

(Very) Quick Introduction to Linear Models in R

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Preliminaries

```
> # clears objects in R workspace
> rm(list = ls())

> # load needed packages
> library(FSA)           # for headtail(), filterD(), fitPlot()
> library(dplyr)         # for mutate(), %>%
> library(multcomp)      # for glht(), mcp()

> ruf <- read.csv("RuffeSLRH.csv") %>%
  mutate(logW=log10(wt), logL=log10(tl), fYear=factor(year))
> headtail(ruf)
   fishID year month day  tl  wt    logW    logL fYear
1    1092 1988     6   1  71  6.0 0.7781513 1.851258 1988
2    1097 1988     6   1  74  6.0 0.7781513 1.869232 1988
3    1132 1988     6   1  75  6.0 0.7781513 1.875061 1988
9949     99 2007     9  20 115 17.9 1.2528530 2.060698 2007
9950     88 2007     9  20 120 18.6 1.2695129 2.079181 2007
9951     60 2007     9  20 134 24.6 1.3909351 2.127105 2007
```

Simple Linear Regression

```
> ruf90 <- filterD(ruf, year==1990)
> fit1 <- lm(logW~logL, data=ruf90)
> coef(fit1)
(Intercept)      logL
-4.936979      3.036223

> confint(fit1)
              2.5 %      97.5 %
(Intercept) -4.996232 -4.877726
logL         3.007323  3.065123

> anova(fit1)
Analysis of Variance Table

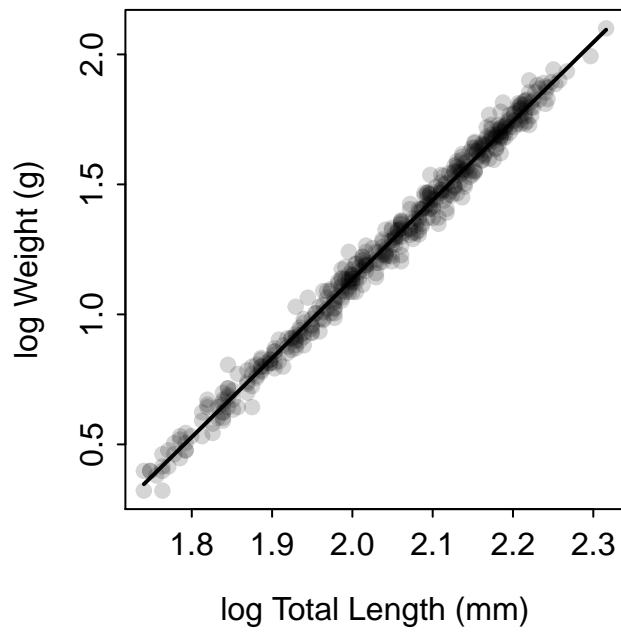
Response: logW
      Df Sum Sq Mean Sq F value    Pr(>F)
logL    1  69.661   69.661   42628 < 2.2e-16
Residuals 455   0.744    0.002

> ( tmp <- range(ruf90$logL) )
[1] 1.740363 2.315970

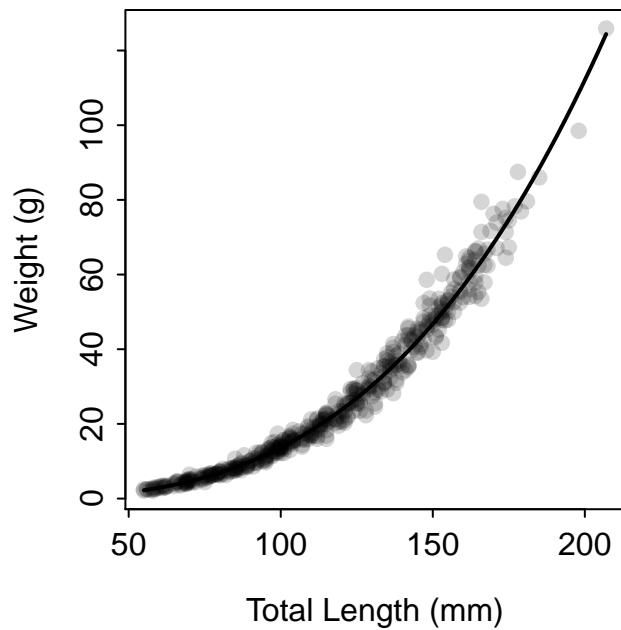
> xs <- seq(tmp[1], tmp[2], length.out=99)
> xs[1:10]
[1] 1.740363 1.746236 1.752110 1.757983 1.763857 1.769730 1.775604 1.781478 1.787351 1.793225

> ys <- predict(fit1, data.frame(logL=xs))
> ys[1:10]
      1      2      3      4      5      6      7      8      9     10
0.3471506 0.3649840 0.3828174 0.4006508 0.4184842 0.4363176 0.4541510 0.4719844 0.4898178 0.5076512
```

```
> plot(logW~logL,data=ruf90,pch=19,col=rgb(0,0,0,1/6),ylab="log Weight (g)",xlab="log Total Length (mm)")
> lines(ys~xs,lwd=2)
```



```
> plot(wt~tl,data=ruf90,pch=19,col=rgb(0,0,0,1/6),ylab="Weight (g)",xlab="Total Length (mm)")
> btxs <- 10^xs
> btys <- 10^ys
> lines(btys~btxs,lwd=2)
```



Dummy Variable Regression (aka ANCOVA)

```
> ruf9000 <- filterD(ruf,year %in% c(1990,2000))
> fit2 <- lm(logW~logL*fYear,data=ruf9000)
> anova(fit2)
Analysis of Variance Table

Response: logW
      Df Sum Sq Mean Sq  F value    Pr(>F)
logL    1 127.429  127.429 74533.553 < 2.2e-16
fYear    1   0.757    0.757  443.043 < 2.2e-16
logL:fYear 1   0.112    0.112   65.463 1.952e-15
Residuals 882   1.508    0.002

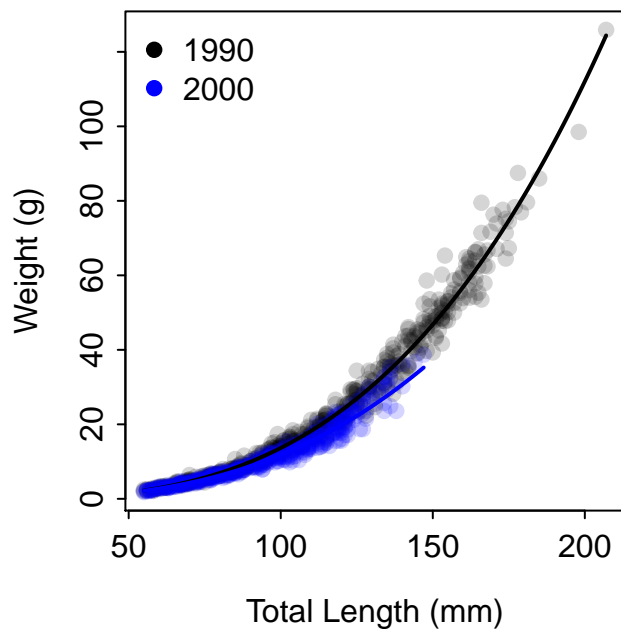
> coef(fit2)
      (Intercept)          logL      fYear2000 logL:fYear2000
      -4.9369786        3.0362230        0.3319951       -0.1976742

> confint(fit2)
              2.5 %      97.5 %
(Intercept) -4.9975070 -4.8764502
logL         3.0067013  3.0657446
fYear2000    0.2362067  0.4277836
logL:fYear2000 -0.2456248 -0.1497235

> tmp <- ruf9000 %>% group_by(fYear) %>% summarize(min=min(tl,na.rm=TRUE),max=max(tl,na.rm=TRUE))
> tmp
Source: local data frame [2 x 3]

   fYear min max
1  1990  55 207
2  2000  55 147

> # base plot
> clrs <- c(rgb(0,0,0,1/6),rgb(0,0,1,1/6))
> plot(wt~tl,data=ruf9000,pch=19,col=clrs[ruf9000$fYear],ylab="Weight (g)",xlab="Total Length (mm)")
>
> # plot line for 1990
> tmpx <- seq(tmp$min[1],tmp$max[1],length.out=99)
> tmpy <- 10^(predict(fit2,data.frame(logL=log10(tmpx),fYear=factor(1990))))
> lines(tmpy~tmpx,lwd=2)
>
> # plot line for 2000
> tmpx <- seq(tmp$min[2],tmp$max[2],length.out=99)
> tmpy <- 10^(predict(fit2,data.frame(logL=log10(tmpx),fYear=factor(2000))))
> lines(tmpy~tmpx,col="blue",lwd=2)
>
> # add a legend
> legend("topleft",c("1990","2000"),pch=19,col=c("black","blue"),bty="n")
```



1-way ANOVA

```
> ruf2 <- filterD(ruf, year %in% c(1990, 1995, 2000, 2006))
> fit3 <- lm(tl ~ fYear, data = ruf2)
> anova(fit3)
Analysis of Variance Table
```

Response: tl

| | Df | Sum Sq | Mean Sq | F value | Pr(>F) |
|-----------|------|--------|---------|---------|-----------|
| fYear | 3 | 120795 | 40265 | 57.25 | < 2.2e-16 |
| Residuals | 1218 | 856640 | 703 | | |

```
> mc1 <- glht(fit3, mcp(fYear = "Tukey"))
> summary(mc1)
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: `lm(formula = tl ~ fYear, data = ruf2)`

Linear Hypotheses:

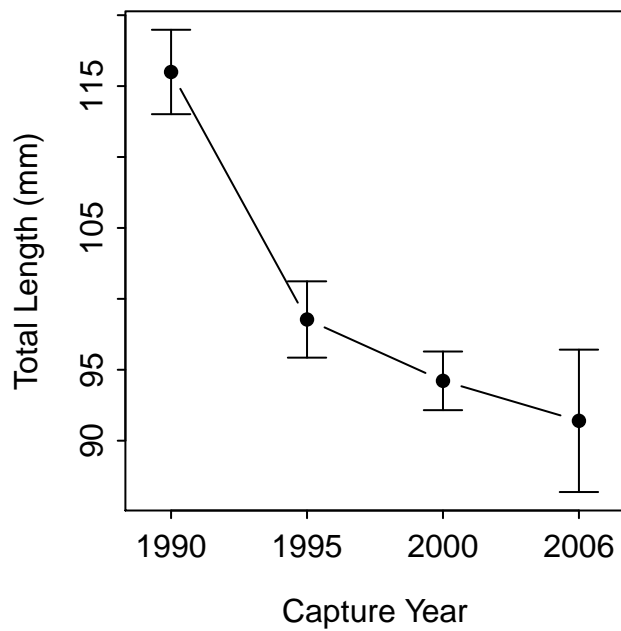
| | Estimate | Std. Error | t value | Pr(> t) |
|------------------|----------|------------|---------|----------|
| 1995 - 1990 == 0 | -17.451 | 1.969 | -8.864 | <0.001 |
| 2000 - 1990 == 0 | -21.777 | 1.783 | -12.215 | <0.001 |
| 2006 - 1990 == 0 | -24.596 | 4.651 | -5.288 | <0.001 |
| 2000 - 1995 == 0 | -4.326 | 1.994 | -2.169 | 0.119 |
| 2006 - 1995 == 0 | -7.145 | 4.736 | -1.509 | 0.410 |
| 2006 - 2000 == 0 | -2.819 | 4.662 | -0.605 | 0.925 |

(Adjusted p values reported -- single-step method)

```
> cld(mc1)
```

```
1990 1995 2000 2006
"b"  "a"  "a"  "a"
```

```
> fitPlot(fit3,ylab="Total Length (mm)",xlab="Capture Year")
```



Application Assignment

Create a script that performs the following tasks:

1. Load the `BLGLW.CSV` data into an R data.frame.
2. Determine if there is a significant relationship between the weight and length of Bluegill in these data.
3. Determine if there is a significant difference between sexes in the relationship between the weight and length of Bluegill.
4. Determine if there is a significant difference among the three lakes in the mean weight of Bluegill.

Save your script!