

id no 17

Assignment - 10 ECON

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Q.9.2

a)

As expected, advertising has ~~the~~ positive impact on sales. The effect is on current week as well as next two weeks or we can say for this week, there is effect from last two weeks. The maximum impact is in next week after advertising.

total effect =

$$1.842 + 3.802 + 2.265 = 7.909$$

sustained one million increase in adv. will increase sales by 7.909 million keeping all else constant.

b)

$$t_c = \frac{5\%}{1}$$

$$t_c = \frac{10\%}{1}$$

One tail 5%	One tail 10%	two tail 5%	two tail 10%
1.66 1.66	1.289	1.98	1.66

$$H_0: \beta_t = 0 \quad \frac{1.842}{\sqrt{1.3916}} = 1.559$$

$$H_1: \beta_t \neq 0$$

Significant at one tail 10%.

$$H_0: \beta_{t-1} = 0 \quad t = \frac{3.802}{\sqrt{2.1606}} = 2.586$$

$$H_1: \beta_{t-1} \neq 0.$$

Significant at 5 & 10% ~~both~~
one & two tail.

$$H_0: \beta_{t-2} = 0 \quad t = \frac{2.265}{\sqrt{1.4214}} = 1.899$$

$$H_1: \beta_{t-2} \neq 0$$

significant at one & two tail 5 & 10%.

$$c) \quad t_c = (0.975, 101) = 1.984$$

impact

$$\text{one period} = 1.842 \pm 1.984 \times \sqrt{1.3946}$$

$$1.842 \pm 2.342$$

$$= -0.5009, 4.184$$

one period

$$\sqrt{\text{var}(b_0) + \text{var}(b_1) + 2 \text{cov}(b_0, b_1)}$$

$$= \sqrt{1.3946 + 2.1606 + 2 \times -1.0406}$$

$$= 1.2140$$

95% conf int

$$= b_0 + b_1 \pm 1.984 \times 1.2140$$

$$= (3.2354, 8.052)$$

total impact

$$\sqrt{\text{var } b_0 + \text{var } b_1 + \text{var } b_2 + 2 \text{cov } b_0 b_1 + 2 \text{cov } b_0 b_2 + \text{cov } b_1 b_2}$$

$$= \sqrt{1.3946 + 2.1606 + 1.4214 + 2 \times -1.0406 + 2 \times 0.0984 + 2 \times -1.0367}$$

$$= 1.009$$

95% conf interval

$$= 7.909 \pm 1.984 \times 1.009$$

$$= (5.907, 9.91)$$

Q.9.4a)

~~Q.9.4~~

$$\beta_1 = \frac{0.0979}{1.5436} = 0.0634$$

$$\beta_2 = \frac{0.1008}{1.5436} = 0.0653$$

$$b) \quad Z_{\text{critical}} = -1.959 \quad +1.959$$

$$Z_{S_1} = \sqrt{10} \times 0.0634 = 0.2004$$

$$\checkmark H_0: S_1 = 0$$

$$H_1: S_1 \neq 0$$

We fail to reject null hyp.

$\therefore S_1$ is not significantly diff from zero

$$Z_{S_2} = \sqrt{10} \times 0.0653 = 0.2064$$

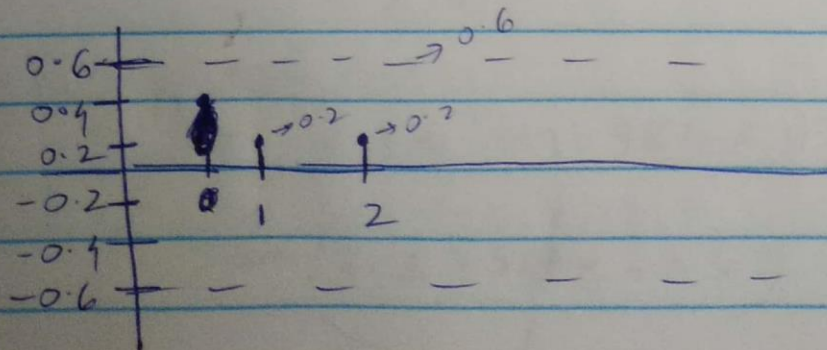
$$\checkmark H_0: S_2 = 0$$

$$H_1: S_2 \neq 0$$

We fail to reject the null hypothesis

$\therefore S_2$ is not significantly diff from zero

$$c) \quad \sqrt{10} \times x = \pm 1.959 \Rightarrow x = \pm 0.61$$



Q.9.6

a) An increase in interest rate by 1 unit (interest rate in percentage) ~~or change of~~ as compared to that of in previous month, will decrease the sale of houses by 53510 units ~~to~~ as compared to previous month

$$\begin{aligned} 951 \text{ conf} &= -53.51 \pm 1.971 \times 16.98 \\ &= -86.9774 - 20.042 \end{aligned}$$

\therefore the number of household sold will decrease by 86977 \pm 20042 as compared to last month.

b) $H_0: \beta_{t-1} = 0$

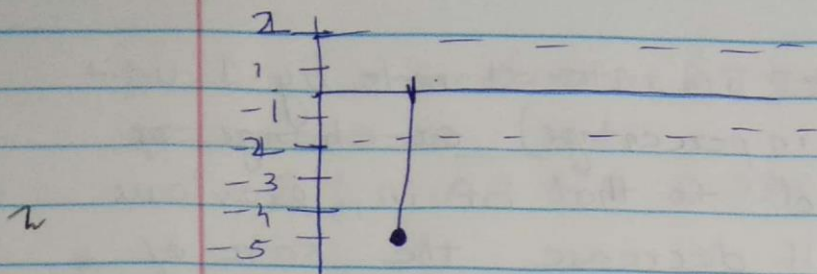
$\checkmark H_1: \beta_{t-1} \neq 0$

$$t_{\text{critical}} = t(0.975, 215) = \pm 1.97$$

$$t_{\text{stat}} = \frac{-0.3306}{0.0649} = -5.09$$

$$t_{\text{stat}} > t_{\text{critical}}$$

\therefore we reject the null hypothesis & conclude that there is autocorrelation



MY

$$\begin{aligned}
 & \text{c) } 95\% \text{ confint} \\
 & = -58.61 \pm 1.971 \times 14.10 \\
 & = \Rightarrow \\
 & = -86.40 \text{ \& } -30.8189
 \end{aligned}$$

The lower bound of confidence interval is approx the same but the upper bound changed significantly. This shows that ignoring the autocorrelation gives incorrect std. errors & lead to incorrect confidence interval.

11/11/22

Q.9-22

a) Plotted using R

b)
$$\text{congwth} = 0.973 + 0.449 \times \text{incgwth}$$

So says 1% increase in income growth rate will increase consumption growth rate by ~~11~~ ~~45~~ 0.45%.

$$AIC = \ln\left(\frac{SSE}{N}\right) + \frac{2K}{N} = -0.856$$

$$BIC = \ln\left(\frac{SSE}{N}\right) + \frac{K}{N} (\ln(N)) = -0.823$$

The correlogram suggests that - there is significant correlation in errors at lag 2.

c) Adding lag for congwth

$$\text{congwth} = 0.671 + 0.271 \text{ Congwth}(t-1) + 0.350 \text{ Incgwth}_t$$

all significant at 5%.

$$\begin{aligned} AIC &= -0.81 \\ BIC &= -0.76 \end{aligned} \left. \vphantom{\begin{aligned} AIC &= -0.81 \\ BIC &= -0.76 \end{aligned}} \right\} \begin{array}{l} \text{both values} \\ \text{reduced compared} \\ \text{to model 1} \end{array}$$

which says this is a better model

correlogram says high correlation in lag 1 & 2.

d) Adding cong with $t-2$

$$\begin{aligned} \text{congwith} &= 0.424 + 0.159 \text{congwith}(t-1) \\ &\quad + 0.2805 \text{congwith}(t-2) \\ &\quad + 0.221 \text{incgwith} \end{aligned}$$

θ_2 is significant at 5%.

$$AIC = -0.910$$

$$BIC = -0.84$$

AIC & BIC further reduced suggesting model 3 is better than 1

f) Adding congwth_{t-3} & incgwth_{t-2}

$$\begin{aligned}\text{congwth} = & 0.322 + 0.012 \text{congwth}_{(t-1)} \\ & + 0.208 \text{congwth}_{(t-2)} + 0.04 \text{congwth}_{t-3} \\ & + 0.341 \text{incgwth}_t + 0.22 \text{incgwth}_{t-1} \\ & - 0.01 \text{incgwth}_{t-2}\end{aligned}$$

The variables congwth_{t-3} & incgwth_{t-2} are not significant

correlogram also same as previous

g) dropping congwth_{t-1}

$$\begin{aligned}\text{congwth} = & 0.340 + 0.214 \text{congwth}_{t-2} \\ & + 0.3554 \text{incgwth}_t + 0.241 \text{incgwth}_t\end{aligned}$$

all significant at 5%

$$AIC = 1.003 \quad \left. \begin{array}{l} \text{lowest AIC} \\ \text{4 BIC} \end{array} \right\}$$

$$BIC = -0.93$$

The model is better than model e

Correleto

correlogram suggests that there is correlation at lag 1.

e) Adding incgwth_{t-1}

$$\begin{aligned}\text{congwth} = & 0.3319 + 0.023 \text{ congwth}_{t-1} \\ & + 0.2101 \text{ congwth}_{t-2} \\ & + 0.349 \text{ incgwth} + 0.233 \text{ incgwth}_{t-1}\end{aligned}$$

all significant except congwth_{t-1} at 5%.

$\text{congwth}_{t-1} \rightarrow d_1$ ~~not~~
 d_1 significant.

$$\text{AIC} \rightarrow -0.993$$

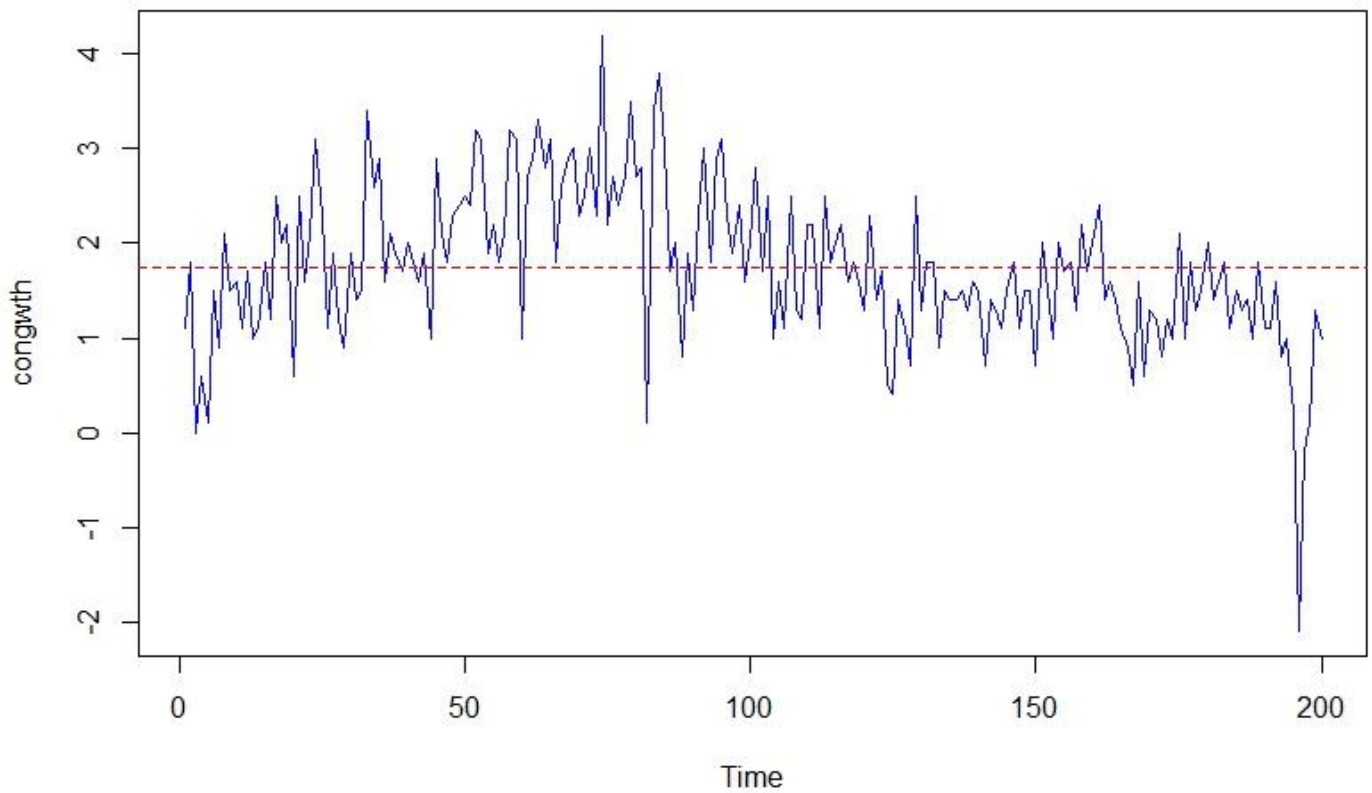
$$\text{BIC} \Rightarrow \text{~~-0.817~~} -0.91 \quad \text{AIC \& BIC}$$

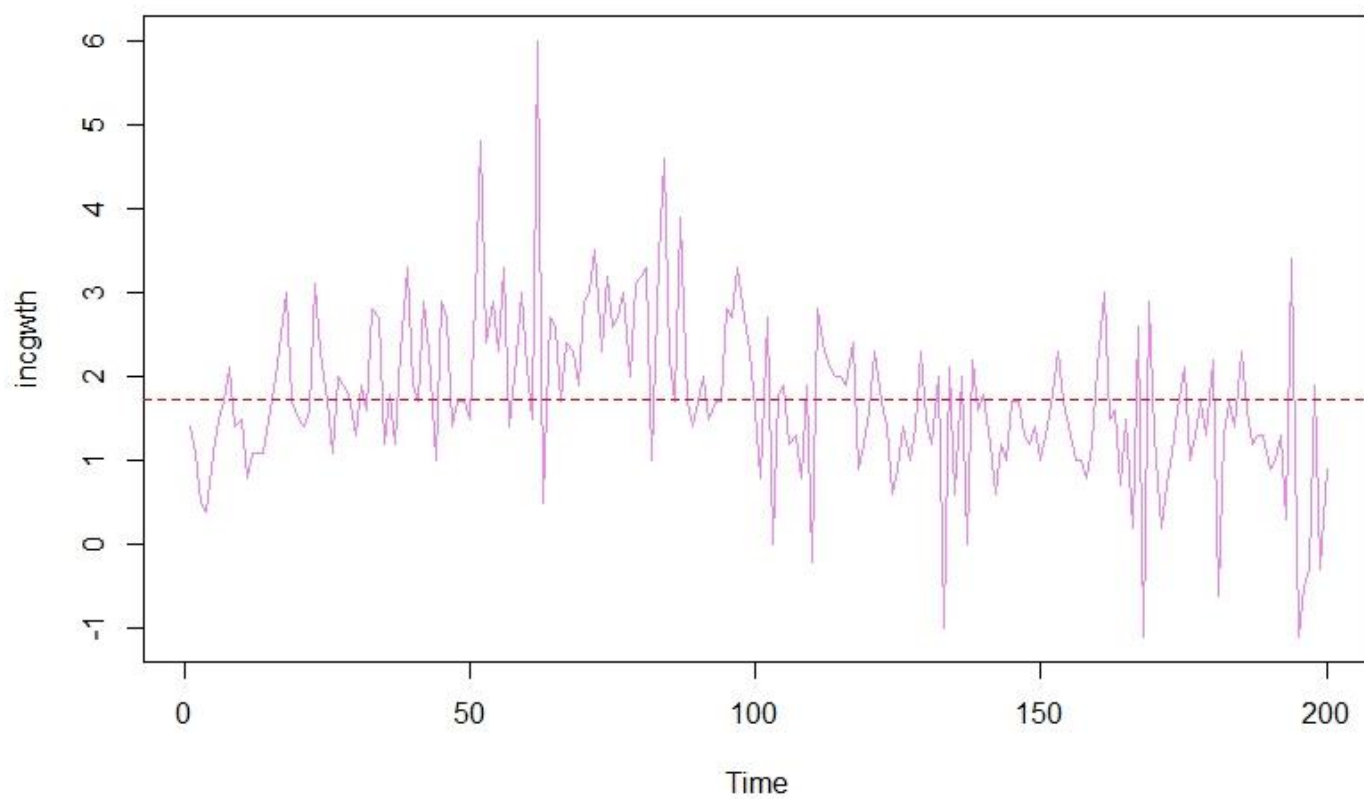
BIC reduced further. Better

model

correlogram shows significant correlation at lag 4

The correlogram shows very little correlation at lag 4.

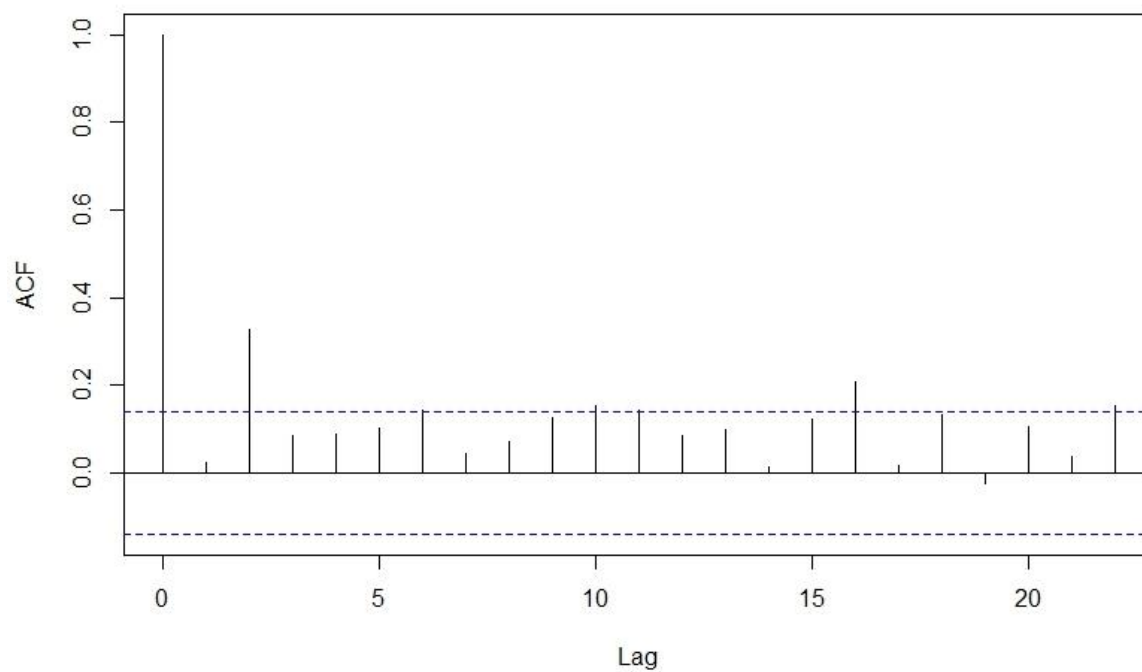




Both Time series appear to be fluctuating around their mean.

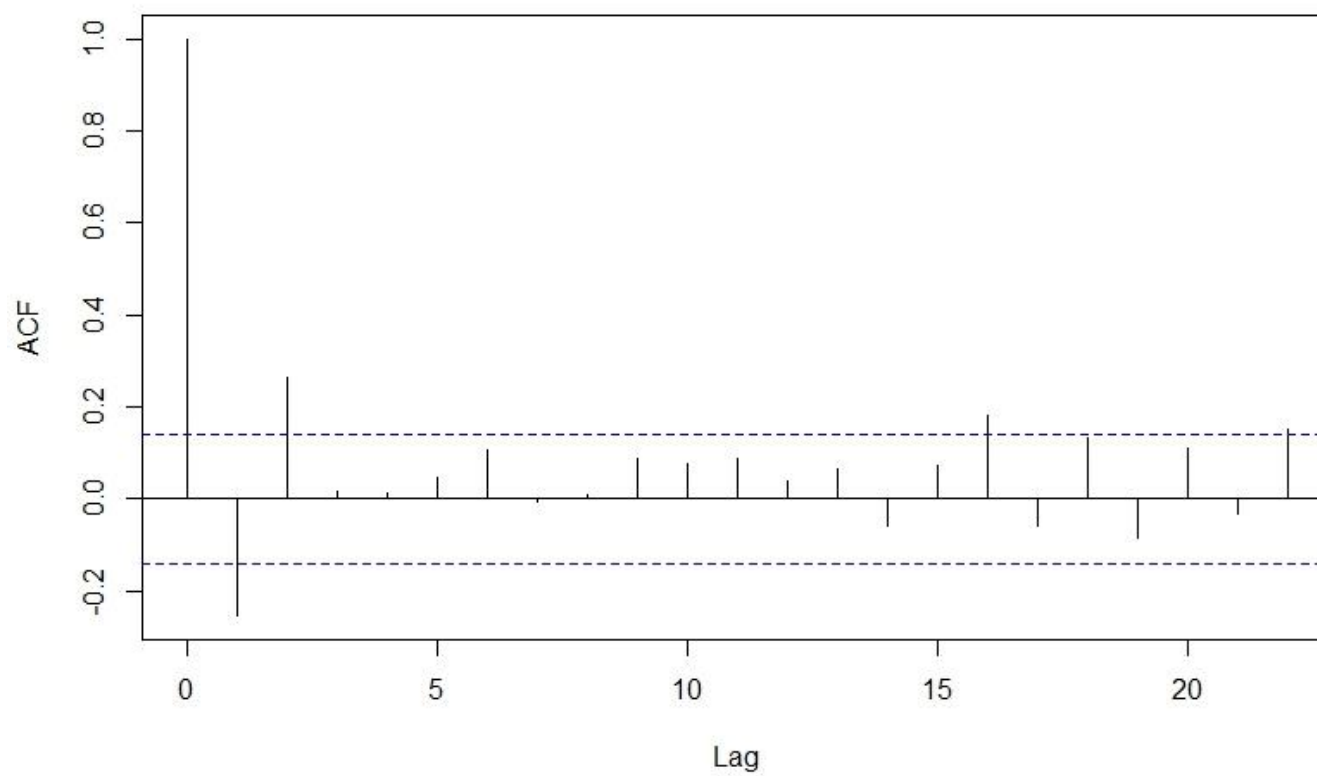
Correlogram part b

Series resid(model)



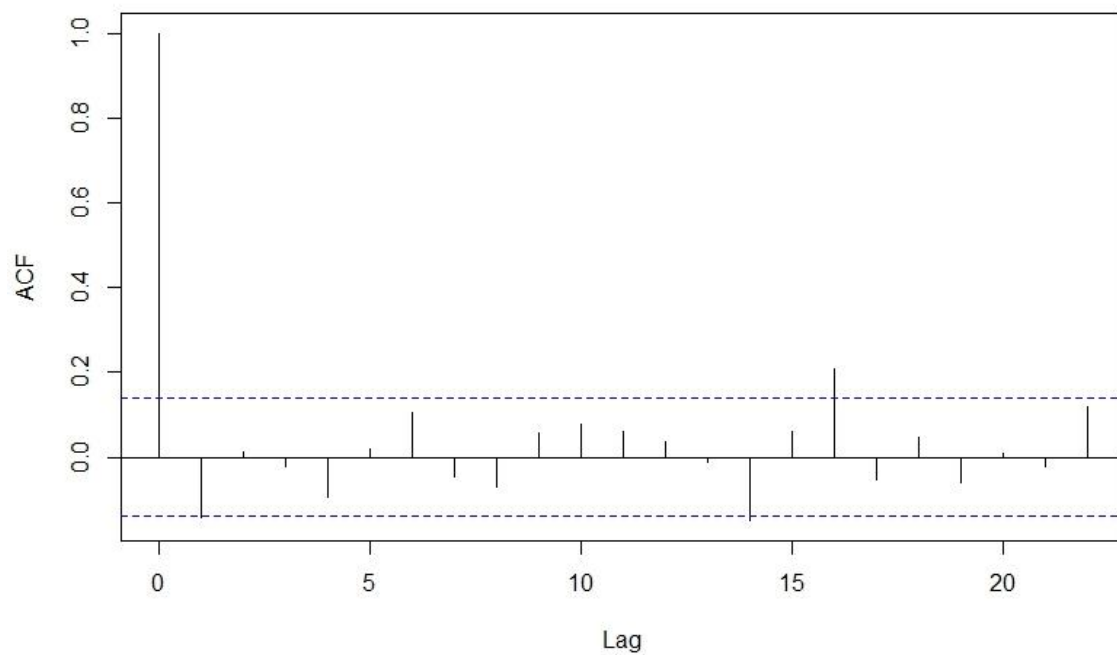
Correlogram part c

Series resid(fit)



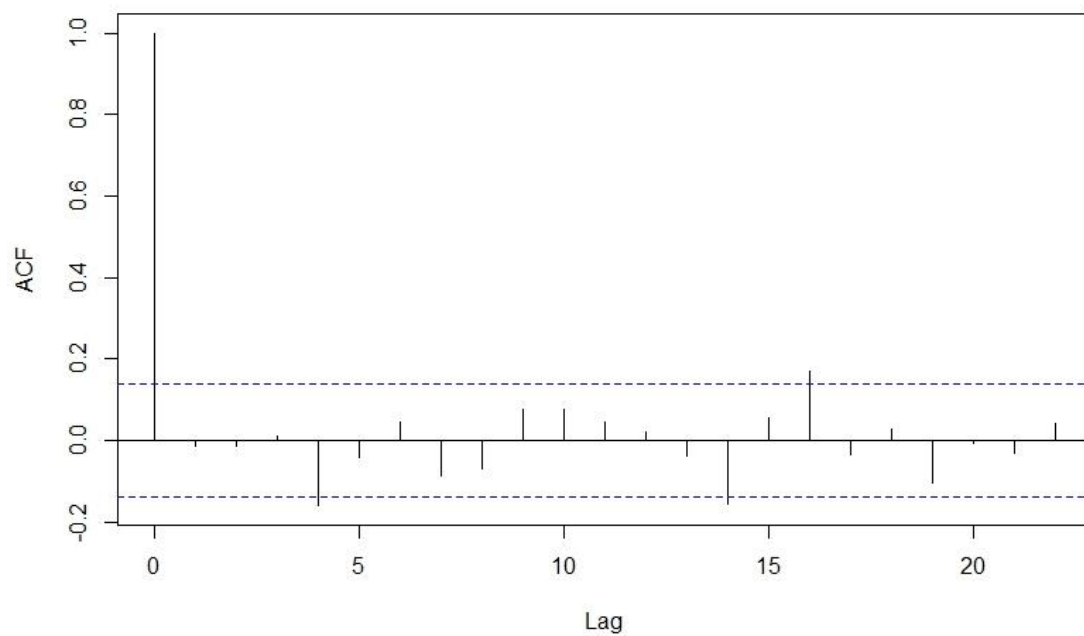
Correlogram part d

Series resid(fit2)



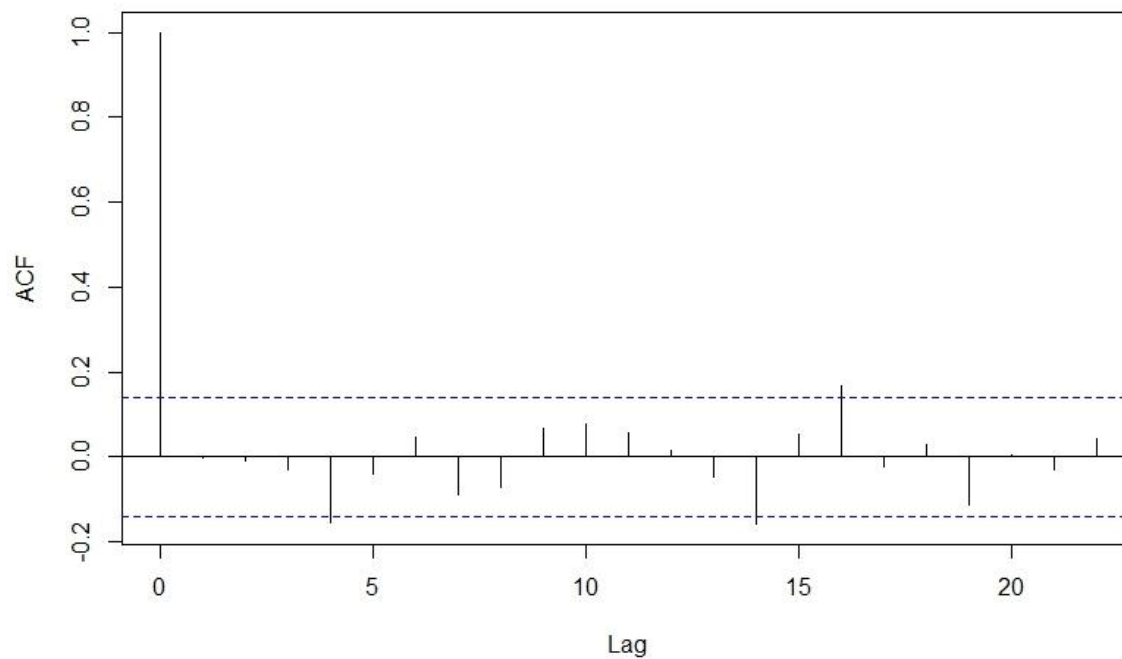
Correlogram part e

Series resid(fit3)



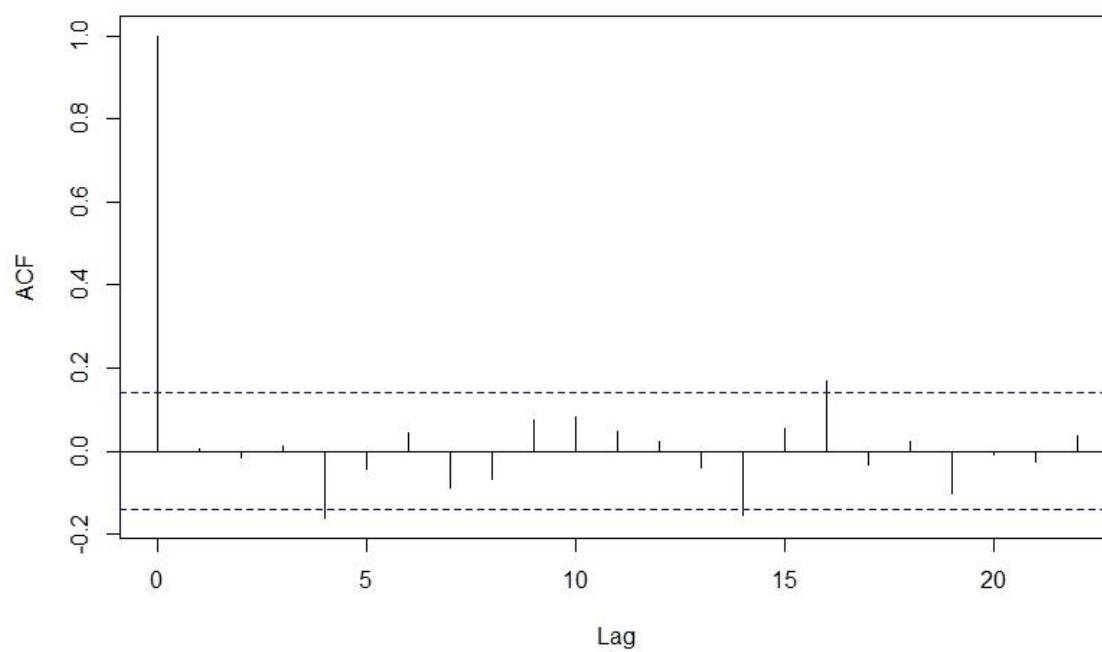
Correlogram Part f

Series resid(fit4)



Correlogram Part g

Series resid(fit5)



R Code:

```

> # -----
> #----- Q 9.2 -----
> #-----
> qt(0.975,101)
[1] 1.983731
> qt(0.95,101)
[1] 1.660081
> qt(0.90,101)
[1] 1.28999
>
> sqrt(1.3946+2.1606-(2*1.0406))
[1] 1.214084
>
> 5.644-(1.984*1.2140)
[1] 3.235424
> sqrt(1.3946+2.1606+1.4214-(2*1.0406)+(2*.0984)-(2*1.0367))
[1] 1.009356
> 7.909-(1.984*1.009)
[1] 5.907144
>
> # *****
> # ***** Q 9.4 *****
> # *****
>
> (-.31*.28)+(.09*.31)-(.03*.09)-(.37*.03)+(.17*.37)+(.39*.17)+(.03*.39)-(.03*.03)+(1.02*.03)
[1] 0.0979
>
> c <- c(0.28, -0.31,-0.09, 0.03, -0.37, -0.17, -0.39, -0.03, 0.03, 1.02)
> View(c)
> d <- c^2
> View(d)

```



```

> sum(d)
[1] 1.5436
> 0.0979/1.5436
[1] 0.06342317
>
> (-.09*.28)-(.03*.31)+(.37*.09)-(.17*.03)+(.39*.37)+(.03*.17)-(.03*.39)-(1.02*.03)
[1] 0.1008
>
> .1008/1.5436
[1] 0.06530189
>
> qnorm(.025)
[1] -1.959964
> qnorm(1-.025)
[1] 1.959964
>
> sqrt(10)*0.0653
[1] 0.2064967
>
> 1.959/sqrt(10)
[1] 0.6194902
>
> # *****
> # ***** Q 9.6 *****
> # *****
>
> qt(.975, 216)
[1] 1.971007
> qt(.975,215)
[1] 1.971059
>
> -53.51 - (1.971*16.98)

```

```

[1] -86.97758
> .3306/.0649
[1] 5.093991
>
> -58.61 - (1.971*14.10)
[1] -86.4011
> ts.plot(congwth, col = "blue")
> abline(h = mean(congwth), col = "red", lty = 2)
>
> ts.plot(incgwth, col = "violet")
> abline(h = mean(incgwth), col = "red", lty = 2)
> lag1congwh <- lag(congwth, -1)
> lag1incgwth <- lag(incgwth, -1)
> lag2congwh <- lag(congwth, -2)
> lag2incgwth <- lag(incgwth, -2)
> data <- cbind(consumptions, lag1congwh, lag2congwh, lag1incgwth, lag2congwh)
> View(data)
>
> data2 <- data[-c(1,2,3), ]
>
> data3 <- as.data.frame(data2)
>
> consumption2 <- consumptn[-c(1,2,3), ]
> incgwth2<- ts((consumption2$incgwth))
>
> congwh2 <- ts((consumption2$congwh))
>
> consumptionTS2 <- cbind(congwh2, incgwth2)
> # *****
> # ** Model 1 *****
> #*****
> model1 <- lm(congwh2 ~ incgwth2)

```

```
> summary(model1)
```

Call:

```
lm(formula = congwth2 ~ incgwth2)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.84905	-0.41333	-0.01333	0.37659	2.10138

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.97384	0.09961	9.776	<2e-16 ***
incgwth2	0.44958	0.04967	9.052	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6878 on 195 degrees of freedom

Multiple R-squared: 0.2959, Adjusted R-squared: 0.2923

F-statistic: 81.94 on 1 and 195 DF, p-value: < 2.2e-16

```
> acf(resid(model))
```

```
> anova(model1)
```

Analysis of Variance Table

Response: congwth2

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
incgwth2	1	38.764	38.764	81.939	< 2.2e-16 ***
Residuals	195	92.253	0.473		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
> log(81.94/197) + (4/197)
```

```
[1] -0.8569119
```



```

>
> log(81.94/197) + 2*(log(197)/197)
[1] -0.8235799
> # *****
> # ***** Adding Lag 1 gor congwth *****
> # *****
> fit <- lm(consumptionts.congwth ~ lag1congwth + consumptionts.incgwth, data = data3)
> summary(fit)

```

Call:

```

lm(formula = consumptionts.congwth ~ lag1congwth + consumptionts.incgwth,
    data = data3)

```

Residuals:

```

      Min       1Q   Median       3Q      Max
-2.65079 -0.45241 -0.00952  0.40190  1.78389

```

Coefficients:

```

              Estimate Std. Error t value Pr(>|t|)
(Intercept)    0.67158   0.11883   5.652 5.61e-08 ***
lag1congwth     0.27137   0.06351   4.273 3.02e-05 ***
consumptionts.incgwth 0.35012   0.05299   6.607 3.68e-10 ***
---

```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6593 on 194 degrees of freedom

(2 observations deleted due to missingness)

Multiple R-squared: 0.3564, Adjusted R-squared: 0.3498

F-statistic: 53.73 on 2 and 194 DF, p-value: < 2.2e-16

```

> anova(fit)

```

Analysis of Variance Table

Response: consumptionts.congwth

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
lag1congwth	1	27.726	27.7258	63.793	1.196e-13 ***
consumptionts.incgwth	1	18.975	18.9747	43.658	3.677e-10 ***
Residuals	194	84.316	0.4346		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

>

> log(84.316/197) +(6/197)

[1] -0.8181752

>

> log(84.316/197) + (3 * log(197)/ 197)

[1] -0.7681772

>

> acf(resid(fit))

>

>

> # *****

> # ***** Adding Lag 1 and lag 2 for congwth *****

> # *****

>

> fit2 <- lm(consumptionts.congwth ~ lag1congwth + lag2congwth + consumptionts.incgwth, data = data3)

> summary(fit2)

Call:

lm(formula = consumptionts.congwth ~ lag1congwth + lag2congwth +
consumptionts.incgwth, data = data3)

Residuals:

Min	1Q	Median	3Q	Max
-2.67655	-0.34503	-0.02734	0.33613	1.53752

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	0.42490	0.12541	3.388	0.000853	***
lag1congwth	0.15940	0.06527	2.442	0.015498	*
lag2congwth	0.28059	0.06146	4.565	8.87e-06	***
consumptionts.incgwth	0.32162	0.05086	6.324	1.73e-09	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6279 on 193 degrees of freedom

(2 observations deleted due to missingness)

Multiple R-squared: 0.4192, Adjusted R-squared: 0.4101

F-statistic: 46.43 on 3 and 193 DF, p-value: < 2.2e-16

> anova(fit2)

Analysis of Variance Table

Response: consumptionts.congwth

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
lag1congwth	1	27.726	27.7258	70.318	1.045e-14 ***
lag2congwth	1	11.422	11.4224	28.969	2.118e-07 ***
consumptionts.incgwth	1	15.770	15.7705	39.997	1.730e-09 ***
Residuals	193	76.098	0.3943		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

>

> log(76.098/197) +(8/197)

[1] -0.9105726

>

> log(76.098/197) + (4 * log(197)/ 197)

[1] -0.8439086


```

>
> acf(resid(fit2))
>
>
> # *****
> # Adding Lag 1 and lag 2 for congwth and lag 1 for inc gwth
> # *****
>
> fit3 <- lm(consumptions.congwth ~ lag1congwth + lag2congwth + consumptions.incgwth + lag1incgwth, data =
data3)
> summary(fit3)

```

Call:

```
lm(formula = consumptions.congwth ~ lag1congwth + lag2congwth +
    consumptions.incgwth + lag1incgwth, data = data3)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.21541	-0.34194	-0.02908	0.33843	1.52952

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	0.33195	0.12193	2.723	0.007075	**
lag1congwth	0.02332	0.06993	0.334	0.739093	
lag2congwth	0.21013	0.06103	3.443	0.000706	***
consumptions.incgwth	0.34931	0.04909	7.116	2.15e-11	***
lag1incgwth	0.23335	0.05392	4.328	2.42e-05	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6009 on 192 degrees of freedom

(2 observations deleted due to missingness)

Multiple R-squared: 0.4708, Adjusted R-squared: 0.4598

F-statistic: 42.7 on 4 and 192 DF, p-value: < 2.2e-16

```
> anova(fit3)
```

Analysis of Variance Table

Response: consumptions.congwth

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
lag1congwth	1	27.726	27.7258	76.778	1.003e-15 ***
lag2congwth	1	11.422	11.4224	31.631	6.511e-08 ***
consumptions.incgwth	1	15.770	15.7705	43.672	3.731e-10 ***
lag1incgwth	1	6.764	6.7641	18.731	2.420e-05 ***
Residuals	192	69.334	0.3611		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
>
```

```
> log(69.334/197) +(10/197)
```

```
[1] -0.9935069
```

```
>
```

```
> log(69.334/197) + (5 * log(197))/ 197)
```

```
[1] -0.9101769
```

```
> sum((residuals(fit3)^2))
```

```
[1] 69.33421
```

```
>
```

```
> acf(resid(fit3))
```

```
>
```

```
> # *****
```

```
> # Adding t-3 and t-2 *****
```

```
> # *****
```

```
>
```

```
> lag3congwth <- lag(congwth, -3)
```

```
>
```

```

> data5 <- cbind(consumptions, lag1congwth, lag2congwth, lag3congwth, lag1incgwth, lag2incgwth)
> View(data5)
>
> data5 <- data5[-c(1,2,3,4), ]
>
> data5 <- as.data.frame(data5)
> data5 <- data5[-c(197,198,199),]
>
> fit4 <- lm(consumptions.congwth ~ lag1congwth + lag2congwth + lag3congwth +
+           consumptions.incgwth + lag1incgwth + lag2incgwth, data = data5)
> summary(fit4)

```

Call:

```

lm(formula = consumptions.congwth ~ lag1congwth + lag2congwth +
    lag3congwth + consumptions.incgwth + lag1incgwth + lag2incgwth,
    data = data5)

```

Residuals:

Min	1Q	Median	3Q	Max
-2.20805	-0.35351	-0.01537	0.35356	1.53709

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.32262	0.12920	2.497	0.01337 *
lag1congwth	0.01221	0.07566	0.161	0.87195
lag2congwth	0.20848	0.07079	2.945	0.00364 **
lag3congwth	0.04094	0.06508	0.629	0.53013
consumptions.incgwth	0.34136	0.05068	6.735	1.91e-10 ***
lag1incgwth	0.22960	0.05600	4.100	6.13e-05 ***
lag2incgwth	-0.01029	0.05736	-0.179	0.85775

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6044 on 189 degrees of freedom

Multiple R-squared: 0.4675, Adjusted R-squared: 0.4506

F-statistic: 27.66 on 6 and 189 DF, p-value: < 2.2e-16

```
> acf(resid(fit4))
```

```
> anova(fit4)
```

Analysis of Variance Table

Response: consumptions.congwth

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
lag1congwth	1	26.509	26.5091	72.5622	5.019e-15 ***
lag2congwth	1	11.706	11.7064	32.0436	5.527e-08 ***
lag3congwth	1	1.517	1.5172	4.1530	0.04295 *
consumptions.incgwth	1	14.194	14.1935	38.8514	2.913e-09 ***
lag1incgwth	1	6.682	6.6819	18.2900	3.008e-05 ***
lag2incgwth	1	0.012	0.0118	0.0322	0.85775
Residuals	189	69.047	0.3653		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
>
```

```
> # *****
```

```
> # Droppping Lag 1 for congwth and lag 1 for inc gwth
```

```
> #*****
```

```
>
```

```
> fit5 <- lm(consumptions.congwth ~ lag2congwth + consumptions.incgwth + lag1incgwth, data = data3)
```

```
> summary(fit5)
```

Call:

```
lm(formula = consumptions.congwth ~ lag2congwth + consumptions.incgwth +  
lag1incgwth, data = data3)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.21171	-0.34415	-0.01992	0.34549	1.52367

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.34074	0.11877	2.869	0.004577 **
lag2congwth	0.21428	0.05961	3.594	0.000413 ***
consumptionts.incgwth	0.35545	0.04539	7.831	3.14e-13 ***
lag1incgwth	0.24144	0.04805	5.025	1.15e-06 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.5995 on 193 degrees of freedom

(2 observations deleted due to missingness)

Multiple R-squared: 0.4705, Adjusted R-squared: 0.4623

F-statistic: 57.16 on 3 and 193 DF, p-value: < 2.2e-16

> anova(fit5)

Analysis of Variance Table

Response: consumptionts.congwth

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
lag2congwth	1	29.136	29.1363	81.057	< 2.2e-16 ***
consumptionts.incgwth	1	23.431	23.4307	65.184	7.181e-14 ***
lag1incgwth	1	9.076	9.0756	25.248	1.145e-06 ***
Residuals	193	69.374	0.3595		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

>

> log(69.374/197) +(8/197)

[1] -1.003082

```
>
```

```
> log(69.334/197) + (4 * log(197)/ 197)
```

```
[1] -0.9369952
```

```
> sum((residuals(fit3)^2))
```

```
[1] 69.33421
```

```
>
```

```
> acf(resid(fit5))
```