### homework 2

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#### Problem 1

Since a strictly stationary depends only on the time lag k, and a Guassian stationary process has the multivariate normal distributions:  $f(y) = \frac{1}{2\pi^{n/2}|\Gamma|^{1/2}}e^{-1/2(y-\mu)'\Gamma^{-1}(y-\mu)}$ ,

it has the same distribution at t+k . So a Guassian stationary is a strictly stationary.

#### Problem 2

- (a)  $\mu_t = Y_t Z_t = \beta_0 + \beta_1 t$
- (b) No, since  $Y_t = \beta_0 + \beta_1 t + Z_t$ ,  $\beta_1 t$  is a factor of  $Y_t$ , which depend on the time t. So it is not stationary.
- (c)  $W_t = Y_t Y_{t-1} = [\beta_0 + \beta_1 t + Z_t] [\beta_0 + \beta_1 (t-1) + Z_{t-1}]$ =  $\beta_1 + Z_t - Z_{t-1}$ , which is not depend on t, so  $W_t$  is stationary.

#### Problem 3

- (a)  $Cov(Y_{t+k}, Y_t) = E[Y_{t+k}Y_t] E[Y_t]E[Y_t + k]$   $= E[Y_{t+k}Y_t] - \mu^2$   $= E[(\mu + \sum_{j=-\infty}^{\infty} \psi_j Z_{t-j})(\mu + \sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k})] - \mu^2$   $= E[\mu^2 + \mu(\sum_{j=-\infty}^{\infty} \psi_j Z_{t-j} + \sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k}) + (\sum_{j=-\infty}^{\infty} \psi_j Z_{t-j})(\sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k})] - \mu^2$   $= \mu^2 + \mu E[\sum_{j=-\infty}^{\infty} \psi_j Z_{t-j}] + \mu E[\sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k}] + E[(\sum_{j=-\infty}^{\infty} \psi_j Z_{t-j})(\sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k})] - \mu^2$   $= E[(\sum_{j=-\infty}^{\infty} \psi_j Z_{t-j})(\sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k})]$  since  $E[\sum_{j=-\infty}^{\infty} \psi_j Z_{t-j}] = 0$  and  $E[\sum_{j=-\infty}^{\infty} \psi_{j+k} Z_{t-j-k}] = 0$ . Since  $Z_t \sim WN(0, \sigma^2)$ ,  $Cov(Y_{t+k}, Y_t) = \sigma^2 \sum_{j=-\infty}^{\infty} \psi_j \psi_{j+k}$
- (b)  $\gamma(k) = Cov(Y_t, Y_{t+k}) = \sigma^2 \sum_{j=-\infty}^{\infty} \psi_j \psi_{j+k}$  Since  $\sum_{j=-\infty}^{\infty} |\psi_j| < \infty$  implys  $\sigma^2 \sum_{j=-\infty}^{\infty} \psi_j \psi_{j+k} < \infty$  and  $\sum_{k=-\infty}^{\infty} \sum_{j=-\infty}^{\infty} \psi_{j+k} < \infty$  So  $\sum_{k=-\infty}^{\infty} |\gamma(k)| = \sigma^2 \sum_{j=-\infty}^{\infty} \psi_j \psi_{j+k} * \sum_{k=-\infty}^{\infty} \sum_{j=-\infty}^{\infty} \psi_{j+k} < \infty * \infty < \infty$
- (c) Since  $\sum_{j=-\infty}^{\infty} \psi_j$  has the same variables with  $\sum_{k=-\infty}^{\infty} \sum_{j=-\infty}^{\infty} \psi_{j+k}$ , so  $\sum_{j=-\infty}^{\infty} \psi_j = \sum_{k=-\infty}^{\infty} \sum_{j=-\infty}^{\infty} \psi_{j+k}$ , then  $v = \sum_{k=-\infty}^{\infty} \gamma(k) = \sigma^2 \sum_{j=-\infty}^{\infty} \psi_j \sum_{j=-\infty}^{\infty} \psi_j$   $= \sigma^2 (\sum_{j=-\infty}^{\infty} \psi_j)^2$

#### Problem 4

```
(a) E(Y_t) = E(e_1 cos(t) + e_2 sin(t)) = E(e_1) cos(t) + E(e_2) sin(t)]

(b) Since e1, e2 have mean 0, then E(Y_t) = 0

Cov(Y_{t+k}, Y_t) = E(Y_{t+k}Y_t) - E(Y_{t+k})E(Y_t)

= E(Y_{t+k}Y_t)

= E[(e_1 cos(t+k) + e_2 sin(t+k))(e_1 cos(t) + e_2 sin(t))]

= E[e_1^2 cos(t+k) cos(t) + e_1 e_2 (cos(t+k) sin(t) + sin(t+k) cos(t)) + e_2^2 sin(t+k) sin(t)]

Since e_1, e_2 are independent, E[e_1 e_2] = E[e_1]E[e_2] = 0, and E[e_1^2] = E[e_2^2] = \sigma^2

then Cov(Y_{t+k}, Y_t) = \sigma^2 cos(t+k) cos(t) + \sigma^2 sin(t+k) sin(t)

= \sigma^2 (cos(t+k) cos(t) + sin(t+k) sin(t))

= \sigma^2 cos(t+k-t)

= \sigma^2 cos(t+k-t)
```

(c) Since  $E[Y_t] = 0$  is not depend on t, and  $Cov(Y_{t+k}, Y_t)$  depends only on lag k, So  $\{Y_t\}$  is stationary.

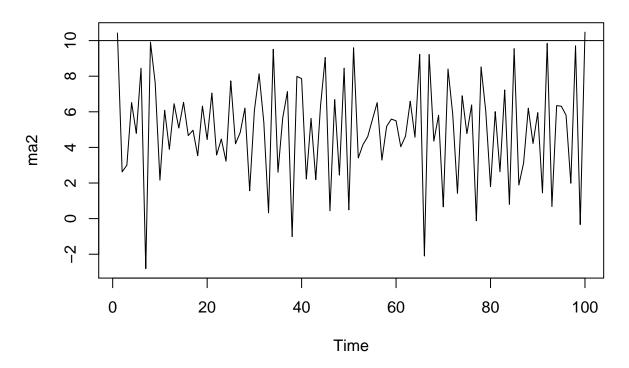
#### Problem 5

```
(a) E(Y_t)=\mu=5.0 \gamma(0)=\sigma_{MA}^2=(1-1.4+0.7)\sigma^2=0.3*2.4=0.72 Since MA(2), when k>2, \gamma(k)=0, \rho(k)=\gamma(k)/\gamma(0)=0 (b)
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```
(acf.ma2 = ARMAacf(ma=c(-1.4,0.7), lag.max=7))
```

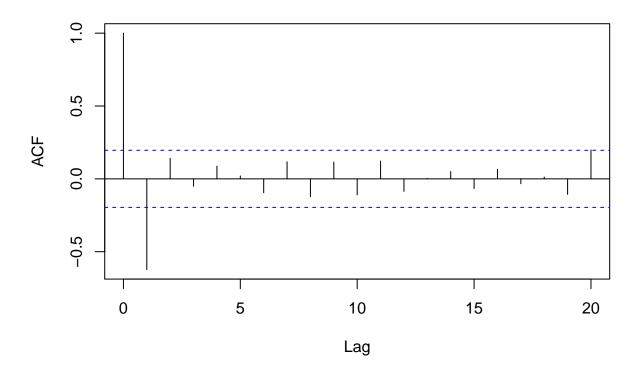
```
ma2 = 5.0+arima.sim(list(order=c(0,0,2), ma=c(-1.4,0.7)), n=100, sd=sqrt(2.4))
ts.plot(ma2, main="MA(2),mu=10,sd=2"); abline(h=10)
```

# MA(2),mu=10,sd=2



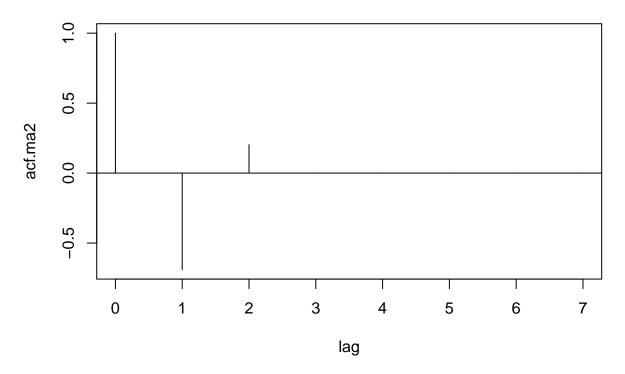
acf(ma2)

## Series ma2



plot(0:7, acf.ma2, type="h",xlab="lag",main="MA(2) theta1=1.4 theta2=-0.7")
abline(h=0)

## MA(2) theta1=1.4 theta2=-0.7



#### Problem 6

- (a) Since  $\lim_{k\to\infty}\rho(k)=0$ , then exist a k=n, where  $\gamma(n)=0$ , then from t=n to  $T=\infty$ ,  $\bar{Y}_n=(1/(T-n))\sum_{t=n}^T Y_t=\mu$  then plug t=1 to t=n-1 into  $\bar{Y}_n$  with T-n times, we can get  $\bar{Y}=(1/T)\sum_{t=1}^T Y_t\approx\mu$
- $\begin{array}{l} \text{(b) Since $\bar{Y}=(1/T)\sum_{t=1}^{T}Y_{t}$} \\ \text{then $Var(\bar{Y})=(1/T)^{2}\sum_{s=1}^{T}\sum_{t=1}^{T}Y_{t-s},$ where s is add as row sums.} \\ \text{Then $Var(\bar{Y})=(1/T)\sum_{k=-(T-1)}^{T-1}(1-\frac{|k|}{T})\gamma(k),$ where $k=t-s$.} \\ \text{Then $TVar(\bar{Y})=\sum_{k=-(T-1)}^{T-1}(1-\frac{|k|}{T})\gamma(k),$} \\ \text{when $T\!\!-\!\!>\!\!\infty,\,\frac{|k|}{T}\!\!-\!\!>\!\!0,$ and then $(1-\frac{|k|}{T})\!\!-\!\!>\!\!1$} \\ \text{then $TVar(\bar{Y})\!\!-\!\!>\!\!\sum_{k=-\infty}^{\infty}\gamma(k)$} \end{array}$