Homework 5

Problem 14 b

```
sim.a <- function(n=100,a=-2,b=1,c=pi/4)
{
    Zt <- rnorm(n+2)
    a + b*Zt[3:(n+2)] + c*Zt[1:n]
}
sim.b <- function(n=100,a=-3,b=1,c=pi/4)
{
    Zt <- rnorm(n+2,sd = sqrt(0.25))
    a + b*Zt[3:(n+2)] + c*Zt[1:n]
}
set.seed(42)</pre>
```

Problem 14 b

```
df <- data.frame()</pre>
for (k in 1:1000){
 row_i <- as.vector(acf(sim.a(), lag.max=4, plot=FALSE)$acf[2:5])</pre>
  df <- rbind(df,row_i)</pre>
acf.tss.mat.1000 <- as.matrix(df)</pre>
colnames(acf.tss.mat.1000) <- c('rho_1','rho_2','rho_3', 'rho_4')</pre>
df_corr <- data.frame()</pre>
for (i in 1:4){
b <- c()
for (j in 1:4)
   b <- c(b, cor(acf.tss.mat.1000[,i],acf.tss.mat.1000[,j]))</pre>
df_corr <- rbind(df_corr,b)</pre>
df_cov <- data.frame()</pre>
for (i in 1:4){
b <- c()
for (j in 1:4)
```

```
b <- c(b, cov(acf.tss.mat.1000[,i],acf.tss.mat.1000[,j]))
}
df_cov <- rbind(df_cov,b)
}
colnames(df_corr) <- c('cor_1','cor_2','cor_3', 'cor_4')
colnames(df_cov) <- c('cov_1','cov_2','cov_3', 'cov_4')</pre>
```

Sample mean

```
colMeans(acf.tss.mat.1000)

## rho_1 rho_2 rho_3 rho_4

## -0.02600225 0.45860763 -0.02268161 -0.02475111
```

Correlation Matrix

```
knitr::kable(df_corr)
```

cor_1	cor_2	cor_3	cor_4
1.0000000	-0.0396955	0.6217270	-0.0538426
-0.0396955	1.0000000	-0.0525408	0.8285797
0.6217270	-0.0525408	1.0000000	-0.0680291
-0.0538426	0.8285797	-0.0680291	1.0000000

Covariance Matrix

```
knitr::kable(df_cov)
```

	cov_2	cov_3	cov_4
0.0233988	-0.0004535	0.0112017	-0.0009680
-0.0004535	0.0055768	-0.0004621	0.0072725
0.0112017	-0.0004621	0.0138732	-0.0009418
-0.0009680	0.0072725	-0.0009418	0.0138139

It is very much in sync with what large sample theory suggests, have done some spot checking below

```
Lets calculate W(4,4) ( Same as W(3,3) ) using the Bartlett matrix ((2*(pi^2))/(16*(1+pi/4)^2) + 1)/100 = 0.01387026
```

```
Also, lets calculate W(1,1) using the Bartlett matrix ((2*pi^2)/(16*(1+pi/4)^2) + 1 + (2*pi/(4*(1+(pi/4)^2))))/100 = 0.02358542
```

Also, lets calculate W(3,1) using the Bartlett matrix

```
(((pi^2))/(16 * (1+pi/4)^2) + ((2*pi/(4*(1+(pi/4)^2))))) /100 = 0.01165029
```

Once in negative, can be taken as tending towards 0, so $W(1,2),W(1,4),\,W(2,3)$ & W(3,4) are all zeros.

Changing the value, a = -2, sd = 4

```
#create matrix from vectors
set.seed(42)
df2 <- data.frame()</pre>
for (k in 1:1000){
  row_i <- as.vector(acf(sim.b(), lag.max=4, plot=FALSE)$acf[2:5])</pre>
  df2<- rbind(df2,row_i)</pre>
acf.tss.mat.1000.v2 <- as.matrix(df2)</pre>
colnames(acf.tss.mat.1000.v2) <- c('rho_1_2','rho_2_2','rho_3_2', 'rho_4_2')
df_corr_2 <- data.frame()</pre>
for (i in 1:4){
b < -c()
for (j in 1:4)
   b <- c(b, cor(acf.tss.mat.1000.v2[,i],acf.tss.mat.1000.v2[,j]))
df_corr_2 <- rbind(df_corr_2,b)</pre>
df cov 2 <- data.frame()</pre>
for (i in 1:4){
b <- c()
for (j in 1:4)
   b <- c(b, cov(acf.tss.mat.1000.v2[,i],acf.tss.mat.1000.v2[,j]))
df_cov_2 <- rbind(df_cov_2,b)</pre>
 colnames(df_corr_2) <- c('cor_1_2','cor_2_2','cor_3_2', 'cor_4_2')</pre>
 colnames(df_cov_2) <- c('cov_1_2','cov_2_2','cov_3_2', 'cov_4_2')</pre>
```

Sample mean 2

```
colMeans(acf.tss.mat.1000.v2)
```

```
## rho_1_2 rho_2_2 rho_3_2 rho_4_2
## -0.02600225 0.45860763 -0.02268161 -0.02475111
```

Covariance Matrix 2

knitr::kable(df_cov_2)

cov_1_2	cov_2_2	cov_3_2	cov_4_2
0.0233988 -0.0004535	-0.0004535 0.0055768	0.0112017 -0.0004621	-0.0009680 0.0072725
0.0112017 -0.0009680 ## Correlatio	-0.0004621 0.0072725 n Matrix 2	0.0138732 -0.0009418	-0.0009418 0.0138139

knitr::kable(df_corr_2)

cor_1_2	cor_2_2	cor_3_2	cor_4_2
1.0000000	-0.0396955	0.6217270	-0.0538426
-0.0396955	1.0000000	-0.0525408	0.8285797
0.6217270	-0.0525408	1.0000000	-0.0680291
-0.0538426	0.8285797	-0.0680291	1.0000000

Changing the variance (sqrt(0.25)) and (a'(-3)) does not seem to change the Mean, Correlation or Covariance matrix. I would say ecause of large sample, it would still be normally distributed.