Homework 6

Devin Etcitty 4/2/2017

STAT4205

```
dce2108
```

```
library(alr4)
library(ggplot2)
library(tidyverse)
```

Problem 7.6

7.6.1 - 7.6.4

stopping dataframe

heteroscedastic homosecdastic

What are theset erms?

```
dim(stopping)
```

```
## [1] 62 2
```

names(stopping)

```
## [1] "Speed" "Distance"
```

sapply(stopping, class)

```
## Speed Distance
## "integer" "integer"
```

head(stopping)

```
##
    Speed Distance
## 1
       4
                2
## 2
       5
## 3
      5
                4
## 4
      5
                8
## 5
       5
                8
       7
## 6
                7
```

plot(stopping)

```
00
    120
                                                                        0
                                                               0
                                                       0
                                                               00
Distance
    80
     9
                                                     0
                                        0
                                      0
                                              0
    4
                                                0
                        20
                0800
           08
             5
                     10
                             15
                                              25
                                                      30
                                     20
                                                               35
                                                                       40
                                       Speed
```

```
qd <- lm(Distance ~ Speed + I(Speed^2),data=stopping)
summary(qd)</pre>
```

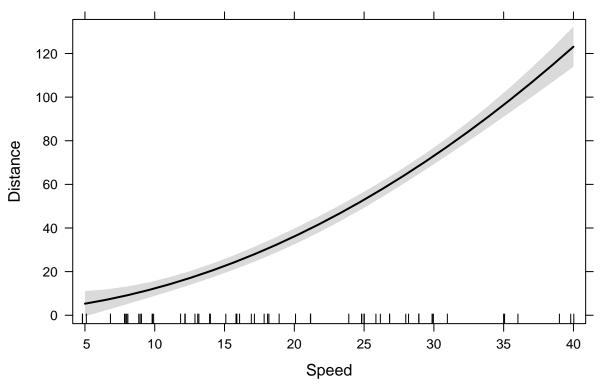
```
##
## Call:
## lm(formula = Distance ~ Speed + I(Speed^2), data = stopping)
##
## Residuals:
##
       Min
                  1Q
                       Median
                                    3Q
                                             Max
## -22.5192 -5.4527 -0.5519
                                3.8442
                                        27.9373
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.58036
                           5.10266
                                     0.310
                                              0.758
## Speed
                0.41607
                           0.55641
                                     0.748
                                               0.458
## I(Speed^2)
                0.06556
                           0.01303
                                     5.033 4.83e-06 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.927 on 59 degrees of freedom
## Multiple R-squared: 0.9144, Adjusted R-squared: 0.9115
## F-statistic: 315.3 on 2 and 59 DF, p-value: < 2.2e-16
p1 <- lm(Distance ~ poly(Speed, 2, raw=TRUE), data = stopping)
summary(p1)
##
## Call:
## lm(formula = Distance ~ poly(Speed, 2, raw = TRUE), data = stopping)
##
## Residuals:
##
       Min
                  1Q
                       Median
                                    3Q
                                             Max
```

3.8442 27.9373

-22.5192 -5.4527 -0.5519

```
##
## Coefficients:
##
                              Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                          5.10266
                                                    0.310
                                                             0.758
                               1.58036
## poly(Speed, 2, raw = TRUE)1 0.41607
                                          0.55641
                                                    0.748
                                                             0.458
## poly(Speed, 2, raw = TRUE)2 0.06556
                                          0.01303
                                                    5.033 4.83e-06 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.927 on 59 degrees of freedom
## Multiple R-squared: 0.9144, Adjusted R-squared: 0.9115
## F-statistic: 315.3 on 2 and 59 DF, p-value: < 2.2e-16
print(plot(Effect("Speed", p1)))
```

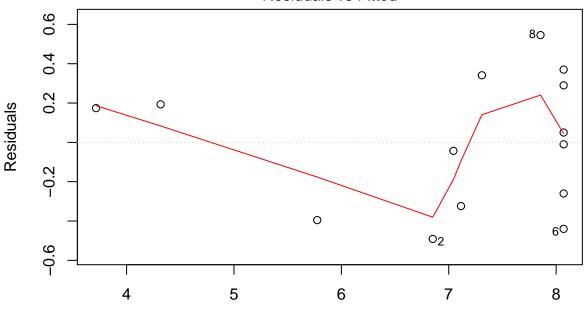
Speed effect plot



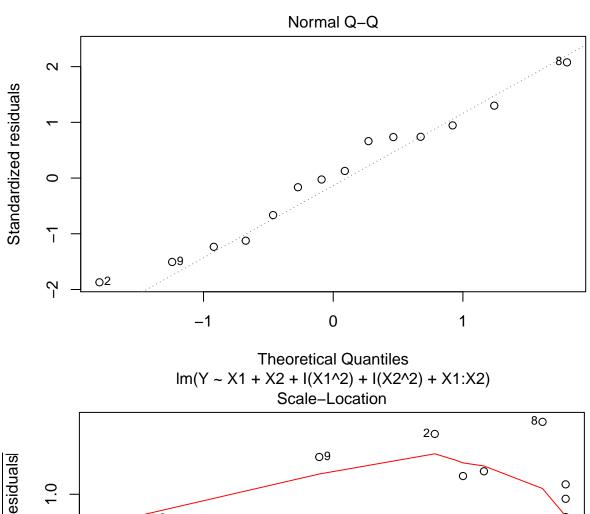
```
mod2 <- lm(Y ~ X1 + X2 + I(X1^2) + I(X2^2) + X1:X2, data=cakes)
summary(mod2)
```

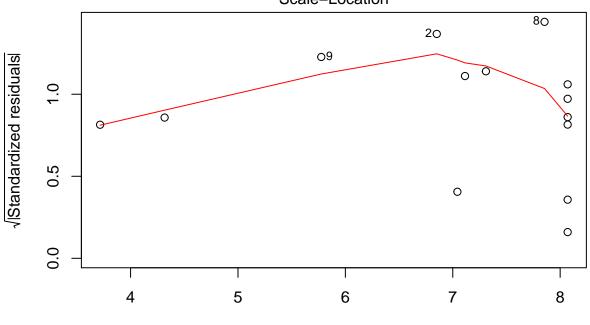
```
## (Intercept) -2.204e+03 2.416e+02 -9.125 1.67e-05 ***
                                      5.563 0.000533 ***
## X1
               2.592e+01 4.659e+00
## X2
               9.918e+00
                          1.167e+00
                                      8.502 2.81e-05 ***
## I(X1^2)
              -1.569e-01
                         3.945e-02
                                    -3.977 0.004079 **
                          1.578e-03
## I(X2^2)
              -1.195e-02
                                     -7.574 6.46e-05 ***
## X1:X2
              -4.163e-02
                          1.072e-02
                                     -3.883 0.004654 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4288 on 8 degrees of freedom
## Multiple R-squared: 0.9487, Adjusted R-squared: 0.9167
## F-statistic: 29.6 on 5 and 8 DF, p-value: 5.864e-05
plot(mod2)
```

Residuals vs Fitted

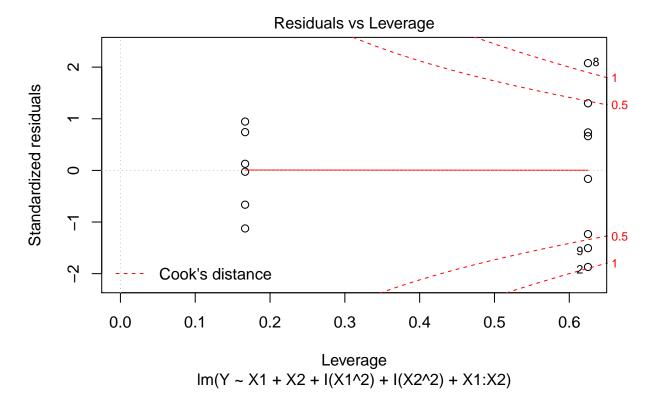


Fitted values $Im(Y \sim X1 + X2 + I(X1^2) + I(X2^2) + X1:X2)$





Fitted values $Im(Y \sim X1 + X2 + I(X1^2) + I(X2^2) + X1:X2)$



```
vcov(mod2)
```

```
##
                 (Intercept)
                                        Х1
                                                      X2
                                                               I(X1^2)
## (Intercept) 58361.2541243 -6.670524e+02 -2.668210e+02 2.488704e+00
                -667.0524374 2.170546e+01 1.642140e+00 -1.089175e-01
## X2
                -266.8209749 1.642140e+00 1.360860e+00 -3.351307e-03
## I(X1^2)
                   2.4887042 -1.089175e-01 -3.351307e-03 1.555964e-03
## I(X2^2)
                   0.3106805 -3.351307e-04 -1.742679e-03 4.787581e-06
## X1:X2
                   1.4075488 -4.021568e-02 -4.021568e-03 7.495493e-17
##
                     I(X2^2)
                                     X1:X2
## (Intercept) 3.106805e-01 1.407549e+00
## X1
               -3.351307e-04 -4.021568e-02
## X2
               -1.742679e-03 -4.021568e-03
## I(X1^2)
                4.787581e-06 7.495493e-17
## I(X2^2)
                2.489542e-06 -3.702888e-18
               -3.702888e-18 1.149019e-04
## X1:X2
```

sqrt(diag(vcov(mod2)))

```
## (Intercept) X1 X2 I(X1^2) I(X2^2)
## 2.415807e+02 4.658911e+00 1.166559e+00 3.944571e-02 1.577828e-03
## X1:X2
## 1.071923e-02
```

summary(mod2)\$coef

```
## (Intercept) -2204.484987 2.415807e+02 -9.125251 1.673889e-05

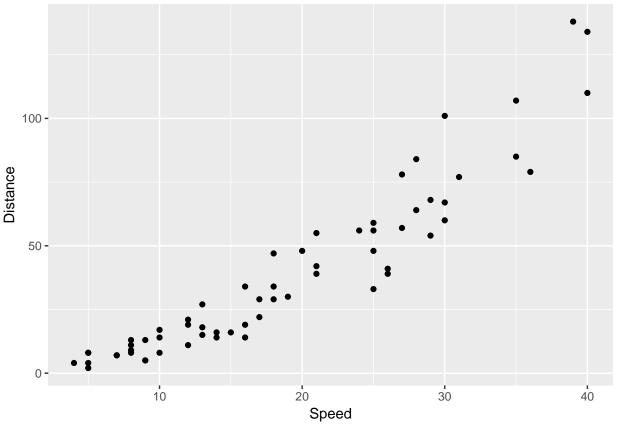
## X1 25.917558 4.658911e+00 5.563007 5.327892e-04

## X2 9.918267 1.166559e+00 8.502156 2.810176e-05

## I(X1^2) -0.156875 3.944571e-02 -3.976985 4.078860e-03

## I(X2^2) -0.011950 1.577828e-03 -7.573701 6.462492e-05
```

```
-0.041625 1.071923e-02 -3.883207 4.653652e-03
## X1:X2
z <- with(cakes, polym(X1, X2, degree=2, raw=TRUE))
\#summary(z)
model.matrix(~ z)
      (Intercept)
                      z1.0
##
                              z2.0
                                       z0.1
                                                 z1.1
                                                          z0.2
## 1
                1 33.00000 1089.00 340.0000 11220.00 115600.0
## 2
                1 37.00000 1369.00 340.0000 12580.00 115600.0
## 3
                1 33.00000 1089.00 360.0000 11880.00 129600.0
                1 37.00000 1369.00 360.0000 13320.00 129600.0
## 4
                1 35.00000 1225.00 350.0000 12250.00 122500.0
## 5
## 6
                1 35.00000 1225.00 350.0000 12250.00 122500.0
## 7
                1 35.00000 1225.00 350.0000 12250.00 122500.0
## 8
                1 37.82843 1430.99 350.0000 13239.95 122500.0
## 9
                1 32.17157 1035.01 350.0000 11260.05 122500.0
## 10
                1 35.00000 1225.00 364.1421 12744.97 132599.5
## 11
               1 35.00000 1225.00 335.8579 11755.03 112800.5
                1 35.00000 1225.00 350.0000 12250.00 122500.0
## 12
## 13
                1 35.00000 1225.00 350.0000 12250.00 122500.0
                1 35.00000 1225.00 350.0000 12250.00 122500.0
## 14
## attr(,"assign")
## [1] 0 1 1 1 1 1
ggplot(stopping, aes(Speed, Distance)) +
 geom_point()
```



```
This graph supports a quadratic regression model because the graph has a curve and less linear.
7.6.2
7.6.3
7.6.4
head(stopping)
     Speed Distance
##
## 1
         4
## 2
         5
                  2
## 3
         5
                  4
         5
                  8
## 4
         5
                  8
## 5
## 6
         7
                  7
summary(stopping)
##
        Speed
                        Distance
          : 4.00
                          : 2.00
##
   Min.
                    Min.
##
   1st Qu.:10.00
                    1st Qu.: 13.25
## Median :17.50
                    Median : 29.50
           :18.92
                          : 39.31
## Mean
                    Mean
## 3rd Qu.:26.75
                     3rd Qu.: 56.75
           :40.00
## Max.
                    Max.
                            :138.00
Problem 7.8
jevons dataframe
dim(jevons)
## [1] 5 6
names(jevons)
## [1] "Age"
                          "Weight" "SD"
                 "n"
                                             "Min"
                                                      "Max"
sapply(jevons, class)
                           Weight
                                         SD
                                                   Min
                                                              Max
## "integer" "integer" "numeric" "numeric" "numeric" "numeric"
head(jevons)
##
     Age
           n Weight
                          SD
                               Min
                                     Max
      1 123 7.9725 0.01409 7.900 7.999
## 2
       2 78 7.9503 0.02272 7.892 7.993
       3 32 7.9276 0.03426 7.848 7.984
## 3
## 4
       4 17 7.8962 0.04057 7.827 7.965
       5 24 7.8730 0.05353 7.757 7.961
summary(jevons)
##
         Age
                                     Weight
                                                        SD
                       n
```

Min.

:7.873

1st Qu.:7.896

Median :7.928

:0.01409

1st Qu.:0.02272

Median : 0.03426

Min.

1st Qu.:2

Median :3

:1

Min.

: 17.0

1st Qu.: 24.0

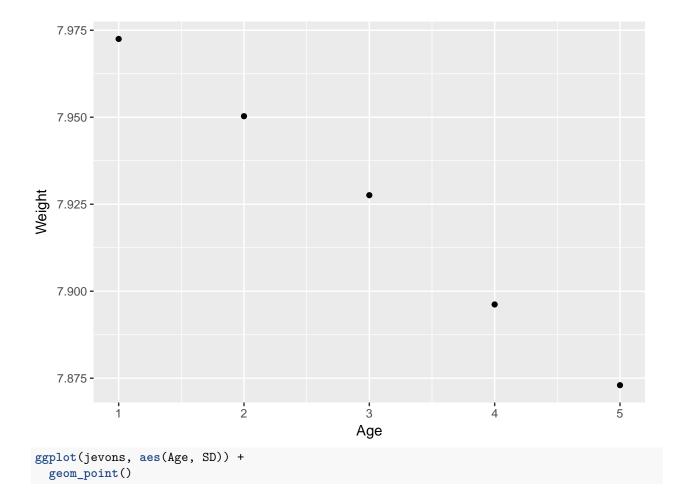
Median: 32.0

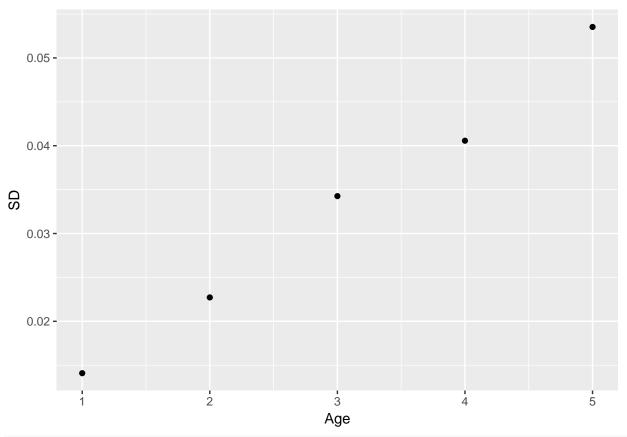
Min.

##

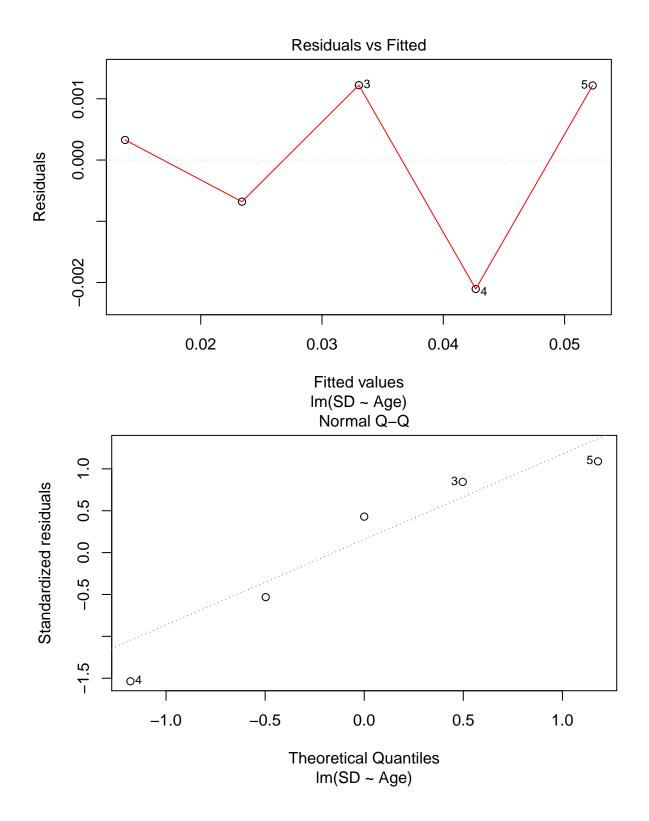
```
Mean :3
              Mean : 54.8 Mean
                                      :7.924
                                               Mean
                                                      :0.03303
   3rd Qu.:4
              3rd Qu.: 78.0 3rd Qu.:7.950
                                               3rd Qu.:0.04057
##
##
   Max.
         :5
              Max.
                     :123.0 Max.
                                      :7.973
                                               Max.
                                                      :0.05353
##
        Min
                        Max
   Min.
          :7.757
                          :7.961
##
                   Min.
##
   1st Qu.:7.827
                   1st Qu.:7.965
  Median :7.848
                   Median :7.984
         :7.845
                          :7.980
## Mean
                   Mean
   3rd Qu.:7.892
                   3rd Qu.:7.993
##
## Max. :7.900
                   Max.
                          :7.999
plot(jevons)
               20 60 100
                                        0.02 0.04
                                                               7.96
                                                                    7.98 8.00
                                                                             2
                                                                  0
      Age
                                          0
                                                              0
                                                                         0
                    n
20
        0
                                 0
                                             0
                                                           0
                                                                       0
                                                              o
                            Weight
        0
                 0
                                             0
                                                           0
          0
                                                                  0
0.02 0.05
                                                         ۰ 。
                                                                  0
                                           SD
                                                              0
                                                                         0
                                             ° 0
                 0
                                                       Min
                                                           0
                                                                   Max
7.96
                                               0
   1 2 3 4 5
                           7.88
                                 7.94
                                                    7.76
                                                         7.84
```

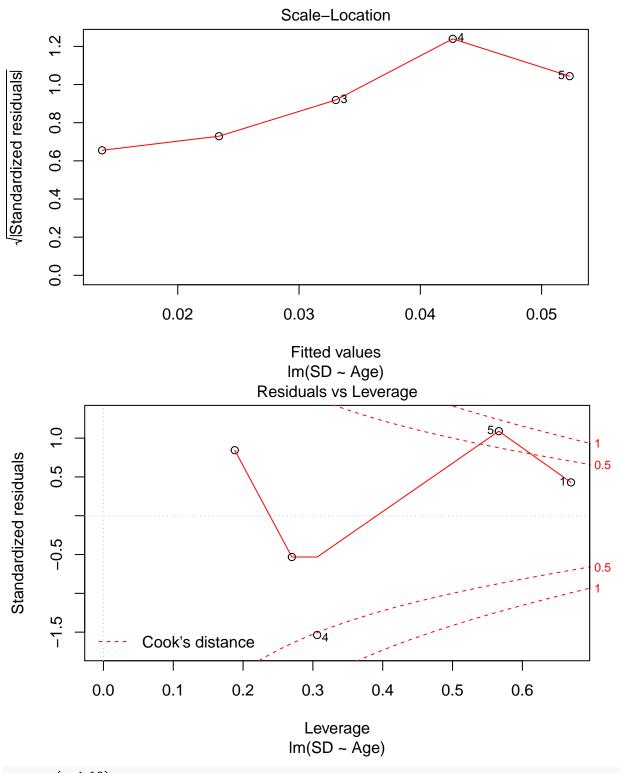
ggplot(jevons, aes(Age, Weight)) +
 geom_point()





```
model.1 <- lm(SD ~ Age, data=jevons)
wts <- (nrow(jevons) * fitted(lm(abs(residuals(model.1)) ~ jevons$SD))^2) / (jevons$SD)^2
model2 <- lm(SD ~ Age, data=jevons, weights=wts)
plot(model2)</pre>
```





```
summary(model2)
```

```
##
## Call:
## lm(formula = SD ~ Age, data = jevons, weights = wts)
##
```

```
## Weighted Residuals:
##
                                3
           1
                     2
## 2.967e-05 -5.463e-05 9.147e-05 -1.539e-04 8.628e-05
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.0041245 0.0014901
                                  2.768 0.069687 .
              0.0096376 0.0004796 20.095 0.000269 ***
## Age
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.0001202 on 3 degrees of freedom
## Multiple R-squared: 0.9926, Adjusted R-squared: 0.9902
## F-statistic: 403.8 on 1 and 3 DF, p-value: 0.0002694
summary(jevons)
##
                                  Weight
                                                   SD
        Age
                    : 17.0 Min.
                                     :7.873
                                                     :0.01409
##
   Min.
         :1
              Min.
                                              Min.
              1st Qu.: 24.0
   1st Qu.:2
                              1st Qu.:7.896
                                              1st Qu.:0.02272
  Median :3
             Median: 32.0
                            Median :7.928
                                              Median :0.03426
              Mean : 54.8
##
   Mean :3
                              Mean :7.924
                                              Mean
                                                     :0.03303
##
   3rd Qu.:4
              3rd Qu.: 78.0
                              3rd Qu.:7.950
                                              3rd Qu.:0.04057
##
   Max. :5
             Max. :123.0
                             Max. :7.973
                                              Max.
                                                    :0.05353
##
        Min
                       Max
##
          :7.757
                         :7.961
  Min.
                  Min.
##
  1st Qu.:7.827
                  1st Qu.:7.965
## Median :7.848
                 Median :7.984
## Mean :7.845
                  Mean :7.980
## 3rd Qu.:7.892
                   3rd Qu.:7.993
## Max. :7.900
                 Max.
                         :7.999
Problem 7.12
mile dataframe
head(mile)
    Year Time
                         Name Country Place Gender
## 1 1861 286.0
                   N.S. Greene
                                  IRL <NA>
                                              Male
## 2 1862 273.0
                                  IRL <NA>
                                              Male
                 George Farran
## 3 1868 269.8 Walter Chinnery
                                  GBR <NA>
                                              Male
## 4 1868 268.8
                                  GBR <NA>
                 William Gibbs
                                              Male
```

GBR <NA>

GBR <NA>

Male

Male

5 1873 268.6 Charles Gunton

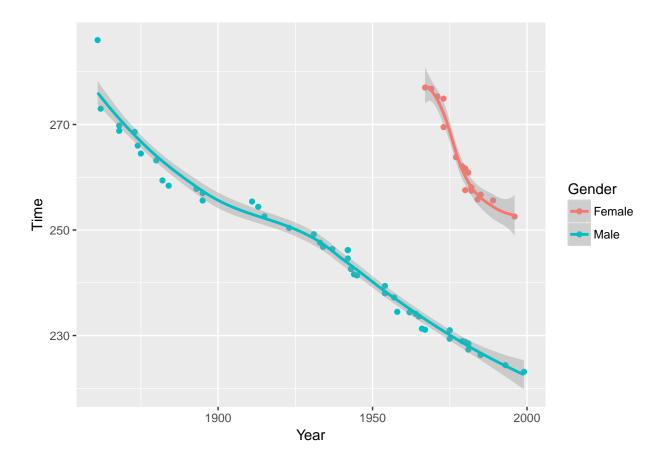
Walter Slade

6 1874 266.0

plot(mile)

```
230 260
                                             5 10
                                                                    1.0 1.4 1.8
      Year
                                                                                  860
                   Time
                               Name
                 a a
                             ૢૻૼૺૹૺઌૢૼૹઌ<u>ૢૼ</u>
                                           Country
                                                         ၀ ၀ထာထာဝ
                               80 80 80 V
                                                          Place
                                           \omega\omega\omega\sigma\sigma
                                                        Gender
                                            റ നാറാന്ത്യ d
      1940
  1860
                             0
                                 20
                                      40
                                                       0 10 20 30
summary(mile)
##
         Year
                         Time
                                                            Country
                                                 Name
##
           :1861
                   Min.
                           :223.1
                                    Mary Decker
                                                   : 4
                                                                 :19
    Min.
                                                         GBR
                   1st Qu.:235.2
##
    1st Qu.:1917
                                    Arne Andersson: 3
                                                         USA
                                                                 :12
    Median:1960
                   Median :252.6
                                    Gunder Hagg
                                                         SWE
                                                                 : 6
##
                                                  : 3
                                    Sebastian Coe : 3
##
   Mean :1945
                   Mean
                         :250.3
                                                         NZL
                                                                 : 4
                   3rd Qu.:261.5
                                    Walter George : 3
##
    3rd Qu.:1980
                                                         ROM
                                                                 : 4
    Max. :1999
                           :286.0
                                    Jim Ryun
                                                   : 2
                                                         AUS
                                                                 : 2
##
                   Max.
                                    (Other)
##
                                                   :44
                                                         (Other):15
          Place
                      Gender
##
                   Female:16
##
    London
            : 4
    Auckland: 3
                   Male :46
##
##
    Goteborg : 3
##
   Oslo
             : 3
   Stockholm: 3
##
##
    (Other) :35
   NA's
             :11
ggplot(mile, aes(Year, Time, color=Gender)) +
 geom_point() +
 geom_smooth(method=)
```

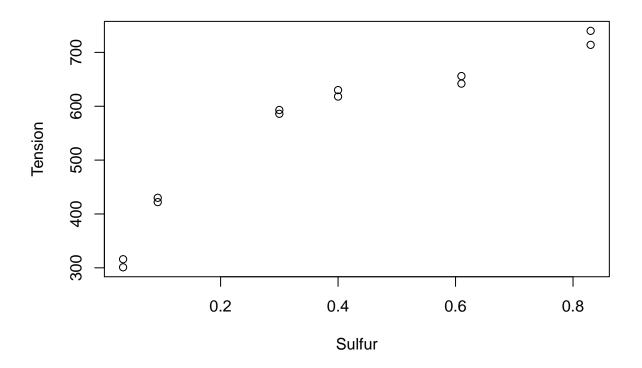
`geom_smooth()` using method = 'loess'



Problem 8.1

```
baeskel dataframe
```

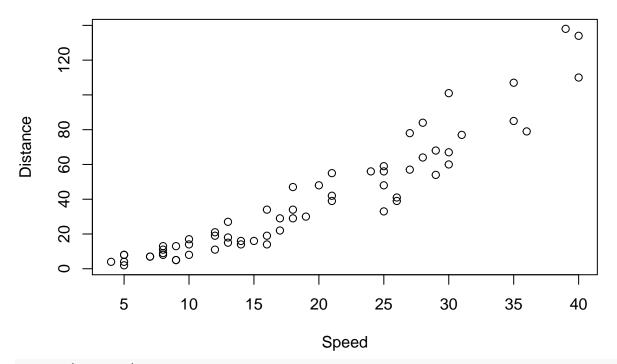
```
dim(baeskel)
## [1] 12 2
names(baeskel)
## [1] "Sulfur" "Tension"
sapply(baeskel, class)
      Sulfur
               Tension
## "numeric" "integer"
head(baeskel)
     Sulfur Tension
##
## 1 0.034
                301
## 2 0.034
                316
## 3 0.093
                430
## 4 0.093
                422
## 5 0.300
                593
## 6 0.300
                586
plot(baeskel)
```



Problem 8.2

```
stopping dataframe
```

```
dim(stopping)
## [1] 62 2
names(stopping)
## [1] "Speed"
                  "Distance"
sapply(stopping, class)
       Speed Distance
## "integer" "integer"
head(stopping)
     Speed Distance
##
## 1
         4
         5
                  2
## 2
## 3
         5
                  4
## 4
         5
                  8
## 5
         5
                  8
         7
## 6
plot(stopping)
```



summary(stopping)

```
##
       Speed
                      {\tt Distance}
   Min. : 4.00
                  Min. : 2.00
##
##
   1st Qu.:10.00
                   1st Qu.: 13.25
   Median :17.50
                   Median : 29.50
   Mean :18.92
                   Mean : 39.31
##
   3rd Qu.:26.75
                   3rd Qu.: 56.75
##
##
  Max.
         :40.00
                          :138.00
                   Max.
```

Problem 8.3

water dataframe

1 1948 9.13 3.58

2 1949 5.28 4.82

3 1950 4.20 3.77

```
dim(water)
## [1] 43 8
names(water)
## [1] "Year"
                 "APMAM"
                           "APSAB"
                                     "APSLAKE" "OPBPC"
                                                          "OPRC"
                                                                    "OPSLAKE"
## [8] "BSAAM"
sapply(water, class)
                 APMAM
                           APSAB
                                  APSLAKE
                                               OPBPC
                                                           OPRC
       Year
## "integer" "numeric" "numeric" "numeric" "numeric" "numeric" "numeric"
##
       BSAAM
## "integer"
head(water)
    Year APMAM APSAB APSLAKE OPBPC OPRC OPSLAKE
                                                   BSAAM
```

6.47

10.26 67567

11.35 66161

54235

3.91 4.10 7.43

5.20 7.55 11.11

3.67 9.52 12.20

```
## 4 1951 4.60 4.46 3.93 11.14 15.15 11.13 68094
## 5 1952 7.15 4.99 4.88 16.34 20.05 22.81 107080
## 6 1953 9.70 5.65 4.91 8.88 8.15 7.41 67594
```

plot(water)

