Linear Regression Part 2

DATA 606 - Statistics & Probability for Data Analytics

Jason Bryer, Ph.D. and Angela Lui, Ph.D.

April 19, 2023

Announcements

Project Presentations

Please take a look at the sign-up sheet even if you already signed up. We had to move the two Thursday evening time slots to Monday evening.

You can sign up for a presentation slot on this Google Sheet:

https://docs.google.com/spreadsheets/d/1xcWRZuM4OO0ePR7rqO43tuXVGLQnc0asK1tODdE-qfo/edit?usp=sharing

Example project slides located here:

https://github.com/jbryer/DATA606Spring2023/tree/main/Project

One Minute Paper Results

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



What important question remains unanswered for you?

```
multivariate distributed residuals

true appointment true can be considered residuals

true appointment true can be considered residuals

true appointment true
```

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?

reportCard Data Frame

BEDSCODE \$	School \$	NumTested2012	Mean2012 🜲	Pass2012 🜲	Charter 🜲	GradeSubject \$	County \$	BOCES	NumTested2013 🖣	Mean2013 🜲	P
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	

Descriptive Statistics

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

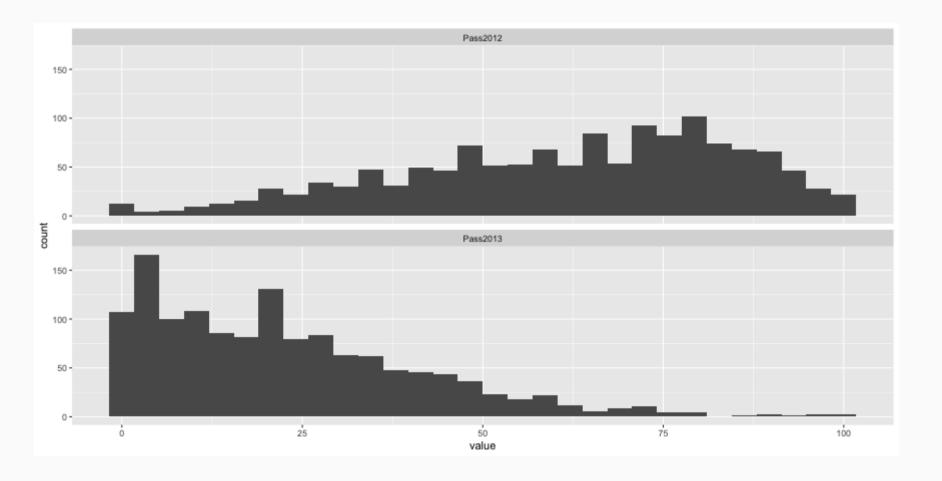
```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.00 46.00 65.00 61.73 80.00 100.00
```

summary(reportCard\$Pass2013)

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

Histograms

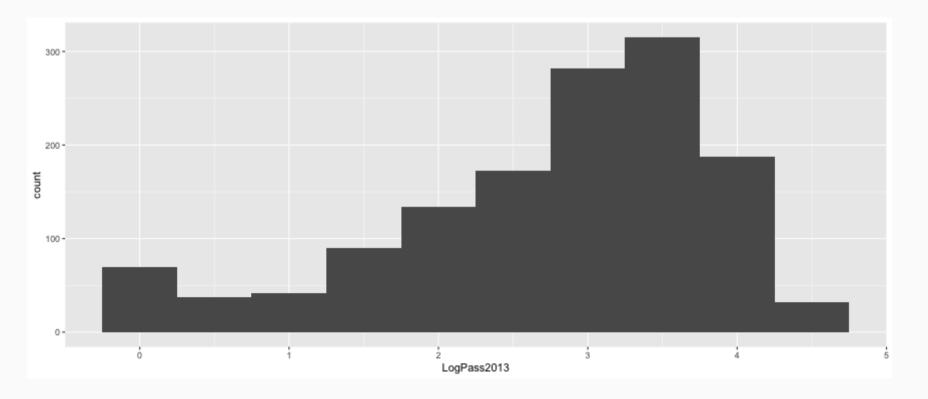
```
melted <- melt(reportCard[,c('Pass2012', 'Pass2013')])
ggplot(melted, aes(x=value)) + geom_histogram() + facet_wrap(~ variable, ncol=1)</pre>
```



Log Transformation

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

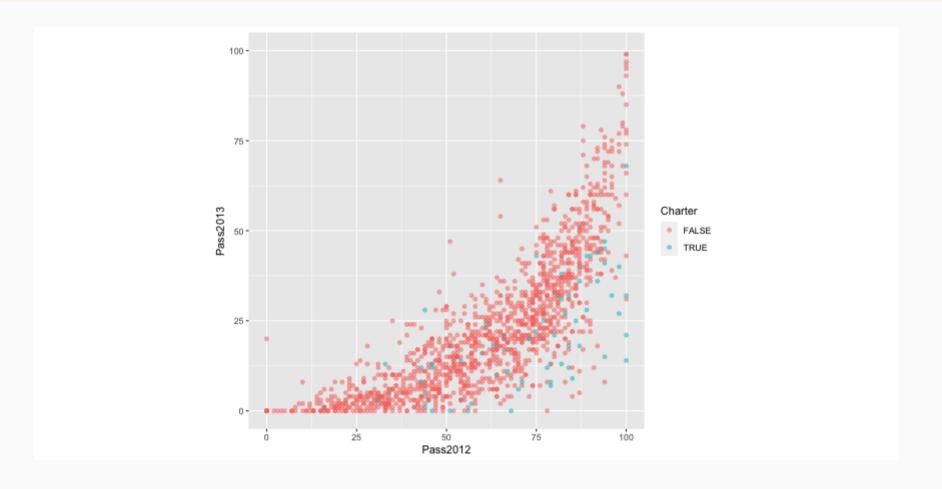
```
reportCard$LogPass2013 <- log(reportCard$Pass2013 + 1)
ggplot(reportCard, aes(x=LogPass2013)) + geom_histogram(binwidth=0.5)</pre>
```





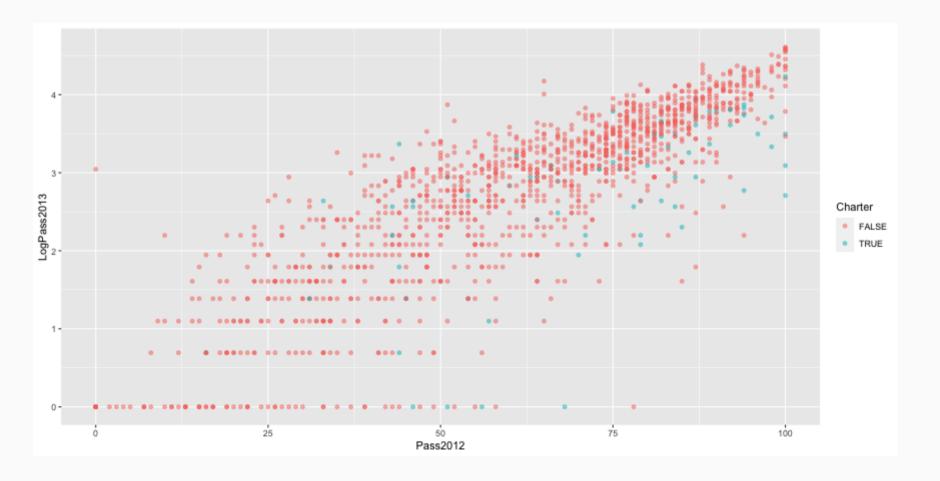
Scatter Plot

```
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</pre>
```

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</pre>
```

Linear Regression

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

```
##
## Call:
## lm(formula = Pass2013 ~ Pass2012, data = reportCard)
##
## Residuals:
   Min 10 Median 30 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)</pre>
```

```
##
## Call:
## lm(formula = LogPass2013 ~ Pass2012, data = reportCard)
##
## Residuals:
   Min
          10 Median 30 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?

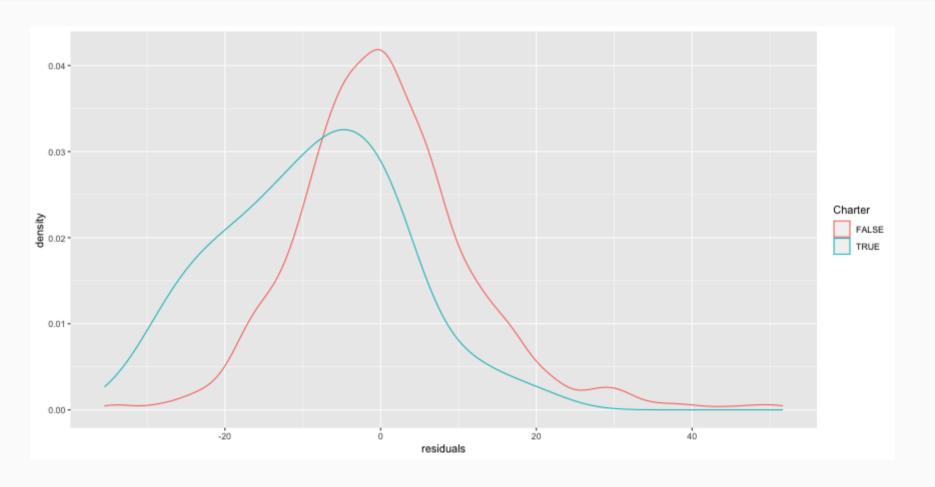
Yes! Whether we log tranform the data or not, the correlations are statistically significant with regression models with \mathbb{R}^2 creater than 62%.

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)</pre>
```

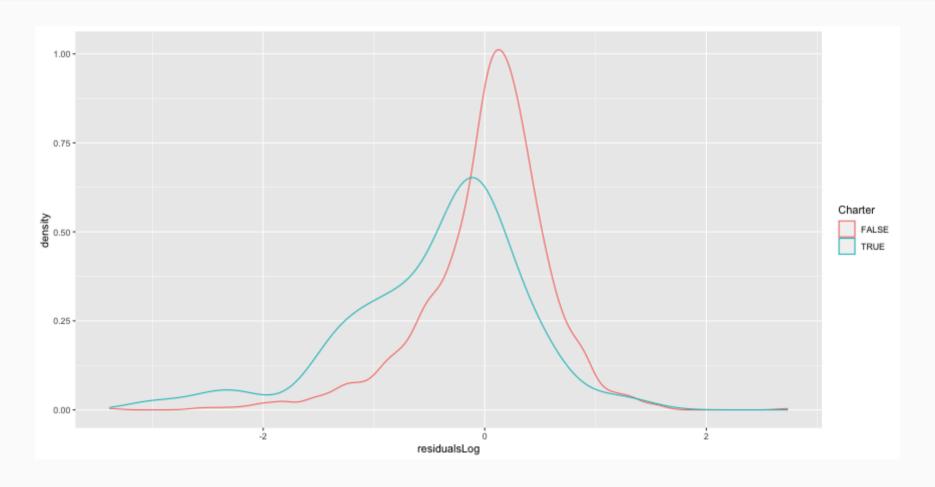
Distribution of Residuals

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 quatric models fairly easily in R, say of the following form:

$$y = b_1 x^2 + b_2 x + b_0$$

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

[1] 0.7065206

summary(lm.out)\$r.squared

[1] 0.6206049

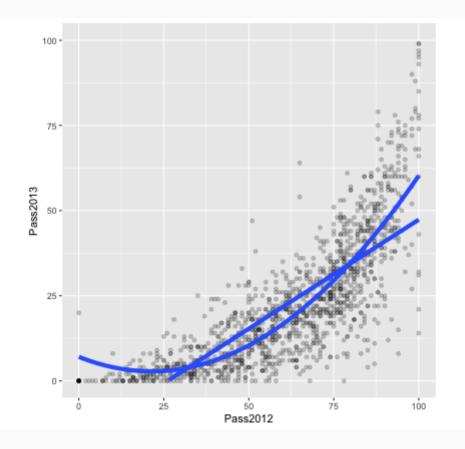
Quadratic Model

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

```
##
## Call:
## lm(formula = Pass2013 ~ I(Pass2012^2) + Pass2012, data = reportCard)
##
## Residuals:
     Min
          10 Median 30 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

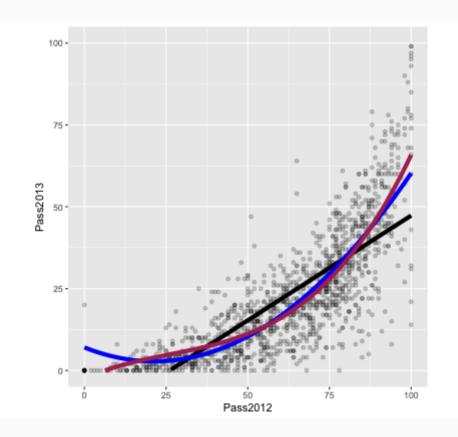
```
ggplot(reportCard, aes(x=Pass2012, y=Pass2013)) + geom_point(alpha=0.2) +
    geom_smooth(method='lm', formula=y ~ x, size=2, se=FALSE) +
    geom_smooth(method='lm', formula=y ~ I(x^2) + x, size=2, se=FALSE) +
    coord_equal() + ylim(c(0,100)) + xlim(c(0,100))
```



Let's go crazy, cubic!

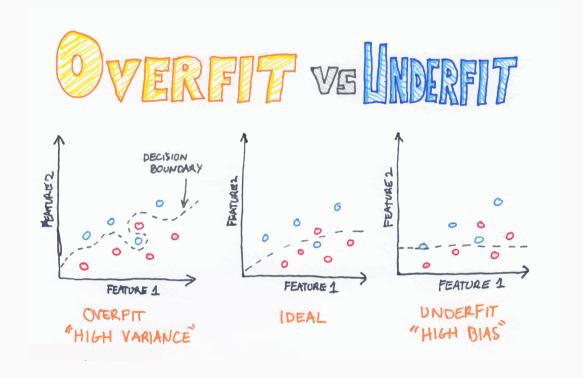
```
cube.out <- lm(Pass2013 ~ I(Pass2012^3) + I(Pass2012^2) + Pass2012, data=reportCard)
summary(cube.out)$r.squared</pre>
```

[1] 0.7168206



Be careful of overfitting...

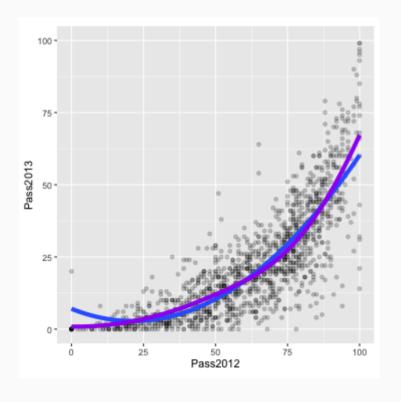




Source: Chris Albon @chrisalbon MachineLearningFlashCards.com

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://jbryer.github.io/VisualStats/articles/loess

Shiny App

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

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

One Minute Paper

Complete the one minute paper:

https://forms.gle/p9xcKcTbGiyYSz368

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