# STAT120 HW 9

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#### Section 9.1

9.4

```
\beta_0 intercept = 7.2777; \beta_0 slope = -0.3560; y = 7.2777 - 0.3560x
```

9.7

Sample Slope = -0.3560;  $H_0: \beta = 0$ ;  $H_a: \beta \neq 0$ ; p-value = 0.087. Since the p-value is greater than the 5% significance value, we fail to reject the null hypothesis.

9.17

a. The scatter plot does not reflect anything that should be of concern.

```
b. y = 2.03 + 0.00189(650) = 3.26
```

- c. The slope is 0.00189, this slope reflects the increase in GPA for every SAT point.
- d. The test statistic if 6.99, the p-value is approximately 0. As such, we can reject the null hypothesis.
- e. The r-squared value is 0.125. This means that 12.5% of the values of students GPAs in this sample is explained by their corresponding verbal SAT score.

9.26

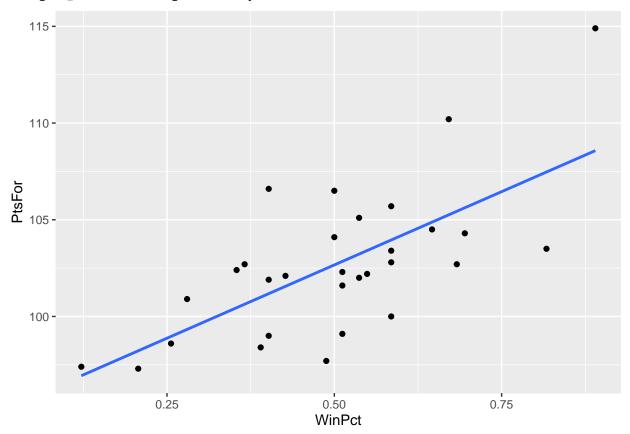
```
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
##
  The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
##
               Df Sum Sq Mean Sq F value
                                            Pr(>F)
##
                  21.05
                                    40.77 6.53e-07 ***
## Beds
                           21.046
               28
                   14.45
                            0.516
## Residuals
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
  a. H_0: \beta = 0; H_a: \beta \neq 0
```

- b. Since we get a r-squared value of 0.593 (59.3%).
- c. We get a f statistic of 40.77, the corresponding p-value is 6.53e-07 (approximately zero)
- d. Since we get such a small p-value, we are able to reject the null hypothesis and find significant evidence that suggests there is correlation between the number of beds and baths in a house in CA. Given the

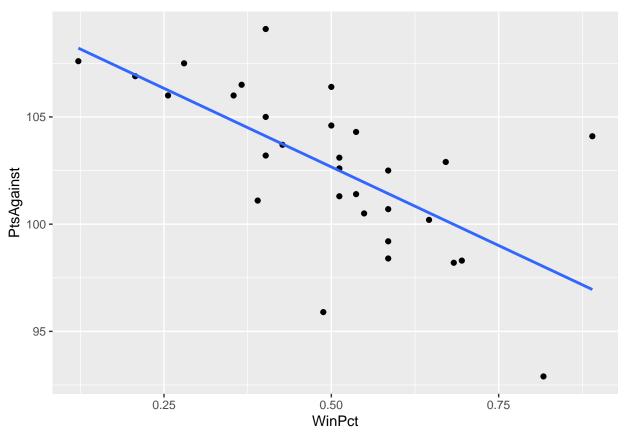
r-squared value of 59.3%, this means that 59.3% of the number of bathrooms of houses in CA in this sample is explained by the number of their bedrooms

## 9.28

## `geom\_smooth()` using formula 'y ~ x'



## `geom\_smooth()` using formula 'y ~ x'



```
##
## Call:
## lm(formula = WinPct ~ PtsFor, data = NBAStandings2016)
## Residuals:
##
       Min
                 1Q
                    Median
                                  3Q
## -0.21858 -0.10572 0.01607 0.07650 0.29177
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
                         0.638332 -4.085 0.000334 ***
## (Intercept) -2.607670
## PtsFor
              0.030270
                         ## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1268 on 28 degrees of freedom
## Multiple R-squared: 0.4587, Adjusted R-squared: 0.4394
## F-statistic: 23.73 on 1 and 28 DF, p-value: 3.941e-05
##
## lm(formula = WinPct ~ PtsAgainst, data = NBAStandings2016)
##
## Residuals:
##
       Min
                 1Q Median
                                  ЗQ
                                         Max
## -0.22935 -0.04506 -0.00873 0.06213 0.43319
##
## Coefficients:
```

```
##
                Estimate Std. Error t value Pr(>|t|)
                           0.657878
                                      5.462 7.85e-06 ***
## (Intercept)
               3.593558
                           0.006404
## PtsAgainst
              -0.030132
                                    -4.705 6.20e-05 ***
## ---
## Signif. codes:
                    '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1288 on 28 degrees of freedom
## Multiple R-squared: 0.4416, Adjusted R-squared: 0.4216
## F-statistic: 22.14 on 1 and 28 DF, p-value: 6.203e-05
```

- a. As one data set shows linear increase while the other shows linear decrease, both graphs seem to fit conditions for a linear model
- b. Equation: WinPct = 3.594 0.030PtsFor, since the p-value (3.941e-05) is so small, we can say that PtsFor is an effective predictor
- c. Equation: WinPct = -2.607 + 0.030PtsAgainst, since the p-value (6.203e-05) is so small, we can say that PtsAgainst is an effective predictor
- d. The r-squared value for PtsFor is 0.4587 while the r-squared value for PtsAgainst is 0.4416. The r-squared value for PtsFor against is larger than that of PtsAgainst.
- e. PtsFor: WinPct = 3.594 0.030(114.9) = 0.147; PtsAgainst: WinPct = -2.607 + 0.030(104.1) = 0.516
- f. Overall, the PtsAgainst seems to be more effective at predicting win percentage. This can be seen from the above example where PtsAginst predicted a 0.516 win percentage, much closer to the actual win percentage as compared to the 0.147 predicted by PtsFor. While the r-squared value for PtsFor is a bit higher than PtsAgainst, it is only approximately 1% and can be seen as not that significant when compared to the huge prediction difference as shown in example e.

#### Section 9.2

9.43

The F-stat is 7.44, the p-value is 0.011. Given this information, we can say that the linear model is appropriate for this. Since the p-value is so small, we can reject the null hypothesis.

#### 9.46

- a. Correlation: -0.366, the p-value is 0.015
- b. The slope is -3.34, the t-value is -2.55, the p-value is 0.015
- c. The f-stat is 6.50, the p-value is 0.015
- d. They are the same
- e. Since our p-value of 0.015 is less than the 5% significance level, we reject the null hypothesis and conclude that there is significant correlation between football and cognition.

```
9.54
```

```
##
  lm(formula = LifeExpectancy ~ Health, data = SampCountries)
##
## Residuals:
##
       Min
                1Q
                    Median
                                 3Q
                                         Max
## -24.034 -5.223
                      2.650
                              7.933
                                     11.874
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
```

```
## (Intercept)
                61.3202
                             4.7024
                                     13.040
                                               <2e-16 ***
## Health
                 0.7286
                             0.3637
                                      2.003
                                               0.0508 .
##
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
## Residual standard error: 10.15 on 48 degrees of freedom
## Multiple R-squared: 0.07715,
                                     Adjusted R-squared:
## F-statistic: 4.013 on 1 and 48 DF, p-value: 0.05083
               Df Sum Sq Mean Sq F value Pr(>F)
##
## Health
                      414
                            413.7
                                    4.013 0.0508 .
                1
## Residuals
                            103.1
               48
                    4948
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
  a. We get a correlation of 0.7286, the associated p-value is 0.05083
  b. The t-stat is 2.003, the associated p-value is 0.05083.
```

- c. We get an f-stat of 4.013 and an associated p-value of 0.0508
- d. Since we have a p-value of 0.0508, it is somewhat higher than a 5% significance level. As such, we can say that this model is not that effective.

Section 9.3

```
9.58
```

A, B; 12

9.63

Confidence Interval: (-0.013, 4.783); Prediction Interval: (-2.797, 7.568)

9.68

```
##
                  Sum Sq Mean Sq F value
                                              Pr(>F)
## Size
                1 9457748 9457748
                                     56.88 3.25e-08 ***
## Residuals
               28 4655513 166268
##
                     '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
                   0
##
          fit.
                   lwr
                             upr
## 1 571.9103 445.2591 698.5614
##
                    lwr
                              upr
## 1 571.9103 -133.2107 1277.031
```

- a. Since we have such a small p-value (3.24e-08), we can say that size (square footage) is an effective predictor.
- b. Price = -369.63 + 470.77(2000) = 941170.37
- c. We get a 90% confidence interval of (445.2591, 698.5614). We are 90% confident that the slope of the linear regression for the house prices with sq footage of 2000 sqft in NY falls between the interval of (419.40, 729.42).
- d. We get a 90% prediction interval of (-133.2107, 1277.031). 90% of all homes in NY with 2000 sq footage have a housing price between the interval of (-133.2107, 1277.031).

9.70

```
a. GPA = 2.03 + 0.00189(500) = 2.975; GPA = 2.03 + 0.00189(700) = 3.353
```

```
##
          fit
                   lwr
                            upr
## 1 2.980022 2.91623 3.043814
##
          fit
                  lwr
                            upr
## 1 2.980022 2.24318 3.716864
##
         fit
                  lwr
                            upr
## 1 3.35861 3.289678 3.427542
##
         fit
                   lwr
                            upr
## 1 3.35861 2.621305 4.095915
```

- b.i. We get a 95% confidence interval of (2.91623, 3.043814). We are 95% confident that the slope of the linear regression for the students GPAs with a 500 verbal SAT score to fall between the interval of (2.91623, 3.043814).
- b.ii. We get a 95% prediction interval of (2.24318, 3.716864). 95% of all students who have a 500 verbal SAT have a GPA between the interval of (2.24318, 3.716864).
- b.iii. We get a 95% confidence interval of (3.289678, 3.427542). We are 95% confident that the slope of the linear regression for the students GPAs with a 700 verbal SAT score to fall between the interval of (3.289678, 3.427542).
- b.iv. We get a 95% prediction interval of (2.621305, 4.095915). 95% of all students who have a 700 verbal SAT have a GPA between the interval of (2.621305, 4.095915).