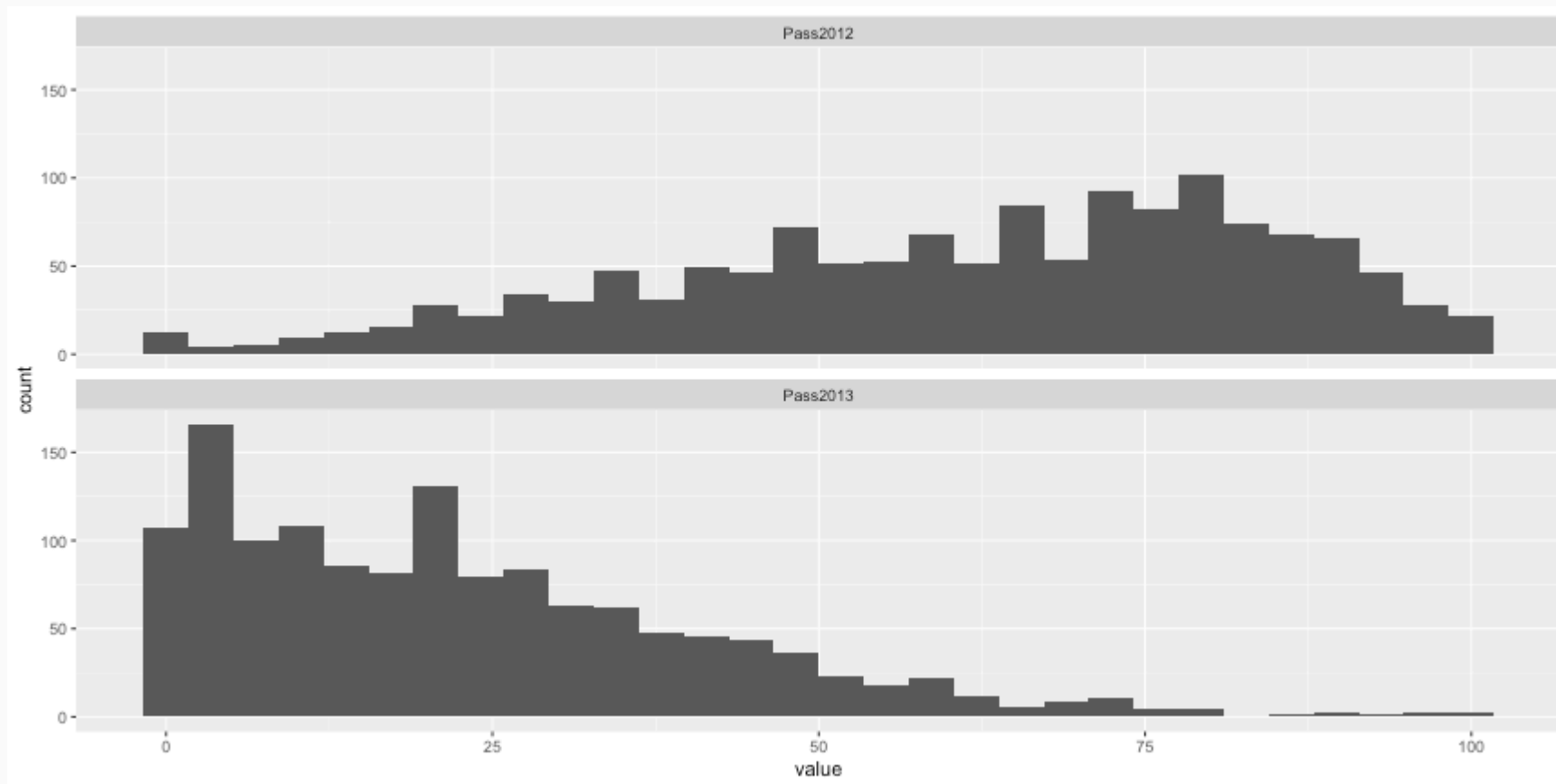
```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00   46.00   65.00   61.73   80.00   100.00
```

```
summary(reportCard$Pass2013)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00    7.00   20.00   22.83   33.00   99.00
```

# Histograms

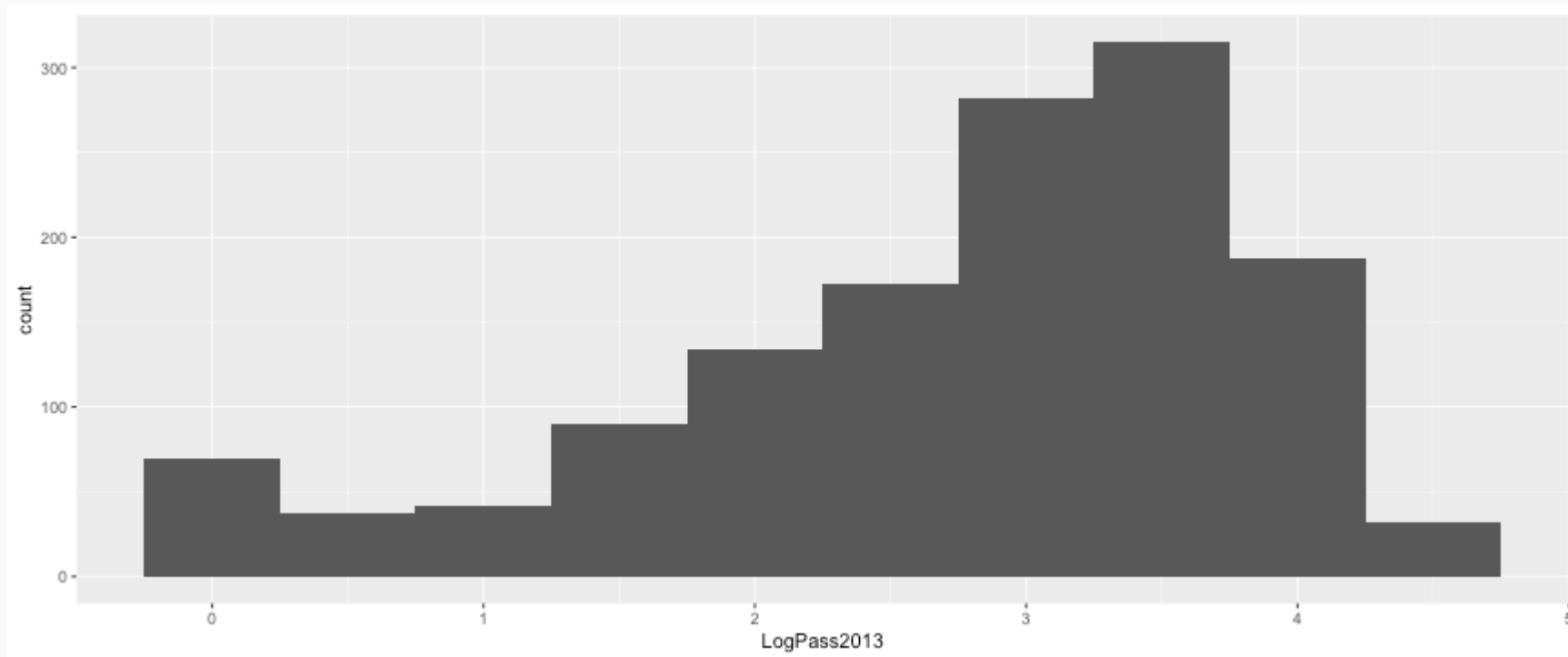
```
melted <- melt(reportCard[,c('Pass2012', 'Pass2013')])  
ggplot(melted, aes(x=value)) + geom_histogram() + facet_wrap(~ variable, ncol=1)
```



# Log Transformation

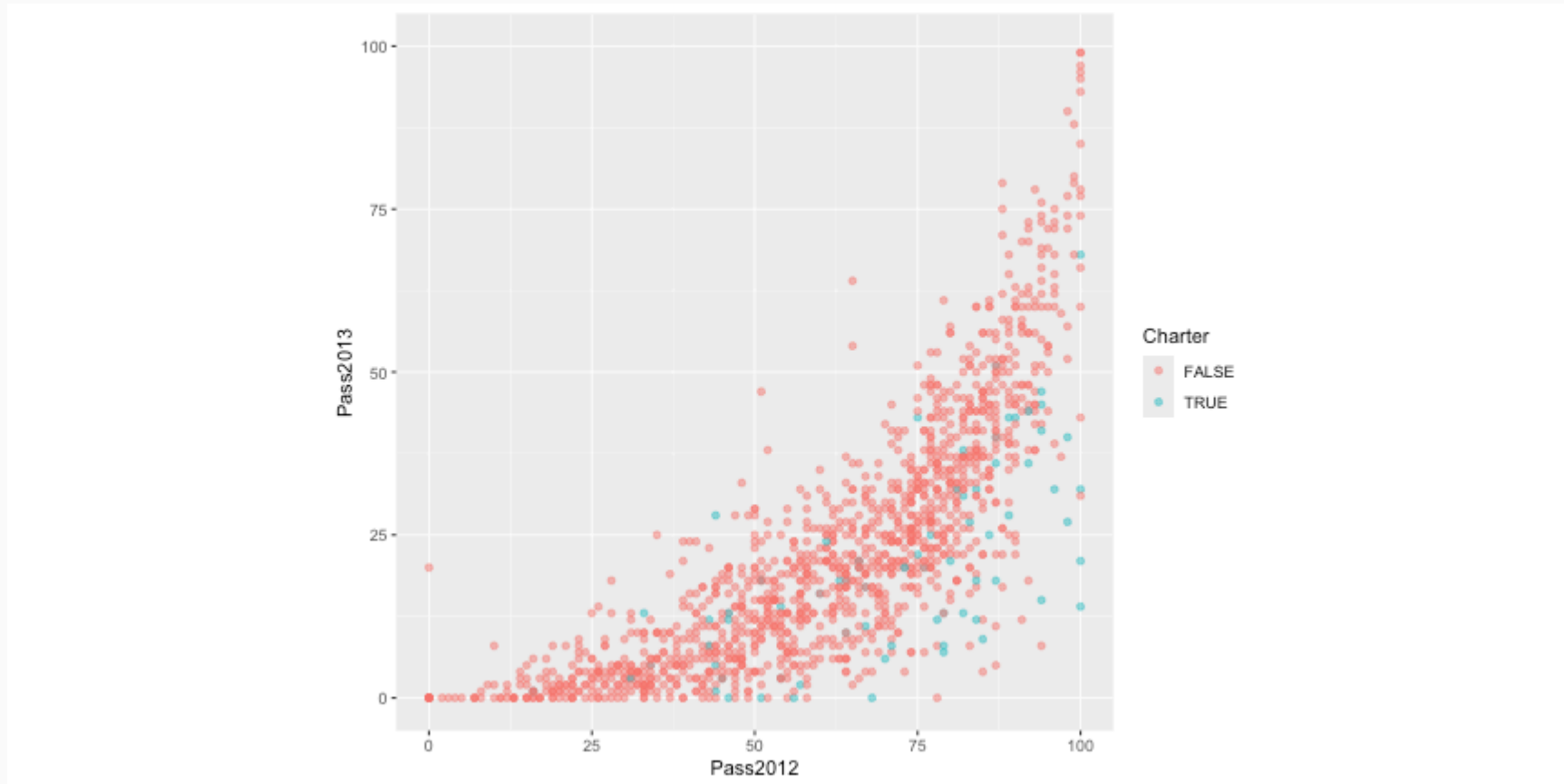
Since the distribution of the 2013 passing rates is skewed, we can log transform that variable to get a more reasonably normal distribution.

```
reportCard$LogPass2013 <- log(reportCard$Pass2013 + 1)  
ggplot(reportCard, aes(x=LogPass2013)) + geom_histogram(binwidth=0.5)
```



# Scatter Plot

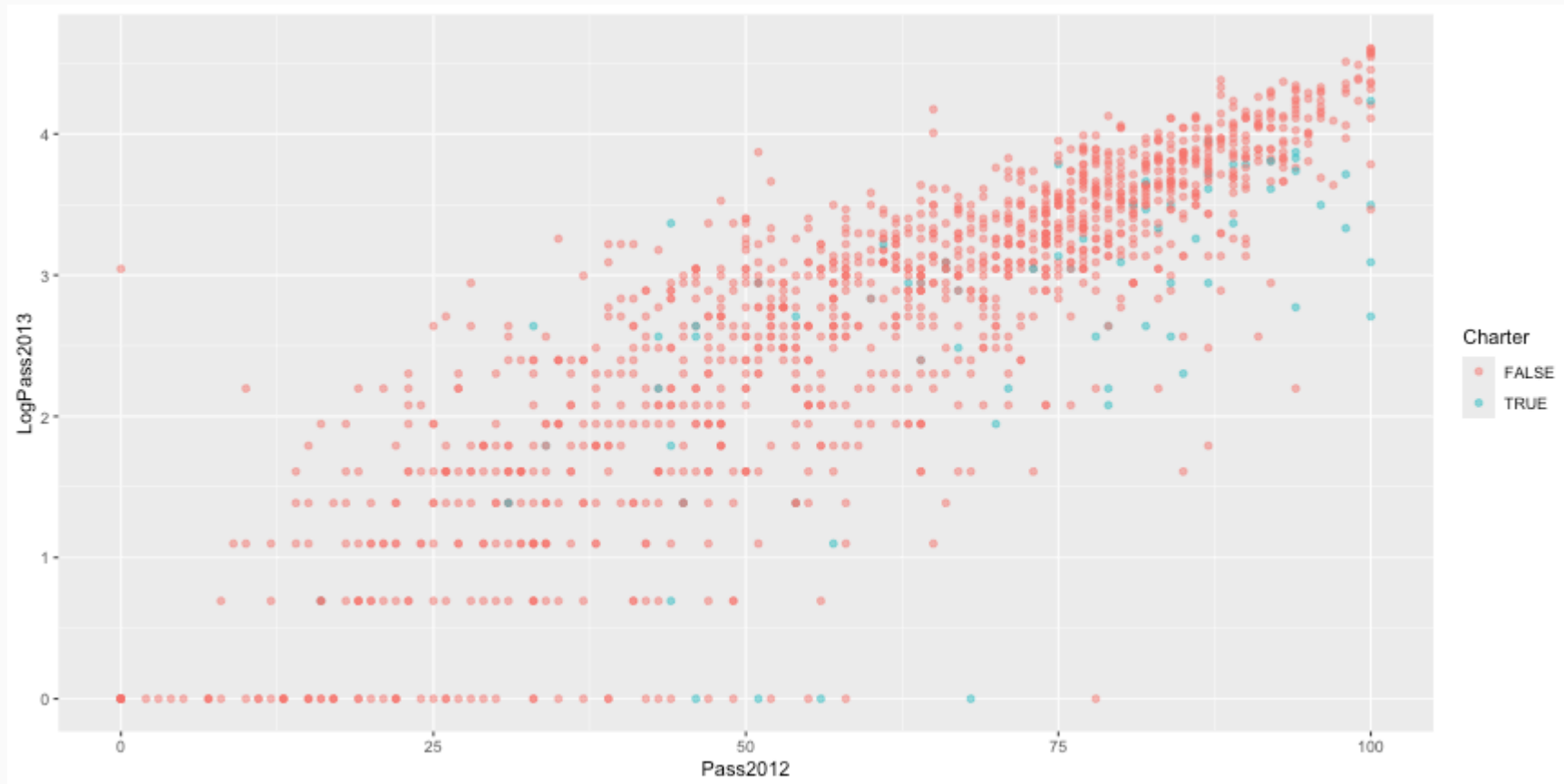
```
ggplot(reportCard, aes(x=Pass2012, y=Pass2013, color=Charter)) +  
  geom_point(alpha=0.5) + coord_equal() + ylim(c(0,100)) + xlim(c(0,100))
```





# Scatter Plot (log transform)

```
ggplot(reportCard, aes(x=Pass2012, y=LogPass2013, color=Charter)) +  
  geom_point(alpha=0.5) + xlim(c(0,100)) + ylim(c(0, log(101)))
```



# Correlation

```
cor.test(reportCard$Pass2012, reportCard$Pass2013)
```

```
##  
##      Pearson's product-moment correlation  
##  
## data:  reportCard$Pass2012 and reportCard$Pass2013  
## t = 47.166, df = 1360, p-value < 2.2e-16  
## alternative hypothesis: true correlation is not equal to 0  
## 95 percent confidence interval:  
##  0.7667526 0.8071276  
## sample estimates:  
##          cor  
## 0.7877848
```

# Correlation (log transform)

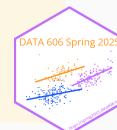
```
cor.test(reportCard$Pass2012, reportCard$LogPass2013)
```

```
##  
##      Pearson's product-moment correlation  
##  
## data:  reportCard$Pass2012 and reportCard$LogPass2013  
## t = 56.499, df = 1360, p-value < 2.2e-16  
## alternative hypothesis: true correlation is not equal to 0  
## 95 percent confidence interval:  
##  0.8207912 0.8525925  
## sample estimates:  
##          cor  
## 0.8373991
```

# Linear Regression

```
lm.out <- lm(Pass2013 ~ Pass2012, data=reportCard)
summary(lm.out)
```

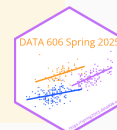
```
##
## Call:
## lm(formula = Pass2013 ~ Pass2012, data = reportCard)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -35.484  -6.878  -0.478   5.965  51.675
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -16.68965    0.89378  -18.67  <2e-16 ***
## Pass2012      0.64014    0.01357   47.17  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 11.49 on 1360 degrees of freedom
## Multiple R-squared:  0.6206,    Adjusted R-squared:  0.6203
## F-statistic: 2225 on 1 and 1360 DF,  p-value: < 2.2e-16
```



# Linear Regression (log transform)

```
lm.log.out <- lm(LogPass2013 ~ Pass2012, data=reportCard)
summary(lm.log.out)
```

```
##
## Call:
## lm(formula = LogPass2013 ~ Pass2012, data = reportCard)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.3880 -0.2531  0.0776  0.3461  2.7368
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.307692   0.046030   6.685 3.37e-11 ***
## Pass2012     0.039491   0.000699  56.499 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5915 on 1360 degrees of freedom
## Multiple R-squared:  0.7012,    Adjusted R-squared:  0.701
## F-statistic: 3192 on 1 and 1360 DF,  p-value: < 2.2e-16
```



# Did the passing rates drop in a predictable manner?

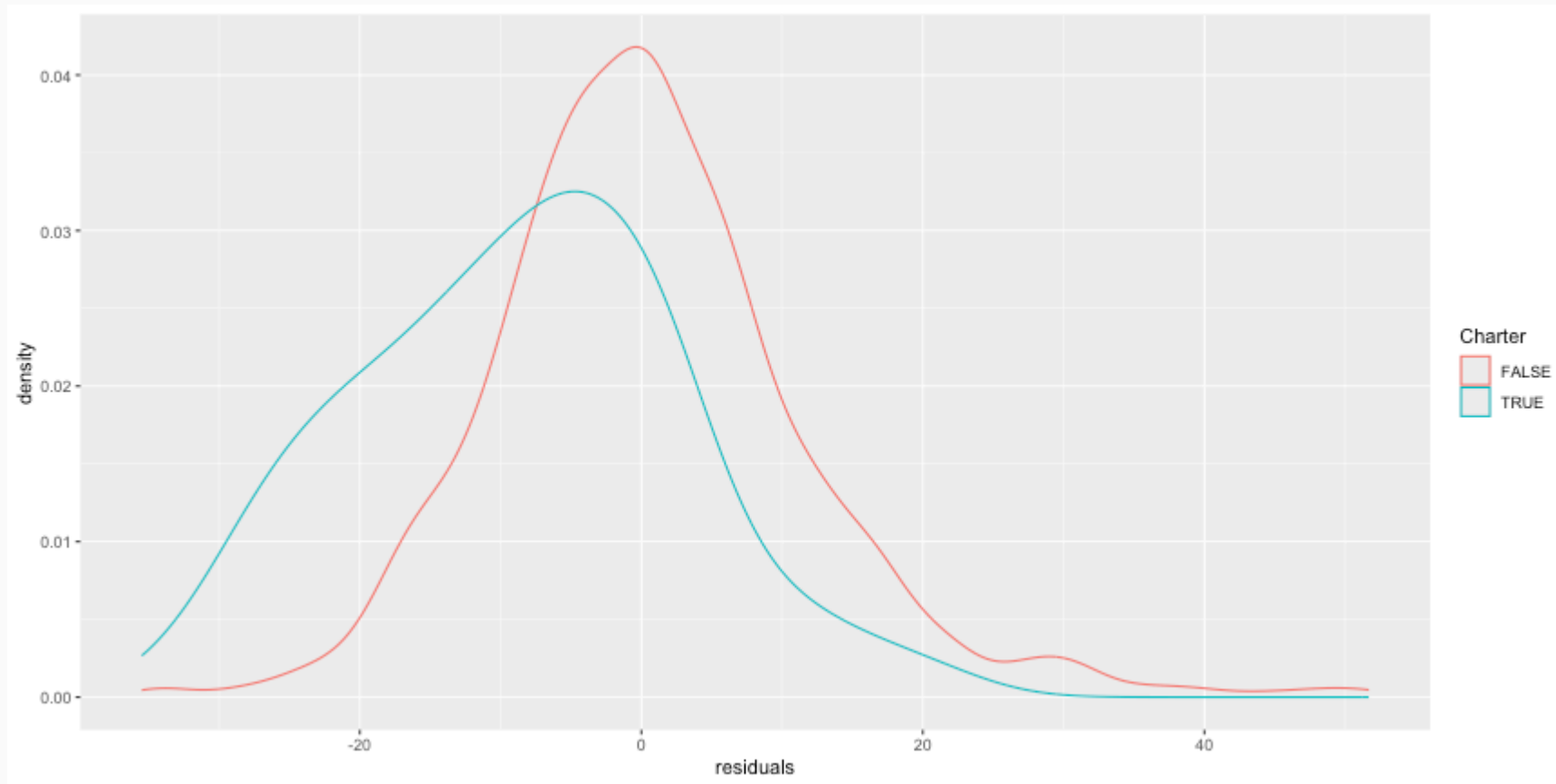
Yes! Whether we log transform the data or not, the correlations are statistically significant with regression models with  $R^2$  greater than 62%.

To answer the second question, whether the drops were different for public and charter schools, we'll look at the residuals.

```
reportCard$residuals <- resid(lm.out)
reportCard$residualsLog <- resid(lm.log.out)
```

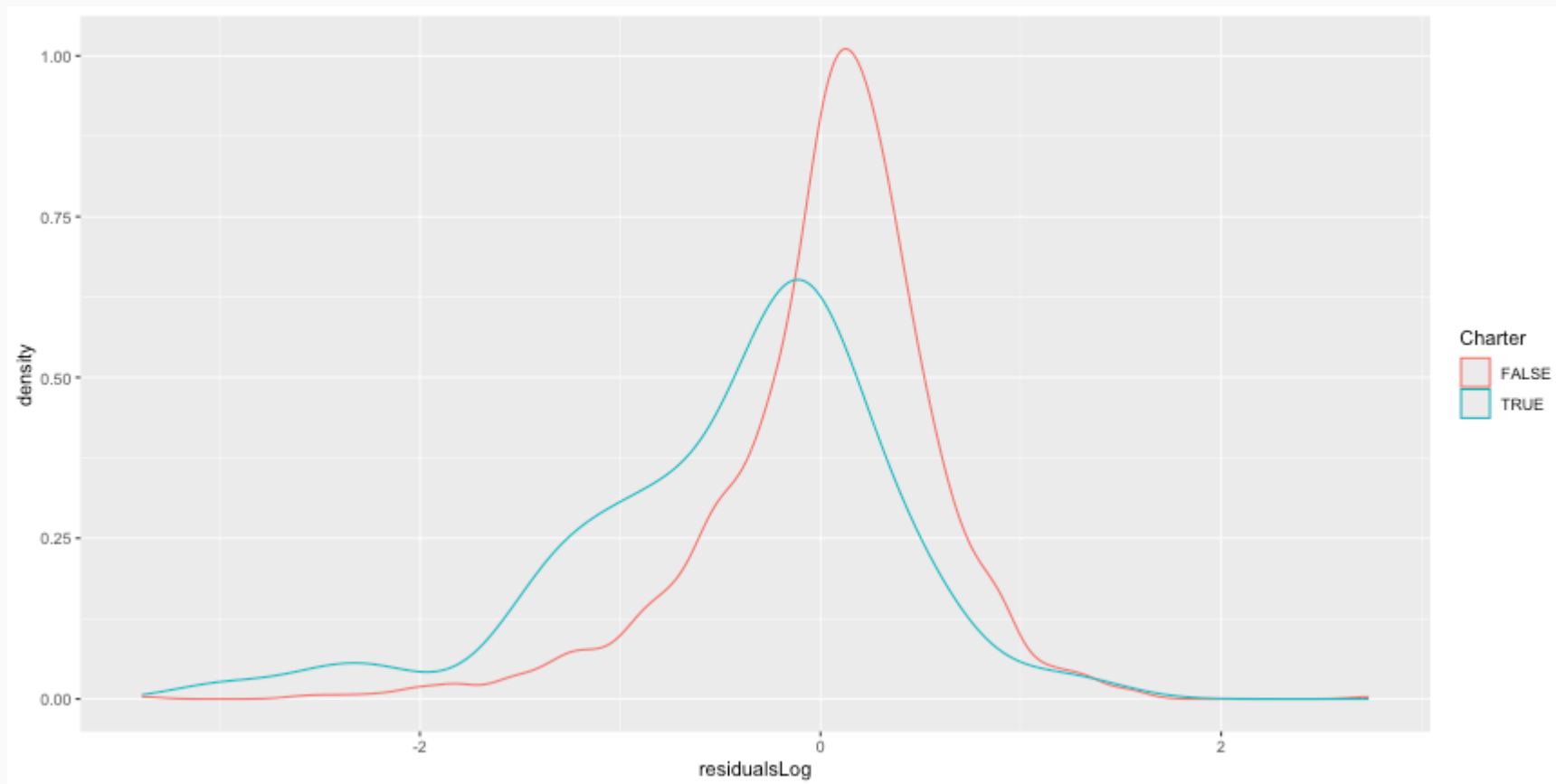
# Distribution of Residuals

```
ggplot(reportCard, aes(x=residuals, color=Charter)) + geom_density()
```



# Distribution of Residuals

```
ggplot(reportCard, aes(x=residualsLog, color=Charter)) + geom_density()
```





# Null Hypothesis Testing

$H_0$ : There is no difference in the residuals between charter and public schools.

$H_A$ : There is a difference in the residuals between charter and public schools.

```
t.test(residuals ~ Charter, data=reportCard)
```

```
##  
##      Welch Two Sample t-test  
##  
## data:  residuals by Charter  
## t = 6.5751, df = 77.633, p-value = 5.091e-09  
## alternative hypothesis: true difference in means between group FALSE and group TRUE is not equal to 0  
## 95 percent confidence interval:  
##  6.411064 11.980002  
## sample estimates:  
## mean in group FALSE  mean in group TRUE  
##           0.479356           -8.716177
```

# Null Hypothesis Testing (log transform)

```
t.test(residualsLog ~ Charter, data=reportCard)
```

```
##  
##      Welch Two Sample t-test  
##  
## data:  residualsLog by Charter  
## t = 4.7957, df = 74.136, p-value = 8.161e-06  
## alternative hypothesis: true difference in means between group FALSE and group TRUE is not equal to 0  
## 95 percent confidence interval:  
##  0.2642811 0.6399761  
## sample estimates:  
## mean in group FALSE  mean in group TRUE  
##      0.02356911      -0.42855946
```

# Polynomial Models (e.g. Quadratic)

It is possible to fit quadratic models fairly easily in R, say of the following form:

$$y = b_1x^2 + b_2x + b_0$$

```
quad.out <- lm(Pass2013 ~ I(Pass2012^2) + Pass2012, data=reportCard)
summary(quad.out)$r.squared
```

```
## [1] 0.7065206
```

```
summary(lm.out)$r.squared
```

```
## [1] 0.6206049
```

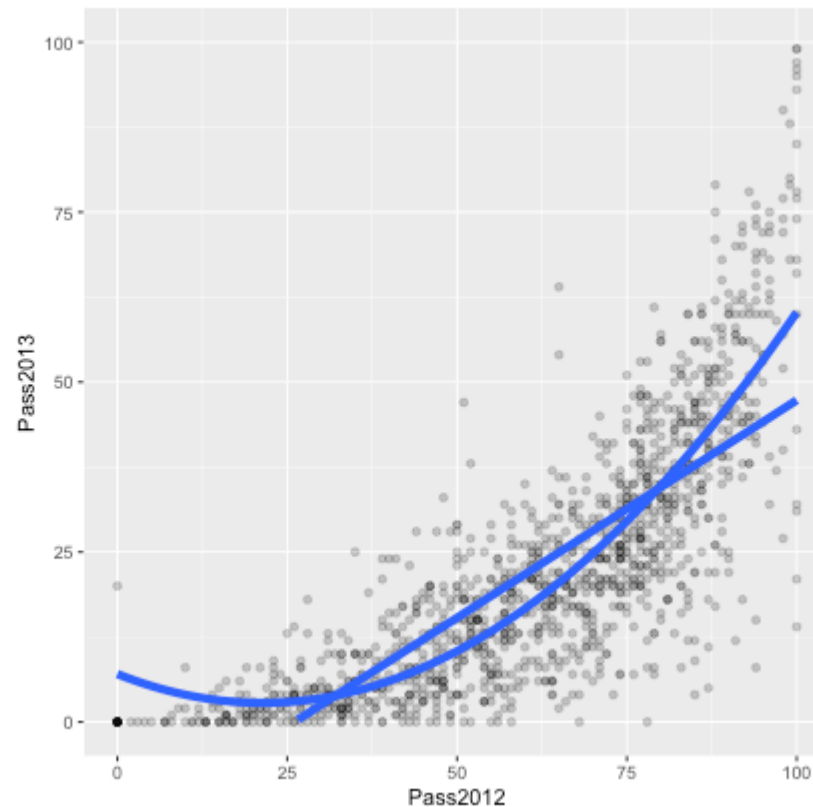
# Quadratic Model

```
summary(quad.out)
```

```
##
## Call:
## lm(formula = Pass2013 ~ I(Pass2012^2) + Pass2012, data = reportCard)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -46.258  -4.906  -0.507   5.430  43.509
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   7.0466153   1.4263773    4.940 8.77e-07 ***
## I(Pass2012^2)  0.0092937   0.0004659   19.946 < 2e-16 ***
## Pass2012     -0.3972481   0.0533631   -7.444 1.72e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.11 on 1359 degrees of freedom
## Multiple R-squared:  0.7065,    Adjusted R-squared:  0.7061
## F-statistic: 1636 on 2 and 1359 DF,  p-value: < 2.2e-16
```

# Scatter Plot

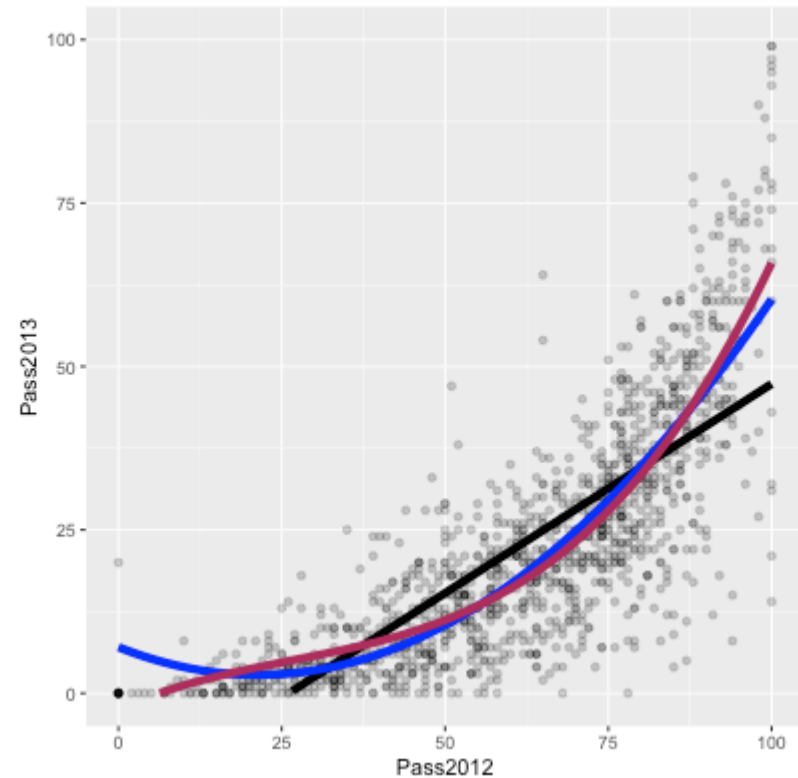
```
ggplot(reportCard, aes(x=Pass2012, y=Pass2013)) + geom_point(alpha=0.2) +  
  geom_smooth(method='lm', formula=y ~ x, size=2, se=FALSE) +  
  geom_smooth(method='lm', formula=y ~ I(x^2) + x, size=2, se=FALSE) +  
  coord_equal() + ylim(c(0,100)) + xlim(c(0,100))
```



# Let's go crazy, cubic!

```
cube.out <- lm(Pass2013 ~ I(Pass2012^3) + I(Pass2012^2) + Pass2012, data=reportCard)  
summary(cube.out)$r.squared
```

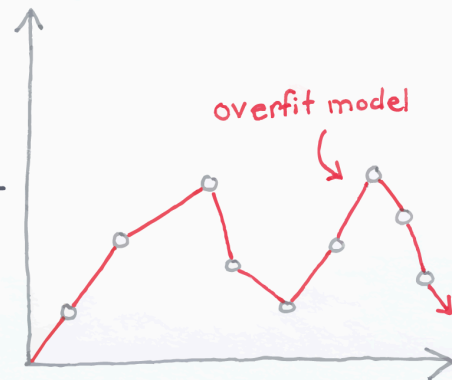
```
## [1] 0.7168206
```



# Be careful of overfitting...

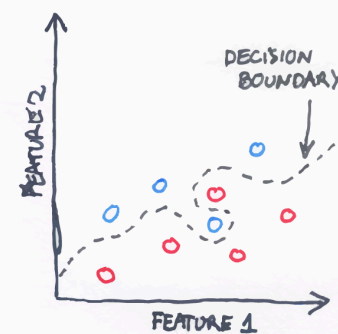
## OVERFITTING

Overfitting occurs when a model starts to memorize the aspects of the training set and in turn loses the ability to generalize

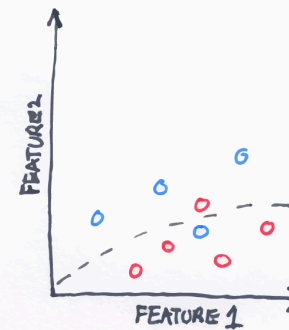


Chris Albon

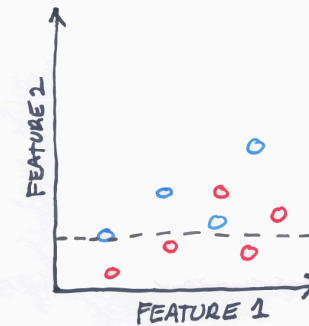
## OVERFIT VS UNDERFIT



OVERFIT  
"HIGH VARIANCE"



IDEAL

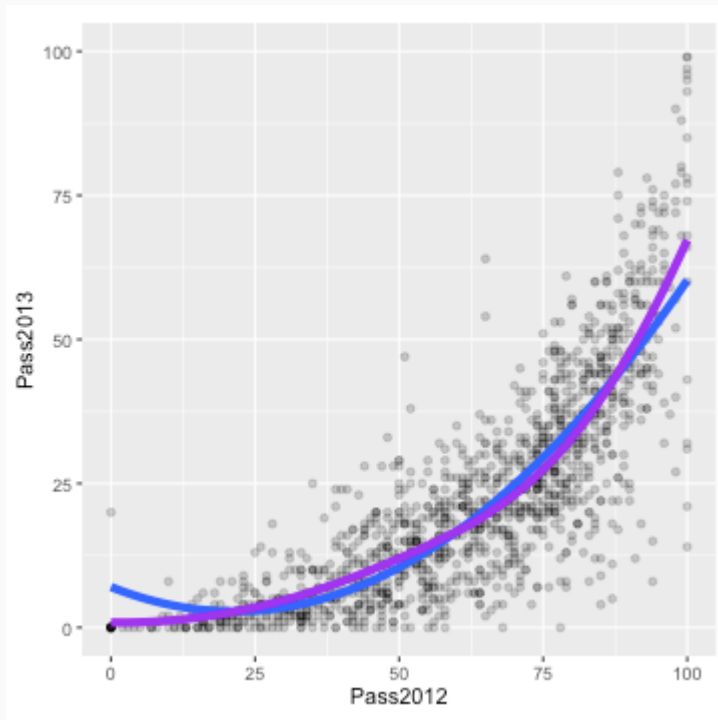


UNDERFIT  
"HIGH BIAS"

Source: Chris Albon @chrisalbon MachineLearningFlashCards.com

# Loess Regression

```
ggplot(reportCard, aes(x=Pass2012, y=Pass2013)) + geom_point(alpha=0.2) +  
  geom_smooth(method='lm', formula=y~poly(x,2,raw=TRUE), size=2, se=FALSE) +  
  geom_smooth(method='loess', formula = y ~ x, size=2, se=FALSE, color = 'purple') +  
  coord_equal() + ylim(c(0,100)) + xlim(c(0,100))
```



```
library('VisualStats')  
library('ShinyDemo')  
shiny_demo('loess', package = 'VisualStats')
```

See this site for more info:

<https://visualstats.bryer.org/loess.html>



# Shiny App

```
shiny::runGitHub('NYSchools','jbryer',subdir='NYSReportCard')
```

See also the Github repository for more information: <https://github.com/jbryer/NYSchools>

# One Minute Paper

1. What was the most important thing you learned during this class?
2. What important question remains unanswered for you?



<https://forms.gle/ETg8tW9YRHQJHjE28>