

Regression Models: Assignment 1

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November 24th, 2020

Importing libraries

```
library(dplyr)
library(MuMIn)
```

Exercise 1

Simulation

```
sim = list()
for (j in 1:1000) {
  vals = c()
  for (i in 1:100) {
    run = 3 + 3*cos(i/10 + 50) + rnorm(1, mean=0, sd=1)
    vals = c(vals, run)
  }
  sim[[j]] = vals
}
sim
```

Exercise 2

Importing the data

```
d <- data.frame(read.table('../data/index.txt', header=TRUE))
```

```
X = d$PovPct
Y = d$Brth15to17
beta1 = cov(X, Y)/var(X)
beta0 = mean(Y) - beta1*mean(X)
beta1
```

```
## [1] 1.373345
```

```
beta0
```

```
## [1] 4.267293
```

Exercise 3

First we have the log-likelihood function for β and σ^2

$$l(\sigma^2|X) = \sum_{i=1}^n \log\left(\frac{1}{\sqrt{2\pi\sigma^2}} - \frac{(Y_i - (\beta_0 + \beta_1 x_{i1} + \dots + \beta_k x_{ik}))^2}{2\sigma^2}\right)$$

Exercise 4

Exercise 5

```
bodyfat <- data.frame(read.table('../data/bodyfat.txt', header=TRUE))
modall <- lm(hwfat ~ ., data = bodyfat)
summary(modall)
```

```
##
## Call:
## lm(formula = hwfat ~ ., data = bodyfat)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.162  -1.858  -0.464   2.502   8.177
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  13.29370    9.63027   1.380  0.1718
## age          -0.32893    0.32158  -1.023  0.3098
## ht           -0.06731    0.16051  -0.419  0.6762
## wt           -0.01365    0.02591  -0.527  0.5999
## abs           0.37142    0.08837   4.203 7.55e-05 ***
## triceps       0.38743    0.13761   2.815  0.0063 **
## subscap       0.11405    0.14193   0.804  0.4243
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.028 on 71 degrees of freedom
## Multiple R-squared:  0.8918, Adjusted R-squared:  0.8827
## F-statistic: 97.54 on 6 and 71 DF,  p-value: < 2.2e-16
```

The sum of residuals is zero:

```
residuals <- sum(resid(modall))
```

The sum of the observed data is equal to the sum of the fitted values

```
Y_hat <- predict(modall, bodyfat[1:length(names(bodyfat))-1])
sum(bodyfat$hwfat) - sum(Y_hat)
```

```
## [1] 4.547474e-13
```

The residuals are orthogonal to the predictors

```
sum(residuals*bodyfat[1:length(names(bodyfat))-1])
```

```
## [1] -3.077268e-10
```

The residuals are orthogonal to the fitted values

```
sum(residuals*Y_hat)
```

```
## [1] -1.568657e-11
```

Exercise 6

```
# rsq <- function(x,y) cor(x,y)^2
# cols <- names(bodyfat)[1:length(names(bodyfat))-1]
# r_2 <- c()
# names(r_2) <- cols
# for (i in 1:length(cols)) {
#   modall <- lm(hwfat ~ cols[i], bodyfat)
#   r_2 <- rsq(predict(hwfat))
# }
# r2
```

```
options(na.action = "na.fail")
modall <- lm(hwfat ~., data = bodyfat)
combs <- dredge(modall)
```

```
## Fixed term is "(Intercept)"
```

```
combs
```

```
## Global model call: lm(formula = hwfat ~ ., data = bodyfat)
```

```
## ---
```

```
## Model selection table
```

	(Intrc)	abs	age	ht	sbscp	trcps	wt	df	logLik	AICc
## 20	10.6200	0.3564	-0.5331			0.4656		5	-194.268	399.4
## 50	5.1120	0.4235				0.4918	-0.028140	5	-194.664	400.2
## 22	13.6200	0.3674		-0.17460		0.4733		5	-194.986	400.8
## 24	15.6100	0.3702	-0.4237	-0.10200		0.4554		6	-193.854	400.9
## 52	9.9400	0.3997	-0.3838			0.4694	-0.015850	6	-193.866	400.9
## 18	2.0590	0.3371				0.5043		4	-196.412	401.4
## 58	5.5980	0.3980			0.14100	0.3972	-0.032400	6	-194.113	401.4
## 28	10.6000	0.3393	-0.5312		0.06218	0.4249		6	-194.153	401.5
## 54	8.7750	0.4124		-0.06753		0.4831	-0.020710	6	-194.569	402.3
## 30	15.0200	0.3379		-0.19550	0.12100	0.3900		6	-194.574	402.3
## 60	9.6920	0.3826	-0.3335		0.11180	0.3974	-0.020840	7	-193.527	402.7
## 32	16.6200	0.3462	-0.3979	-0.12330	0.09771	0.3893		7	-193.583	402.8
## 56	13.2800	0.3897	-0.3805	-0.06232		0.4616	-0.009103	7	-193.784	403.2
## 26	2.0700	0.3189			0.06632	0.4607		5	-196.288	403.4
## 62	9.5590	0.3856		-0.07290	0.14300	0.3865	-0.024440	7	-194.001	403.6
## 64	13.2900	0.3714	-0.3289	-0.06731	0.11410	0.3874	-0.013650	8	-193.430	404.9
## 42	7.4660	0.4762			0.42120		-0.043120	5	-198.493	407.8
## 14	21.4300	0.3960		-0.28230	0.38890			5	-198.642	408.1
## 16	23.0300	0.4042	-0.4012	-0.20930	0.36490			6	-197.735	408.7
## 44	11.5500	0.4608	-0.3326		0.39220		-0.031600	6	-197.973	409.1
## 46	15.5400	0.4463		-0.15050	0.40980		-0.026090	6	-198.053	409.3
## 12	13.1700	0.4014	-0.6434		0.34470			5	-199.292	409.4
## 48	19.2200	0.4325	-0.3228	-0.14520	0.38200		-0.015510	7	-197.558	410.7
## 10	2.9280	0.3826			0.37920			4	-202.064	412.7
## 4	15.8000	0.6277	-0.7675					4	-204.471	417.5
## 8	23.5900	0.6402	-0.5846	-0.16290				5	-203.642	418.1
## 25	1.7770				0.35830	0.6036		4	-204.923	418.4
## 27	7.6410		-0.3661		0.36830	0.5852		5	-204.109	419.1
## 6	21.2400	0.6511		-0.26800				4	-205.331	419.2
## 36	15.3200	0.6623	-0.6555				-0.012050	5	-204.292	419.4
## 29	10.1000			-0.12580	0.40470	0.5636		5	-204.352	419.5

## 63	23.5200	-0.5416	-0.27070	0.26730	0.5295	0.045310	7	-202.096	419.8
## 61	17.8800		-0.29330	0.32620	0.5370	0.030900	6	-203.355	419.9
## 59	8.9530	-0.5902		0.27760	0.5924	0.021350	6	-203.449	420.1
## 40	26.0400	0.6147	-0.6296	-0.20570		0.010020	6	-203.573	420.3
## 57	1.4210			0.34310	0.6065	0.003343	5	-204.900	420.6
## 55	24.7500	-0.6974	-0.28320		0.7331	0.064000	6	-203.733	420.6
## 34	7.2360	0.7265				-0.033440	4	-206.112	420.8
## 31	11.1600	-0.2855	-0.07273	0.39290	0.5661		6	-203.955	421.1
## 51	9.5250	-0.7546			0.8072	0.039630	5	-205.154	421.1
## 38	19.3700	0.6714	-0.22600			-0.008267	5	-205.280	421.4
## 2	3.6540	0.6325					3	-207.960	422.2
## 53	17.3600		-0.31810		0.8051	0.049850	5	-205.846	422.5
## 17	1.5790				0.9765		3	-208.638	423.6
## 49	-0.5971				0.8959	0.020900	4	-207.554	423.7
## 19	6.6240	-0.3154			0.9696		4	-208.089	424.7
## 21	2.4020		-0.01249		0.9773		4	-208.633	425.8
## 23	4.1000	-0.3768	0.05320		0.9648		5	-208.005	426.8
## 47	34.5900	-0.5836	-0.43270	0.69370		0.056720	6	-208.474	430.1
## 45	28.6700		-0.45950	0.76370		0.041360	5	-209.717	430.3
## 13	18.8100		-0.24280	0.89990			4	-211.248	431.0
## 11	10.6500	-0.4874		0.86910			4	-211.939	432.4
## 9	2.8940			0.87680			3	-213.126	432.6
## 15	19.8100	-0.2609	-0.19480	0.89120			5	-210.970	432.8
## 43	11.7800	-0.6758		0.79840		0.017830	5	-211.563	434.0
## 41	3.2050			0.88800		-0.002960	4	-213.111	434.8
## 39	71.3700	-1.7370	-0.93490			0.211500	5	-227.610	466.1
## 37	63.1800		-1.22300			0.212600	4	-235.343	479.2
## 35	26.1600	-2.4350				0.170900	4	-237.409	483.4
## 33	-9.2640					0.152500	3	-250.046	506.4
## 7	-0.6592	-1.8330	0.65440				4	-276.779	562.1
## 3	31.9500	-1.1270					3	-279.011	564.3
## 1	14.2400						2	-280.162	564.5
## 5	-9.6720		0.35800				3	-279.391	565.1
##	delta weight								
## 20	0.00	0.181							
## 50	0.79	0.122							
## 22	1.43	0.088							
## 24	1.52	0.085							
## 52	1.54	0.084							
## 18	2.00	0.066							
## 58	2.04	0.065							
## 28	2.12	0.063							
## 54	2.95	0.041							
## 30	2.96	0.041							
## 60	3.28	0.035							
## 32	3.40	0.033							
## 56	3.80	0.027							
## 26	4.04	0.024							
## 62	4.23	0.022							
## 64	5.58	0.011							
## 42	8.45	0.003							
## 14	8.75	0.002							
## 16	9.28	0.002							
## 44	9.76	0.001							

```

## 46  9.92  0.001
## 12 10.05  0.001
## 48 11.35  0.001
## 10 13.31  0.000
## 4  18.12  0.000
## 8  18.75  0.000
## 25 19.02  0.000
## 27 19.68  0.000
## 6  19.84  0.000
## 36 20.05  0.000
## 29 20.17  0.000
## 63 20.42  0.000
## 61 20.52  0.000
## 59 20.71  0.000
## 40 20.96  0.000
## 57 21.26  0.000
## 55 21.28  0.000
## 34 21.40  0.000
## 31 21.72  0.000
## 51 21.77  0.000
## 38 22.02  0.000
## 2  22.87  0.000
## 53 23.15  0.000
## 17 24.23  0.000
## 49 24.29  0.000
## 19 25.36  0.000
## 21 26.44  0.000
## 23 27.47  0.000
## 47 30.76  0.000
## 45 30.90  0.000
## 13 31.67  0.000
## 11 33.05  0.000
## 9  33.21  0.000
## 15 33.40  0.000
## 43 34.59  0.000
## 41 35.40  0.000
## 39 66.68  0.000
## 37 79.86  0.000
## 35 84.00  0.000
## 33 107.05 0.000
## 7  162.74 0.000
## 3  164.98 0.000
## 1  165.11 0.000
## 5  165.74 0.000
## Models ranked by AICc(x)

```