Bio144, 6./7. April 2017

Solution to practical part 7: Exercise on linear algebra

1. The matrices and vectors can be generated in R:

a) > 2*A

b) > A + B

c) Calculating the results by hand:

$$\begin{bmatrix} 4 & 2 & 3 \\ 1 & 4 & 6 \end{bmatrix} \cdot \begin{bmatrix} 0 & -1 \\ 3 & -1 \\ 6 & 0 \end{bmatrix} = \begin{bmatrix} 4 \cdot 0 + 2 \cdot 3 + 3 \cdot 6 & 4 \cdot (-1) + 2 \cdot (-1) + 3 \cdot 0 \\ 1 \cdot 0 + 4 \cdot 3 + 6 \cdot 6 & 1 \cdot (-1) + 4 \cdot (-1) + 6 \cdot 0 \end{bmatrix} = \begin{bmatrix} 24 & -6 \\ 48 & -5 \end{bmatrix}$$

Checking with R:

$$> A%*%t(B)$$

d) Again, start by hand:

$$\begin{bmatrix} 4 & 2 & 3 \\ 1 & 4 & 6 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 2 \\ -3 \end{bmatrix} = \begin{bmatrix} 4 \cdot (-1) + 2 \cdot 2 + 3 \cdot (-3) \\ 1 \cdot (-1) + 4 \cdot 2 + 6 \cdot (-3) \end{bmatrix} = \begin{bmatrix} -9 \\ -11 \end{bmatrix}$$

1

And check by R:

e) Not defined, wrong dimensions.

- f) Not defined, wrong dimensions.
- g) > A %*% t(A)

```
[,1] [,2]
```

- [1,] 29 30
- [2,] 30
- h) > t(A) %*% A

53

- [1,] 17 12 18
- [2,] 12 20 30
- [3,] 18 30 45
- i) > t(x)%*%x

[1,] 14

j) > x % * % t(x)

- [1,] 1 -2 :
- [2,] -2 4 -6
- [3,] 3 -6 9

2.

a) In the lecture we have defined the covariate vectors $x^{(1)}$ and $x^{(2)}$, the data matrix \tilde{X} , the β vector, and the response vector:

$$> x1 \leftarrow c(0,1,2,3,4)$$

- $> x2 \leftarrow c(4,1,0,1,4)$
- > Xtilde <- matrix(c(rep(1,5),x1,x2),ncol=3)
- > t.beta <- c(10,5,-2)
- > t.y <- Xtilde%*%t.beta
- > t.e <- rnorm(5,0,1)
- > t.Y < -t.y + t.e
- > r.lm <- lm(t.Y ~x1 + x2)
- > summary(r.lm)\$coef

Estimate Std. Error t value Pr(>|t|)

(Intercept) 10.424761 1.1593521 8.991885 0.012143152

x1 4.996217 0.3895548 12.825454 0.006024432

x2 -2.494555 0.3292339 -7.576848 0.016976675

> solve(t(Xtilde) %*% Xtilde) %*% t(Xtilde) %*% t.Y

[,1]

- [1,] 10.424761
- [2,] 4.996217
- [3,] -2.494555

b) Generate a matrix of true value t.y and a matrix of residual errors t.E (instead of generating a separate t.e each time). The 100 observed vectors are then stored in a 5 × 100 matrix t.Y:

```
> t.E <- matrix(rnorm(500,0,1),nrow=5)
> t.Y <- t.y + t.E

100 iterations of the regression:
> r.coef <- matrix(NA,ncol=3, nrow=100)
> for (i in 1:100) {
+    r.coef[i,] <- lm(t.Y[,i] ~x1 + x2)$coefficients
+ }</pre>
```

> t.y <- matrix(rep(t.y,100),nrow=5,byrow=F)</pre>

- c) We procude the required graphs using ggplot. Load the libraries:
 - > library(ggplot2)
 - > library(tidyr)
 - > library(dplyr)

For the histograms, first convert r.coef into a data frame and rename the columns:

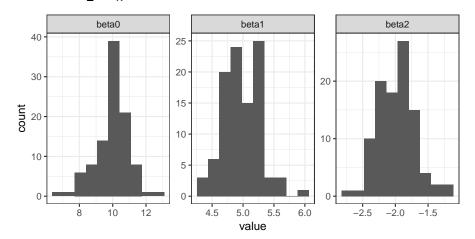
```
> r.coef <- data.frame(r.coef)
> names(r.coef) <- c("beta0","beta1","beta2")</pre>
```

Then either produce three separate plots using

```
> ggplot(r.coef,aes(x=beta0)) + geom_histogram()
> ggplot(r.coef,aes(x=beta1)) + geom_histogram()
> ggplot(r.coef,aes(x=beta2)) + geom_histogram()
```

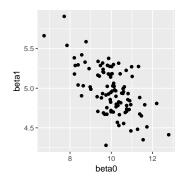
Or use the gather() function from the tidyr package, which has the advantage that you can then use facet_wrap():

```
> ggplot(gather(r.coef, key=variable, value=value), aes(value)) +
+ geom_histogram(bins=10) + facet_wrap(~variable, scales = "free") +
+ theme_bw()
```

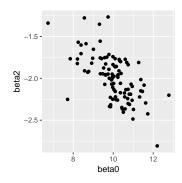


For the scatterplots we can continue to use r.coef:

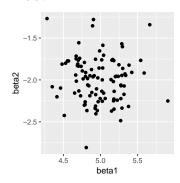
> ggplot(r.coef,aes(x=beta0,y=beta1)) + geom_point()



> ggplot(r.coef,aes(x=beta0,y=beta2)) + geom_point()



> ggplot(r.coef,aes(x=beta1,y=beta2)) + geom_point()



Observation: β_0 seems to be correlated with β_1 and β_2 .