

Time Series

Homework 2

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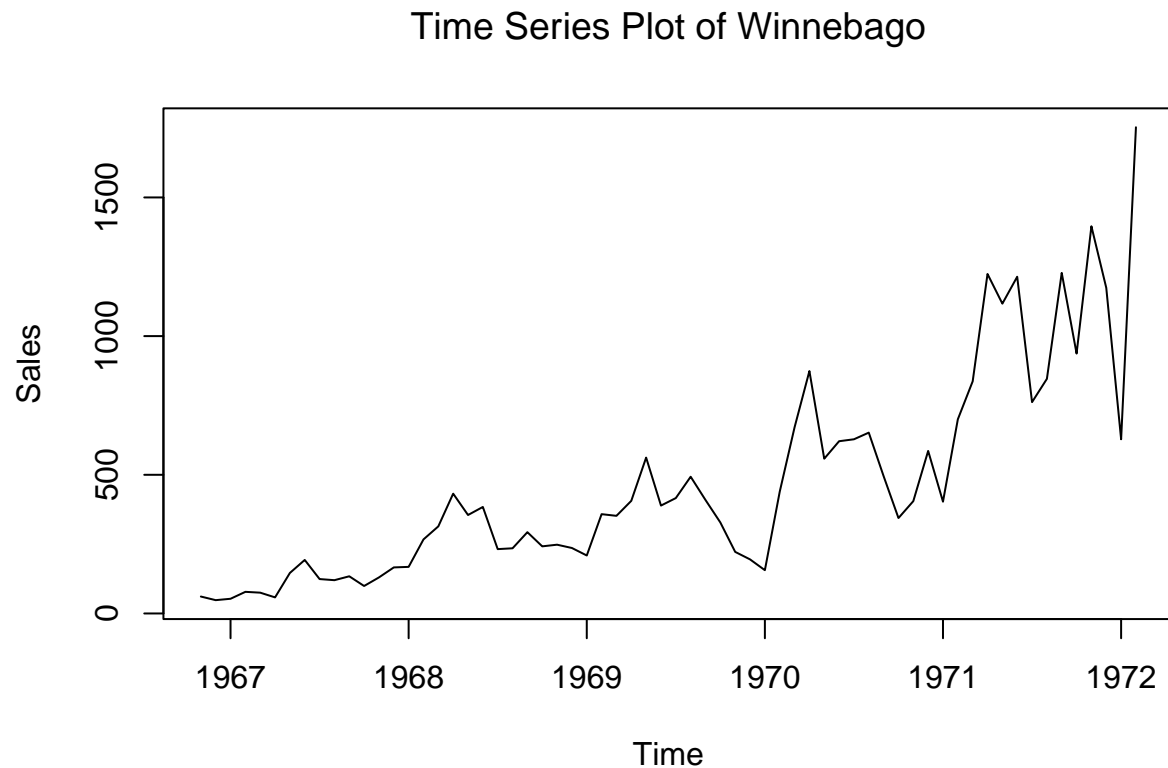
Problem 1

Problem 2

a)

```
library(TSA)
library(tseries)
```

```
data(winnebago)
plot(winnebago, xlab = expression("Time"), ylab = expression("Sales"), main = expression("Time Series Plot of Winnebago Sales"))
```



Interpretation: - The time series shows an overall upwards trend between the years 1967 and 1972. - Between the years 1970 and 1972 the increase is at its highest.

b)

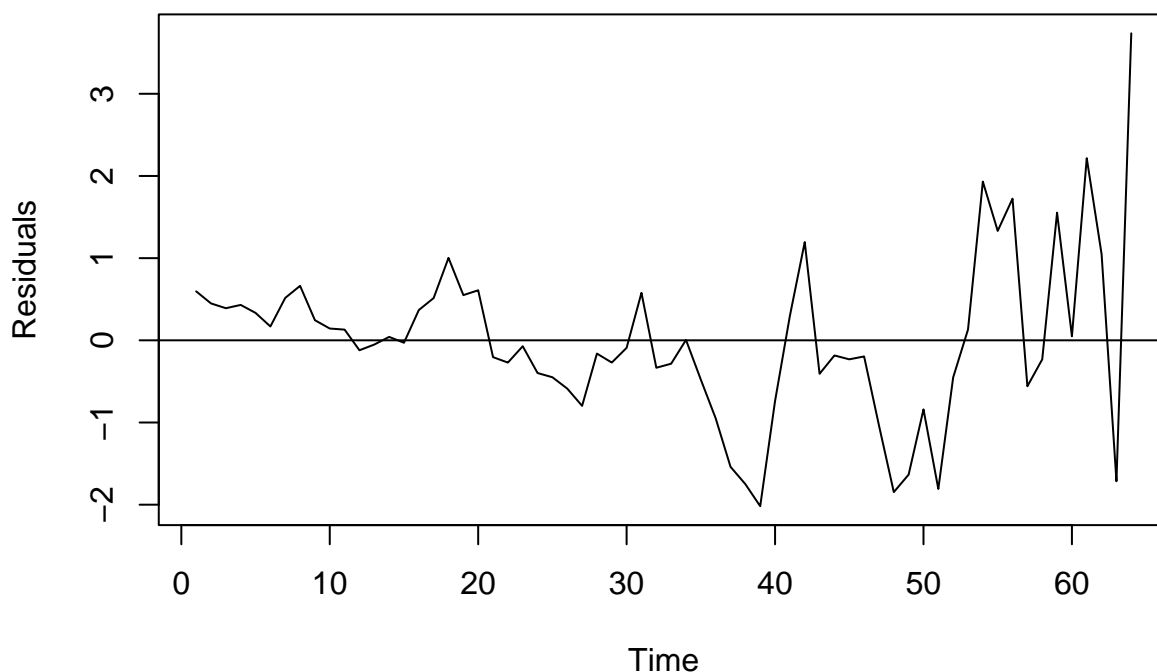
```
modell1 <- lm(winnebago ~ time(winnebago))
summary(modell1)
```

```
##
## Call:
## lm(formula = winnebago ~ time(winnebago))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -419.58  -93.13  -12.78   94.96  759.21
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -394885.68   33539.77  -11.77  <2e-16 ***
## time(winnebago)    200.74     17.03   11.79  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 209.7 on 62 degrees of freedom
## Multiple R-squared:  0.6915, Adjusted R-squared:  0.6865
## F-statistic: 138.9 on 1 and 62 DF,  p-value: < 2.2e-16
```

- We expect the wages to increase by \$200.74 per year
- 69.15 of the variance can be explained by the predictor variables
- the linear trend is significant because the p-value $2.2 \cdot 10^{-16}$ is smaller than 10^{-12}

```
res <- as.ts(rstandard(modell1))
plot(res, xlab = expression("Time"), ylab = expression("Residuals"), main = "Plot of Residuals versus T",
abline(h = 0))
```

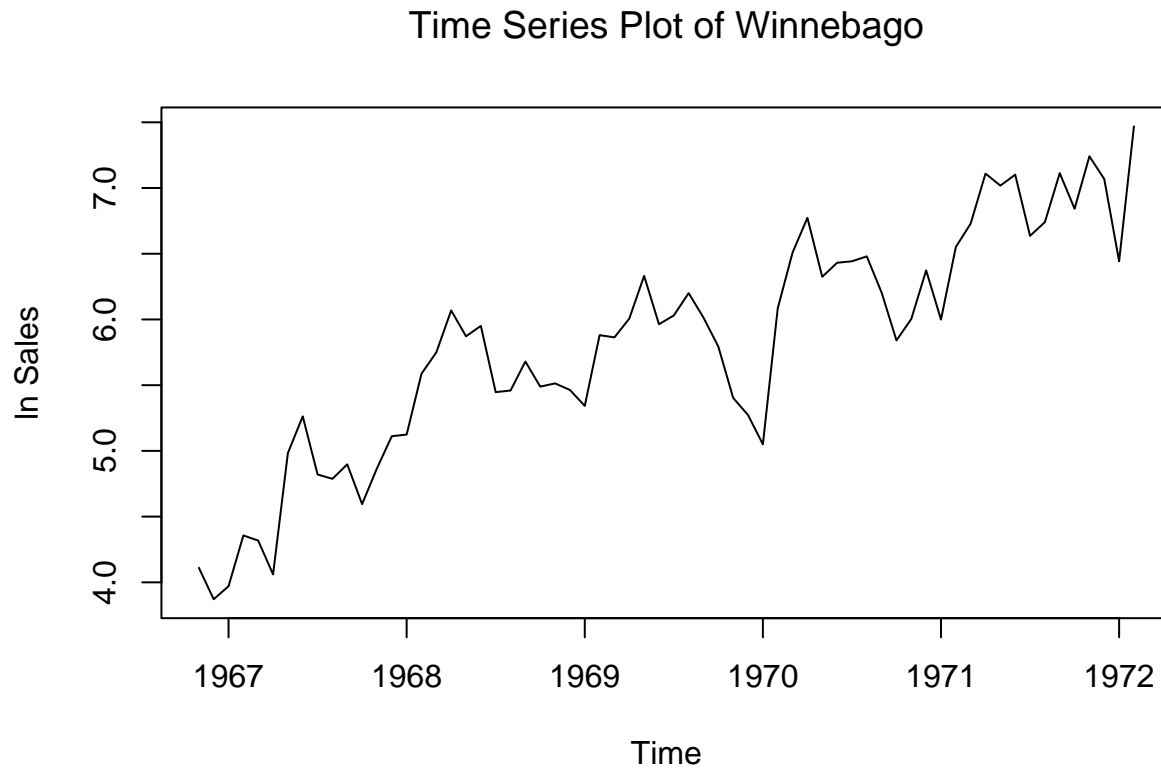
Plot of Residuals versus Time



Interpretation: - The residuals plot shows somewhat random movement around zero. - More uneven spread between ca. 35 to 65 in comparison to 0 to 35. - There may be a “seasonal” cyclical trend.

c)

```
ln_winnebago <- log(winnebago)
plot(ln_winnebago, xlab = expression("Time"), ylab = expression("ln Sales"), main = expression("Time Series Plot of Winnebago"))
```



Interpretation: - The time series plot of the transformed values shows a linear upward trend (which corresponds with the untransformed data) - There is one dip around the year 1970 - The “seasonal” trend seems more pronounced

d)

```
model2 <- lm(ln_winnebago ~ time(ln_winnebago))
summary(model2)
```

```
##
## Call:
## lm(formula = ln_winnebago ~ time(ln_winnebago))
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-1.03669	-0.20823	0.04995	0.25662	0.86223

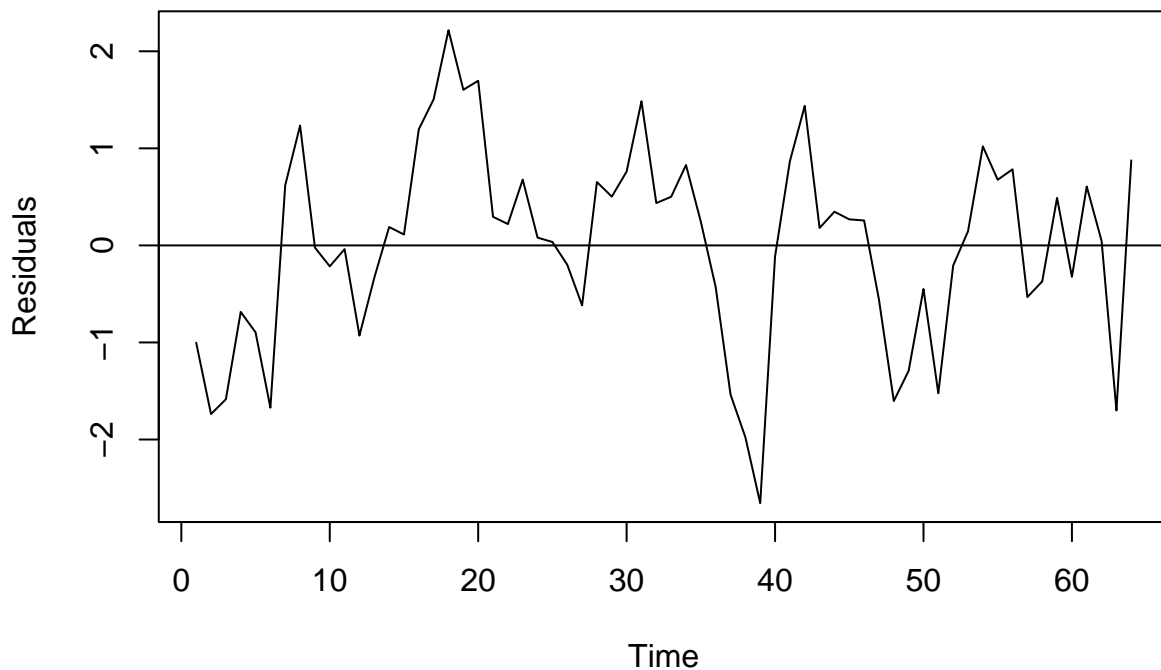
```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-984.93878	62.99472	-15.63	<2e-16 ***
time(ln_winnebago)	0.50306	0.03199	15.73	<2e-16 ***

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3939 on 62 degrees of freedom
## Multiple R-squared:  0.7996, Adjusted R-squared:  0.7964
## F-statistic: 247.4 on 1 and 62 DF,  p-value: < 2.2e-16

res2 <- as.ts(rstandard(model2))
plot(res2, xlab = expression("Time"), ylab = expression("Residuals"), main = "ln Plot of Residuals versus Time",
abline(h = 0))
```

ln Plot of Residuals versus Time



Interpretation: - the ln-transformed residuals plot shows random movement around 0 - There seems to be an overall cyclical trend

e) & f)

```
months <- season(winnebago)
model3 <- lm(ln_winnebago~months-1)
summary(model3)

##
## Call:
## lm(formula = ln_winnebago ~ months - 1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.94319 -0.40444  0.02541  0.59421  1.71807
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```

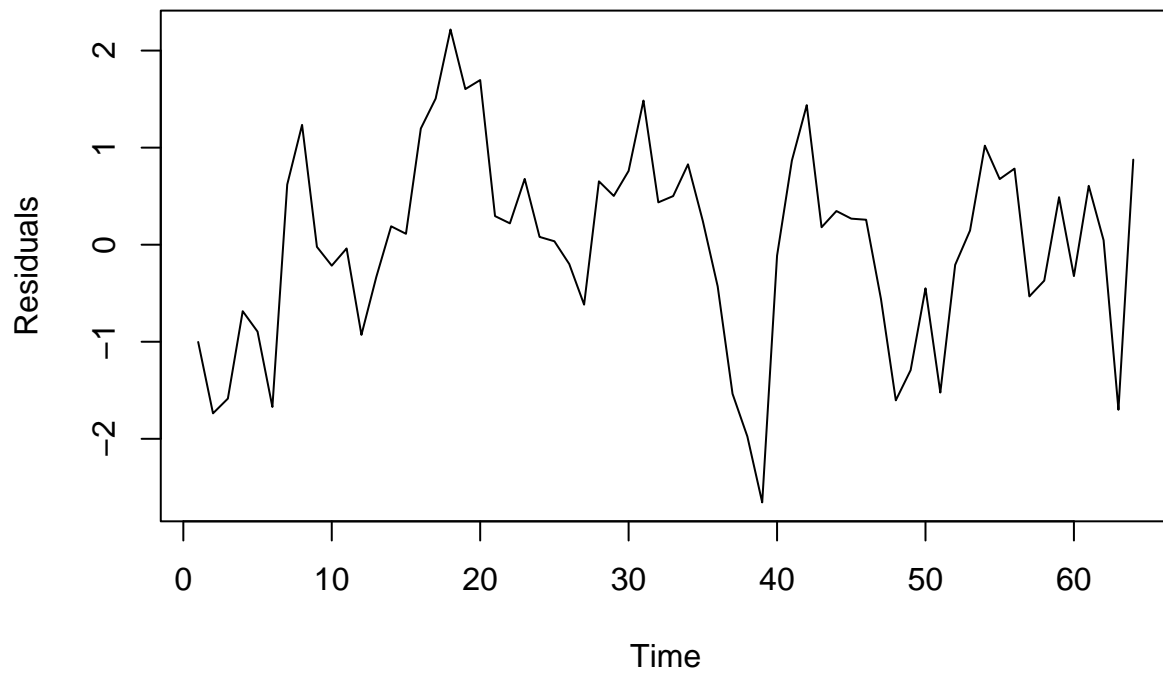
- 98.01 of the variance can be explained by the predictor variables
- the linear trend is significant because the p-value $2.2 \cdot 10^{-16}$ is smaller than 10^{-12}

```
plot(rstudent(model3), x = as.vector(time(winnebago)), xlab="Time", ylab="Residuals", main="Standardized  
points(y=rstudent(model3), x=as.vector(time(winnebago)), pch = as.vector(season(winnebago)))
```



```
res2 <- as.ts(rstandard(model2))
plot(res2, xlab=expression("Time"), ylab=expression("Residuals"), main="Plot of Residuals versus Time")
```

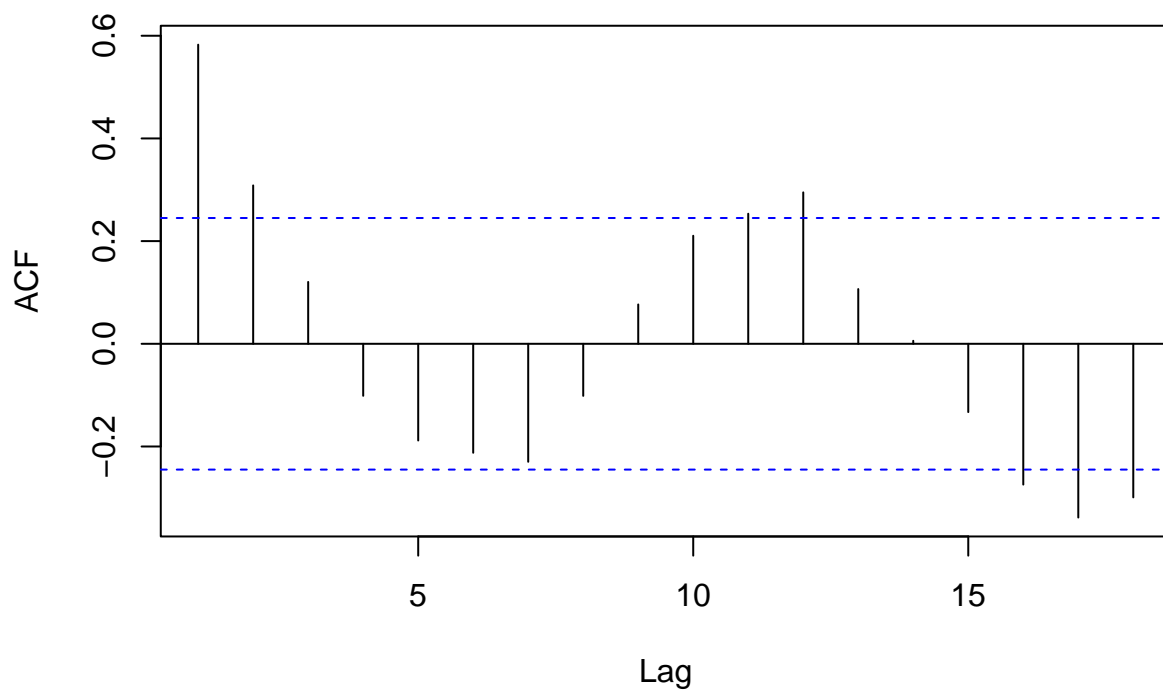
Plot of Residuals versus Time



h)

```
acf(res2, main="Autocorrelation Plot of Residuals")$acf
```

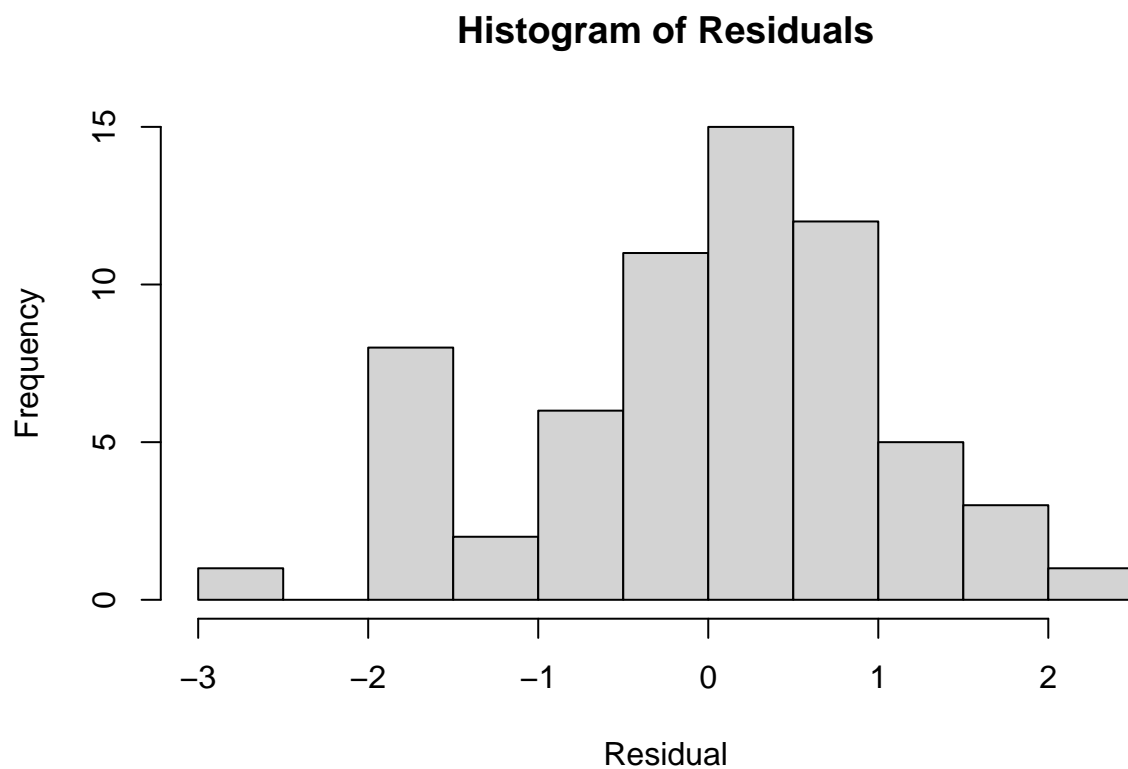
Autocorrelation Plot of Residuals



```
## , , 1
##
##      [,1]
## [1,] 0.582621682
## [2,] 0.308516762
## [3,] 0.120483855
## [4,] -0.101296290
## [5,] -0.188376915
## [6,] -0.212242495
## [7,] -0.229810429
## [8,] -0.101442132
## [9,] 0.076663936
## [10,] 0.210470866
## [11,] 0.253334338
## [12,] 0.294944901
## [13,] 0.106610155
## [14,] 0.005900711
## [15,] -0.132994388
## [16,] -0.274233631
## [17,] -0.338397733
## [18,] -0.299142743
```

h) i)

```
hist(res2, xlab="Residual", main="Histogram of Residuals")
```



```
shapiro.test(res2)
```

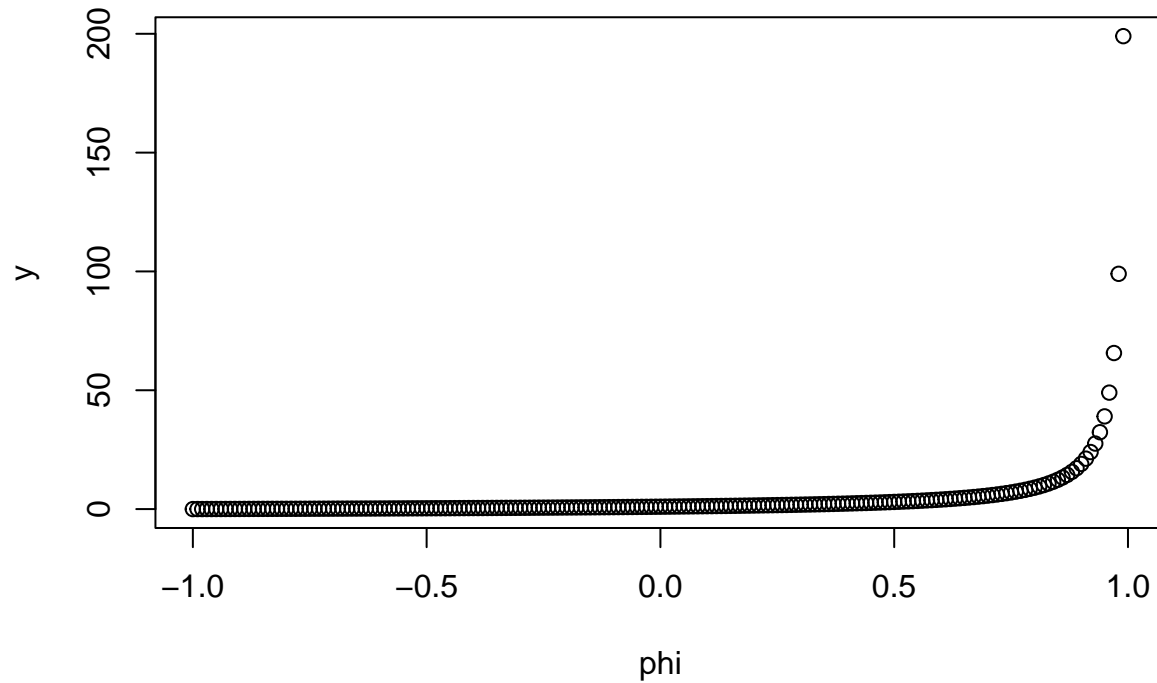
```
##
## Shapiro-Wilk normality test
```

```
##  
## data:  res2  
## W = 0.97939, p-value = 0.3603
```

Exercise 3

c)

```
phi <- seq(-1, 1, by = 0.01)  
y <- (1 + phi)/(1-phi)  
plot(phi, y)
```



The closer ϕ is to -1, the closer the variance is to 0, thus the precision increases. The closer ϕ is to 1, the more the variance increases, and thus the precision decreases.