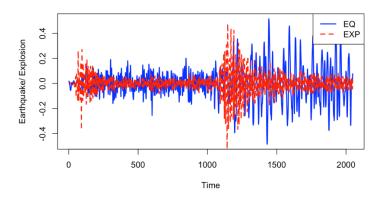
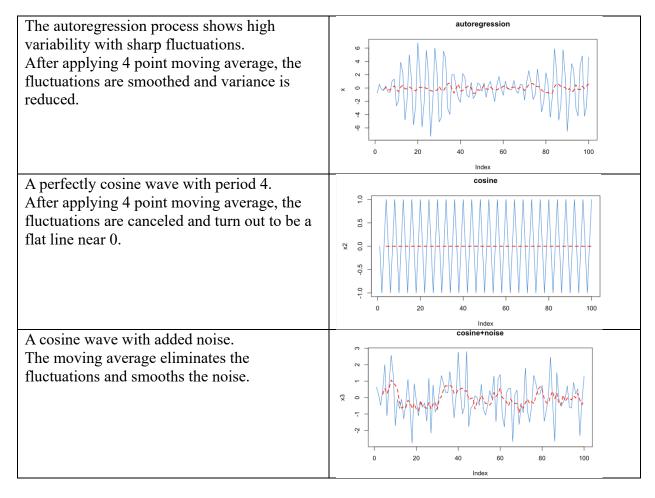
1.1



1.3



a) It is not stationary because the mean is not constant. $E(x)=\beta_1+\beta_2t$, this formula depends on t, which means it is not stationary unless $\beta_2=0$.

b)
$$y_t = x_t - x_{t-1} = \beta_1 + \beta_2 t + w_{t-1} (\beta_1 + \beta_2 (t-1) + w_{t-1}) = \beta_2 + (w_t - w_{t-1})$$

 $E(y_t) = \beta_2$, which is a constant.

$$\begin{split} \gamma_y(0) = &Var\left(w_{t-1}\right) = Var(w_t) + Var(w_{t-1}) - 2Cov(w_t, w_{t-1}) = 2\sigma_w^2, \text{ where xuw}_t \text{ is white noise} \\ \gamma_y\left(1\right) = &\gamma_y\left(-1\right) = Cov(w_t - w_{t-1}, w_{t-1} - w_{t-2}) = Cov(w_t, w_{t-1}) - Cov(w_t, w_{t-2}) - Cov(w_{t-1}, w_{t-1}) + Cov(w_{t-1}, w_{t-2}) = -\sigma_w^2, \end{split}$$

COV (A+B, (+D) = GV (A,C) + GV (A,D) + GV (B,C) + GV (B,D)

$$\gamma_v$$
 (h)=0 for |h|>1,

All moments are time-invariant, which is stationary.

c)

$$\begin{split} E\left(V_{4}\right) &= \frac{1}{2q+1} \sum_{j=-q}^{q} \left(\beta_{i} + \beta_{2}(t-j)\right) = \frac{1}{2q+1} \left[\sum_{j=-q}^{q} \beta_{i} + \sum_{j=-q}^{q} \beta_{i} + \sum_{j=-q}^{q} \beta_{j} \right] \\ &= \frac{1}{2q+1} \left[\left(2q+1\right)\beta_{i} + \left(2q+1\right)\beta_{i} + \left(2q+1\right)\beta_{i$$

$$y_{v}(h) = Cov[(\beta_{1}+\beta_{2}+b) + u_{4}, \beta_{1}+\beta_{2}(t-h) + U_{t-h}]$$

$$= Cov[(U_{t}, U_{t-h})] \qquad U_{t} = \frac{1}{2q+1} \sum_{j=-q}^{q} W_{t-j} \qquad U_{t-h} = \frac{1}{2q+1} \sum_{k=-q}^{q} W_{t-h-k}$$

$$= \frac{1}{2q+1} \sum_{j=-q}^{q} \sum_{k=-q}^{q} Cov(W_{t-j}, W_{t-h-k})$$
when $t-j=t-h-k \Rightarrow j=h+k$

1.8.

(a)
$$X_{t} = \delta + X_{t+1} + W_{t}$$
, with $X_{0} = 0$
 $X_{1} = \delta + X_{0} + W_{1} = \delta + W_{1}$
 $X_{2} = \delta + X_{1} + W_{2} = \delta + \delta + W_{1} + W_{2} = 2 \delta + W_{1} + W_{2}$
 $X_{3} = \delta + X_{2} + W_{3} = 3 \delta + W_{1} + W_{2} + W_{3}$
 $X_{4} = \delta + \sum_{k=1}^{4} W_{k}$

(b) $X_{5} = X_{5} + W_{5} + W_{5}$

 $X_{t} = \begin{cases} t + \xi_{ij} W_{k} \\ W_{i} \end{cases} = \begin{cases} t + \xi_{ij} W_{k} \\ W_{k} \end{cases} = \begin{cases} t + \xi_{ij} W_{k} \\$

- c) mean=function depends on time, t
- d) $ex(t-1,t)=\frac{(t-1)O_w^2}{\sqrt{(t+1)O_w^2}}=\sqrt{\frac{t-1}{t}}$ when $t\to\infty$, the result $\to 1$

The adjacent levels are nearly perfectly correlated for large t

The process is nonstationary and level-based correlations can be misleading.

(mean: ft, variance = tow)

e) $y_{t=\chi_t-\chi_{t-1}} = S + W_t$ $f(y_t) = S$, $Var(y_t) = Uu^{\dagger}$, which are both constant $GV(y_t, y_{t+1})$ $GV(S + W_t, S + W_{t+1}) = 0$ when $h \neq 0$

1.9

1.9. $X_t = U_1$ Gin (rrubt) $+ U_2$ as (rrubt) , let rrubt = a , rrw. (t-h) = bwith U_1 , U_2 are independent random variables with 0 mean and $E(U^*) = E(U^*) = \sigma^2 = 1$ $E(U_1) = E(U_2) = 0$ $E(X_t) = E(U_1) \cdot \sin(2\pi W_1 t) + E(U_2) \cdot \cos(2\pi W_1 t) = 0$ =7 mean is constant 0

YCh) = COU (Xt, Xth) = [(Xt·Xt-h) - [(Xt)·E(Xth))

 $= \overline{b}(X_t \cdot X_{t-h})$

= [(U, Gina+D2 65 a) (U, Ginb+O2 64b)

= El Uizsina sinb+ Ui Uz sina sob+ Ui Uzsinbusa+ Uz asaersb)

= E(U125,nasinb+0205aasb)

= E(Ui) Sinasinit E(Ui) usaasb

= 02 (Gina unbtasa asb)

= 02 05 (a-6)

= 02 as (270 Woh)

Therefore this series is nearly stationary with out avaiance function 4(h)= 0203(12200h)

(A)
$$MSE(A) = E[(X_{E+1} - AX_{E})^2]$$
, X_E with Zero mean
$$= E[X_{E+1}] + A^2 E[X_{E}^2] - 2A E[X_{E+1}, X_{E}]$$

$$= Y_{(D)} + A^2 (Y_0) - 2A Y(U)$$

$$\frac{d}{dA} MSE(A) = 2AY(0) - 2Y(U) = 0$$

$$\therefore A = \frac{Y(U)}{Y_{(D)}} = C(U)$$

$$\therefore A^2 = 0$$

$$\therefore C(U) \text{ is the nimmum value.}$$

b) MFF NOTE Y(0)+
$$Q(l)^{2}Y(0) - 2Q(l)Y(l)$$

$$= Y(0) [1 + Q(l)^{2} - 2Q(l)Y(l)] => Q(l)$$

$$= Y(0) [1 + Q(l)^{2} - 2Q(l)^{2}]$$

$$= Y(0) [1 + Q(l)^{2}]$$

c) if
$$X_{t+l} = AX+b$$

$$\int_{GU(X, X) = lor(X)} for zero naew, unothy sharmany concess$$
Then. $GU(X_{t+l}, X_t) = GU(AX_t, X_t) = A Go (X_t, X_t) = A Var(X_t) = A V(0)$

$$Var(X_{t+l}) = Var(AX_t) = E[AX_t - E(AX_t)^2] = E[(A(X_t - E(X_t))^2)] = A^2[E(X_t - E(X_t)^2)] = A^$$

$$= \frac{G_0V(AX_{+}, X_{+})}{\int V_{Ar}(X_{+}) \cdot V_{Ar}(X_{+})} = \frac{AY(5)}{\int A^2 V_{Ar}(X_{+}) \cdot V_{Ar}(X_{+})}$$

$$= \frac{AY(5)}{|A V_{Ar}(X_{+})|}$$

1.16

1.16. (a)
$$E(\chi_{+}) = \int_{0}^{1} \zeta_{11}(2\lambda n t) du = \left[\frac{OS(2\lambda l t)}{2\lambda t} \right]_{0}^{1} = \frac{COS(2\lambda l t)}{2\lambda t} + 1$$

$$= (-1 - 0)$$

$$Y(L) = E(X_{+} \cdot X_{+} \cdot L) = \int_{0}^{1} \zeta_{11}(2\lambda n t) \zeta_{11}(2\lambda n t) + 2\lambda t (n + 2\lambda l t) + 2\lambda t$$