

Hw 1 R Code

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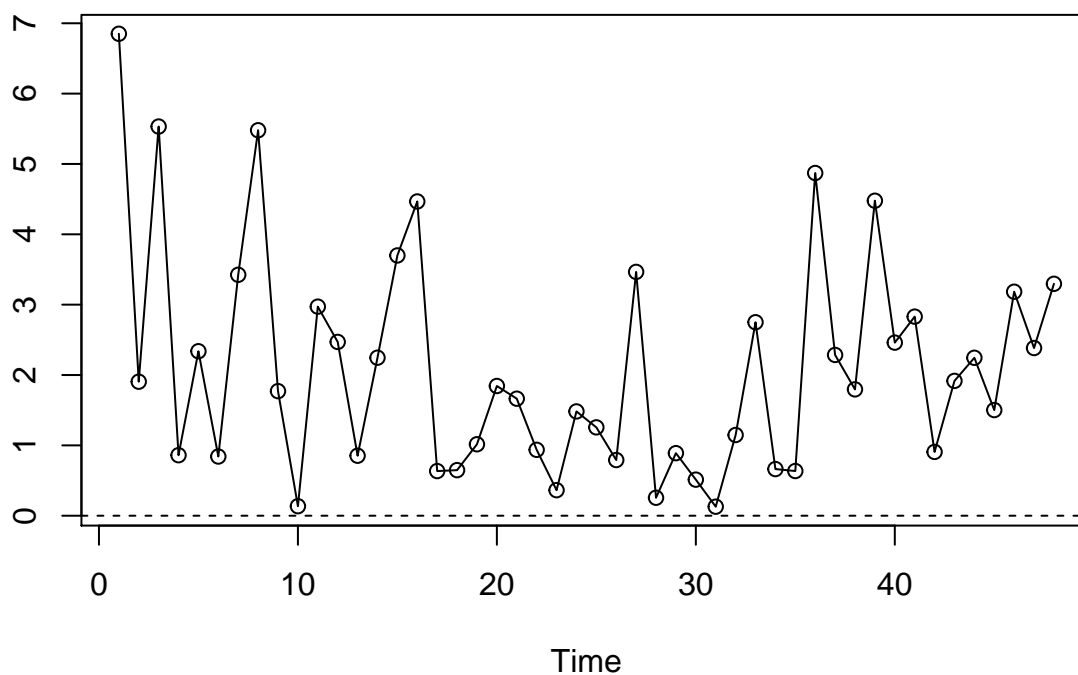
1/31/2021

1.4

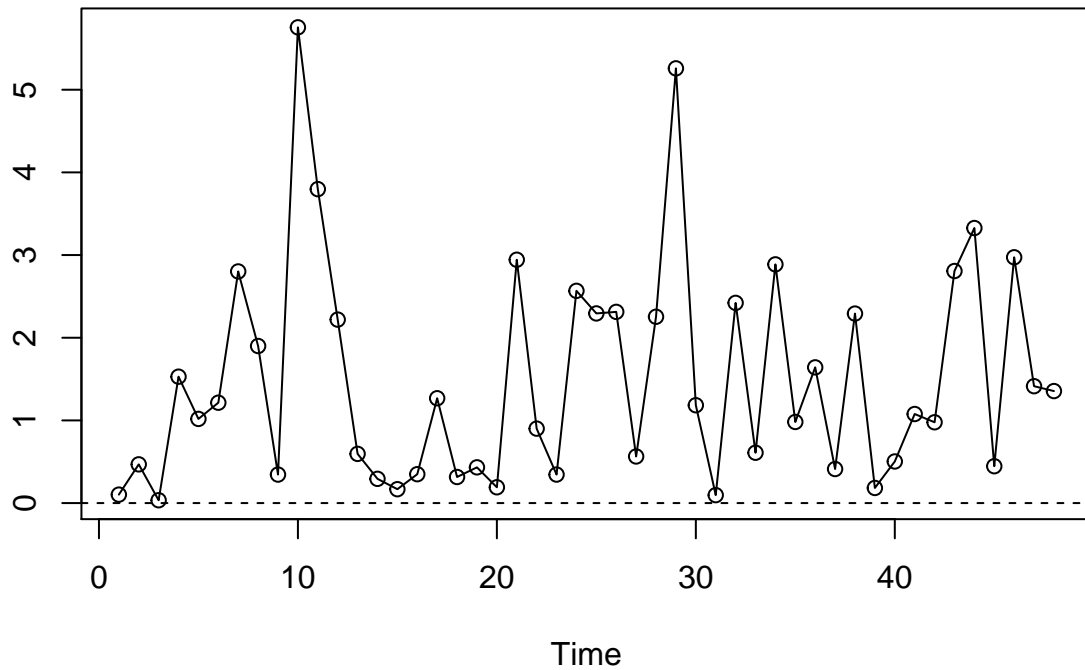
Below are two plots of different simulations of the chi-squared distribution in following code block. Both appear to follow non-normal distributions but they are not random - they follow a chi-squared distribution.

```
set.seed(7)
chiSquare <- ts(rchisq(n = 48, df = 2))
plot(chiSquare, type = "o", ylab = "", main = "Chi-Squared Distribution, n = 48, df = 2")
abline(h = 0, lty = 2)
```

Chi-Squared Distribution, n = 48, df = 2



Chi-Squared Distribution, $n = 48$, $df = 2$

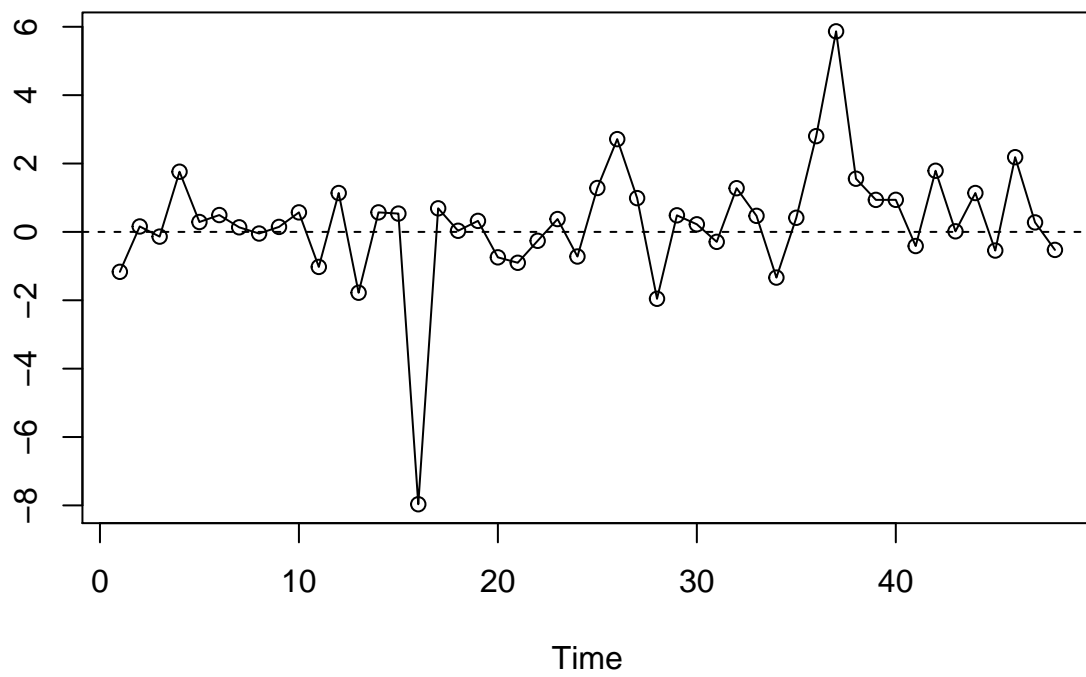


1.5

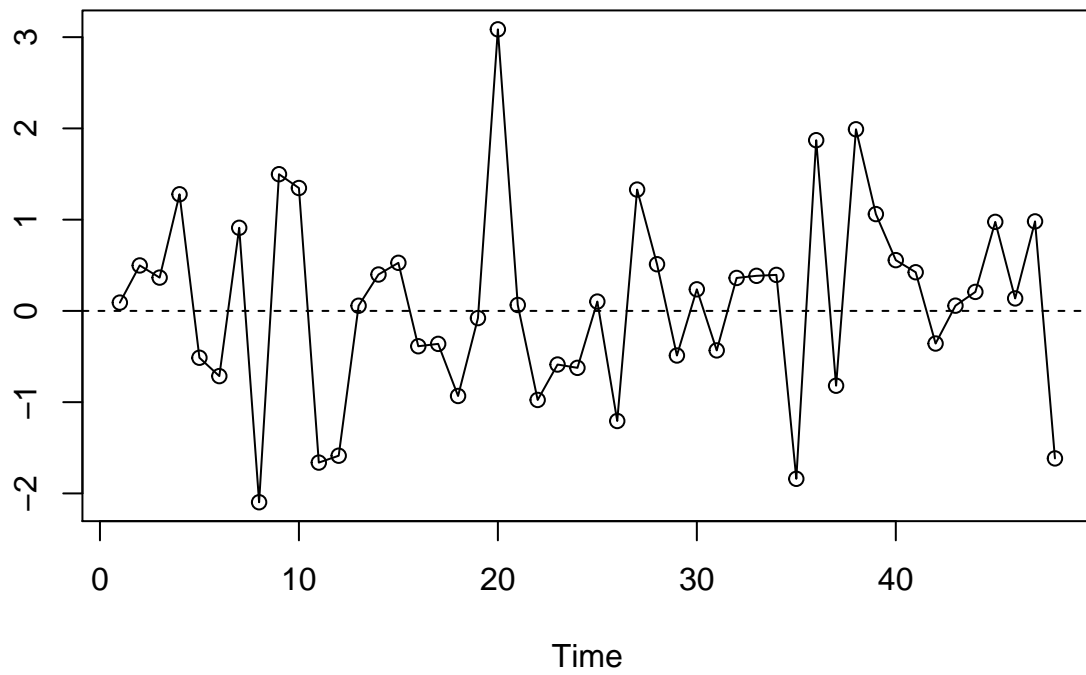
Below are two plots of different simulations of the t distribution in following code block. Both appear to follow normal and do not appear to be random.

```
set.seed(17)
chiSquare <- ts(rt(n = 48, df = 5))
plot(chiSquare, type = "o", ylab = "", main = "t Distribution, n = 48, df = 5")
abline(h = 0, lty = 2)
```

t Distribution, $n = 48$, $df = 5$

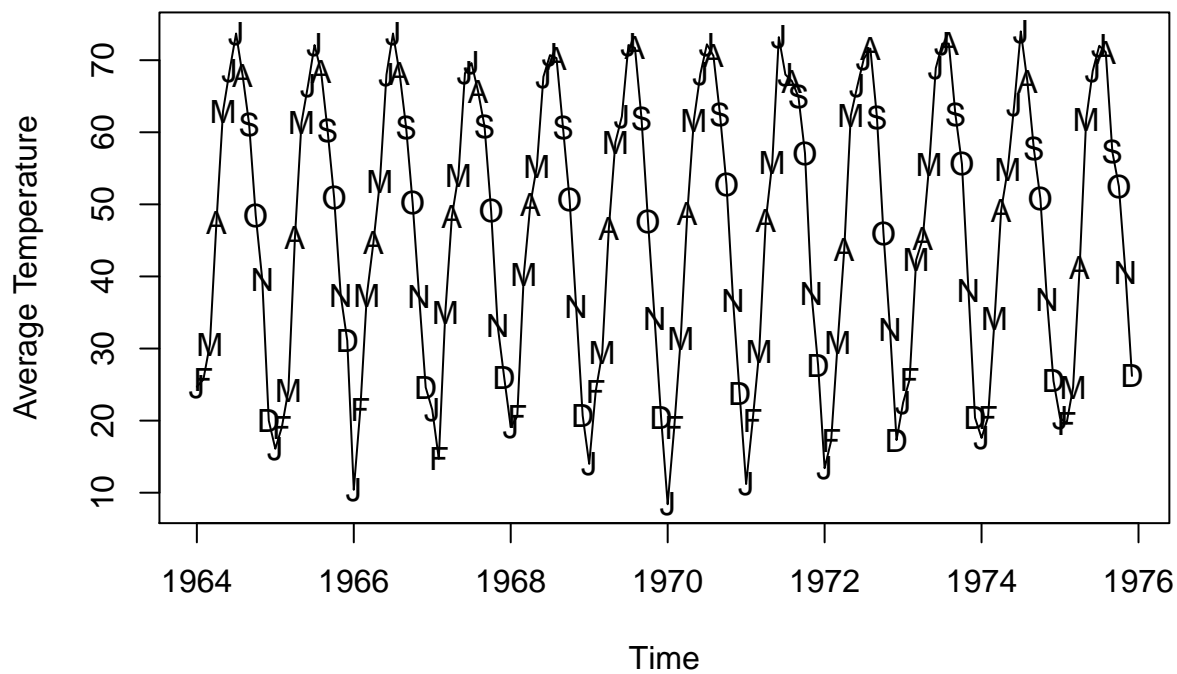


t Distribution, n = 48, df = 5



1.6

```
data("tempdub")
plot(tempdub, type = "l", ylab = "Average Temperature")
points(y = tempdub, x = time(tempdub), pch = as.vector(season(tempdub)))
```



2.9 a) $Y_t = \beta_0 + \beta_1 t + X_t$
 $E[Y_t] = \beta_0 + \beta_1 t + E[X_t] = \beta_0 + \beta_1 t + \mu_{t,x}$
 \rightarrow DEPENDS ON t , \therefore NOT STATIONARY

$$W_t = \nabla Y_t = Y_t - Y_{t-1}$$

$$E[W_t] = E[Y_t - Y_{t-1}] = E[\beta_0 + \beta_1 t + X_t - (\beta_0 + \beta_1(t-1) + X_{t-1})]$$

$$= \beta_0 + \beta_1 t + E[X_t] - \beta_0 - \beta_1(t-1) - E[X_{t-1}]$$

$$= \beta_0 + \beta_1 t - \beta_0 - \beta_1 t + \beta_1 = \beta_1 \rightarrow \text{NOT DEPENDENT ON } t$$

$$\text{cov}(W_t) = \text{cov}(Y_t, Y_{t-1}) = \text{cov}(\beta_0 + \beta_1 t + X_t, \beta_0 + \beta_1(t-1) + X_{t-1})$$

$$= \text{cov}(X_t, X_{t-1}) = \gamma_h \rightarrow \text{NOT DEPENDENT ON } t$$

$\therefore W_t$ IS STATIONARY

2.10 a) $Y_t = \mu_t + \sigma_t X_t$
 $E[Y_t] = E[\mu_t + \sigma_t X_t] = \mu_t + \sigma_t E[X_t] = \mu_t + \sigma_t(0) = \mu_t$
 $\text{cov}(Y_t, Y_{t-h}) = \text{cov}(\mu_t + \sigma_t X_t, \mu_{t-h} + \sigma_{t-h} X_{t-h}) = \sigma_t \sigma_{t-h} \text{cov}(X_t, X_{t-h})$
 $= \sigma_t \sigma_{t-h} \rho_h$

b) $\text{Var}[Y_t] = \text{Var}[\mu_t + \sigma_t X_t] = \sigma_t^2 \text{Var}[X_t] = \sigma_t^2(1) = \sigma_t^2$

$$\text{corr}[Y_t, Y_{t-h}] = \frac{\sigma_t \sigma_{t-h} \rho_h}{\sqrt{\text{Var}[Y_t] \text{Var}[Y_{t-h}]}} = \frac{\sigma_t \sigma_{t-h} \rho_h}{\sigma_t \sigma_{t-h}} = \rho_h$$

Y_t COULD BE STATIONARY AS LONG AS μ_t DOESN'T DEPEND ON t , BUT WE DON'T HAVE THAT INFORMATION

c) ρ_h IS FREE OF t BUT AGAIN, WE DON'T KNOW IF σ_t DEPENDS ON t .

2.13 a) $Y_t = e_t - \theta(e_{t-1})^2$
 $E[Y_t] = E[e_t - \theta(e_{t-1})^2] = E[e_t] - \theta E[e_{t-1}^2] = \theta \text{Var}[e_{t-1}] = -\theta \sigma_e^2$
 $\text{Var}[Y_t] = \text{Var}[e_t - \theta(e_{t-1})^2] = \text{Var}[e_t] + \theta^2 \text{Var}[e_{t-1}^2]$
 $= \sigma_e^2 + \theta^2 (E[e_{t-1}^4] - E[e_{t-1}^2]^2) = \sigma_e^2 + \theta^2 (3\sigma_e^4 - \sigma_e^4)$
 $= \sigma_e^2 + 2\theta^2 \sigma_e^4$

$$\begin{aligned}
\text{cov}(Y_t, Y_{t-1}) &= \text{cov}(e_t - \theta e_{t-1}^2, e_{t-1} - \theta e_{t-2}^2) \\
&= \text{cov}(e_t, e_{t-1}) - \theta \text{cov}(e_t, e_{t-2}^2) - \theta \text{cov}(e_{t-1}^2, e_{t-1}) \\
&\quad + \theta^2 \text{cov}(e_{t-1}^2, e_{t-2}^2) \\
&= \theta \text{cov}(e_{t-1}^2, e_{t-1}) = -\theta(E[e_{t-1}^3] + \mu_{t-1} + \mu_t) = 0
\end{aligned}$$

$$Y_{t,s} = \frac{\text{cov}(Y_t, Y_{t-1})}{\sqrt{\text{Var}[Y_t] \text{Var}[Y_{t-1}]}} = 0$$

b) AUTOCORRELATION IS 0 AND $E[Y_t]$ IS CONSTANT
 $\therefore Y_t$ IS STATIONARY

2.14 a) $Y_t = \theta_0 + t e_t$

$$E[Y_t] = E[\theta_0 + t e_t] = \theta_0 + t E[e_t] = \theta_0 + t(0) = \theta_0$$

\rightarrow NOT DEPENDENT ON t

$$\text{Var}[Y_t] = \text{Var}[\theta_0 + t e_t] = 0 + t^2 \text{Var}[e_t] = t^2 \sigma_e^2$$

\rightarrow DEPENDENT ON t

Y_t IS NOT STATIONARY

b) $W_t = \nabla Y_t = Y_t - Y_{t-1} = t e_t - (t-1) e_{t-1}$

$$\begin{aligned}
E[W_t] &= E[\theta_0 + t e_t - \theta_0 - (t-1) e_{t-1}] = t E[e_t] - (t-1) E[e_{t-1}] \\
&= t E[e_t] - t E[e_{t-1}] + E[e_{t-1}] = 0
\end{aligned}$$

\rightarrow NOT DEPENDENT ON t

$$\begin{aligned}
\text{Var}[W_t] &= \text{Var}[t e_t] - \text{Var}[(t-1) e_{t-1}] = t^2 \text{Var}[e_t] - (t-1)^2 \text{Var}[e_{t-1}] \\
&= t^2 \sigma_e^2 - (t-1)^2 \sigma_e^2 = (2t-1) \sigma_e^2
\end{aligned}$$

\rightarrow DEPENDENT ON t

W_t IS NOT STATIONARY

c) $Y_t = e_t e_{t-1}$

$$E[Y_t] = E[e_t e_{t-1}] = E[e_t] E[e_{t-1}] = 0$$

\rightarrow NOT DEPENDENT ON t

$$\begin{aligned}
\text{cov}(Y_t, Y_{t-1}) &= \text{cov}(e_t e_{t-1}, e_{t-1} e_{t-2}) = E[(e_t e_{t-1} - \mu_t^2)(e_{t-1} e_{t-2} - \mu_{t-1}^2)] \\
&= E[e_t] E[e_{t-1}] E[e_{t-1}] E[e_{t-2}] = 0
\end{aligned}$$

\rightarrow NOT DEPENDENT ON t

Y_t IS STATIONARY

$$2.19 \quad a) \quad Y_1 = \theta_0 + e_1$$

$$t > 1 \quad Y_t = \theta_0 + Y_{t-1} + e_t$$

$$Y_1 = \theta_0 + e_1$$

$$Y_2 = \theta_0 + \theta_0 + e_2 + e_1$$

$$Y_t = \theta_0 + \theta_0 + \dots + e_t + e_{t-1} + \dots + e_1$$

$$Y_t = t\theta_0 + e_t + e_{t-1} + \dots + e_1$$

$$b) \quad E[Y_t] = E[t\theta_0 + e_t + e_{t-1} + \dots + e_1] = t\theta_0 + E[e_t] + E[e_{t-1}] + \dots + E[e_1] \\ = t\theta_0 + 0 + 0 + \dots + 0 = t\theta_0$$

$$c) \quad \text{cov}(Y_t, Y_{t-h}) = \text{cov}(t\theta_0 + e_t + e_{t-1} + \dots, (t-h)\theta_0 + e_{t-h} + \dots) \\ = \text{cov}(e_t + e_{t-1} + \dots, e_{t-h} + e_{t-h-1} + \dots) \\ = \text{Var}[e_{t-h} + e_{t-h-1} + \dots] = (t-h)\sigma_e^2$$