

DSA 8020 R Session 11: Time Series Analysis

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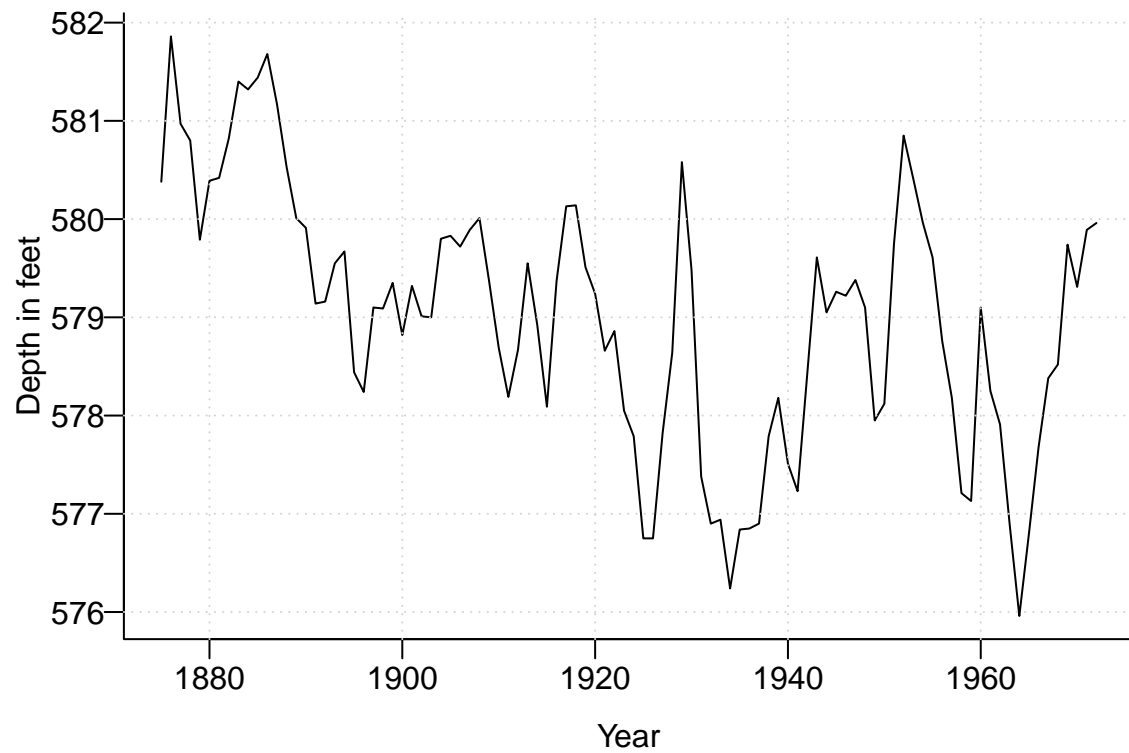
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Time Series Data

Lake Huron Time Series

Annual measurements of the level of Lake Huron in feet

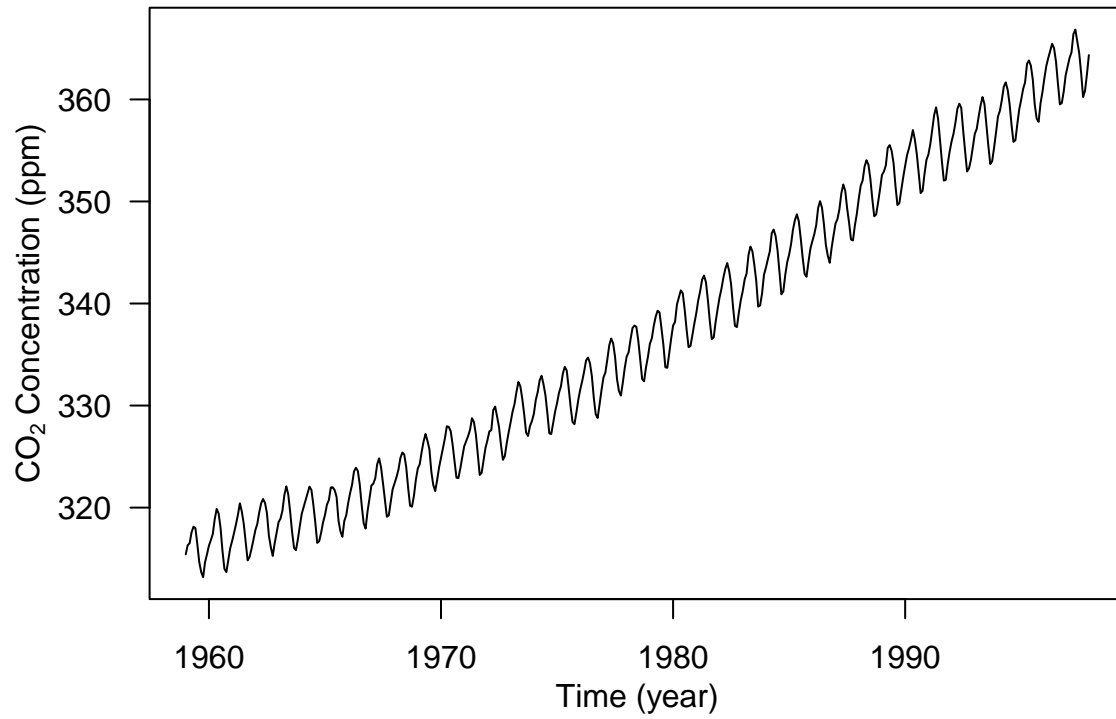
```
par(mar = c(3.2, 3.2, 0.5, 0.5), mgp = c(2, 0.5, 0), bty = "L")
data(LakeHuron)
plot(LakeHuron, ylab = "Depth in feet", xlab = "Year", las = 1)
grid()
```



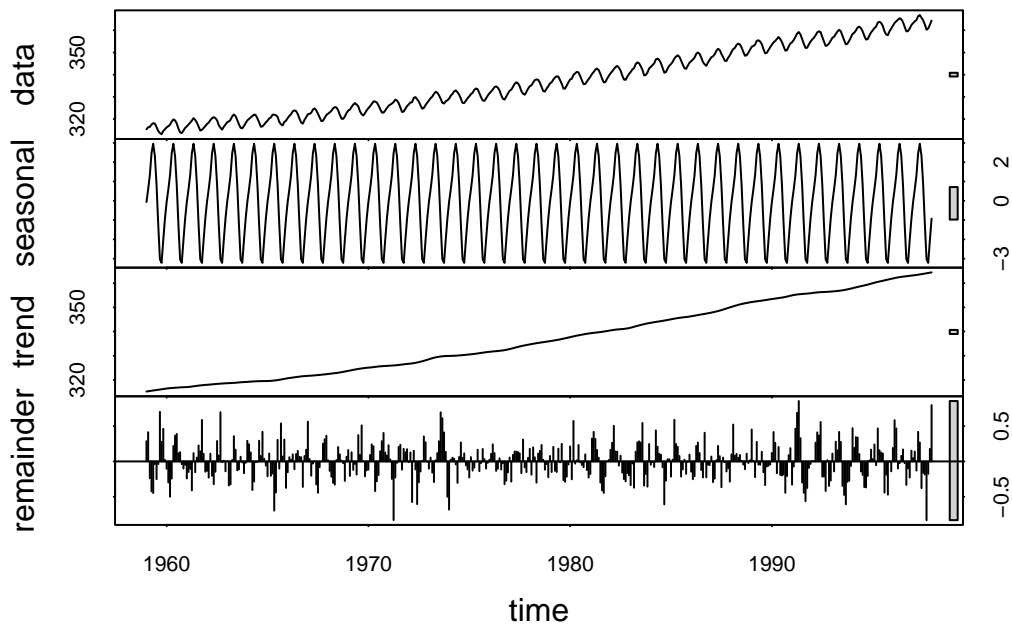
CO₂ Concentration

Atmospheric concentrations of CO₂ are expressed in parts per million (ppm) and reported in the preliminary 1997 SIO manometric mole fraction scale.

```
data(co2)
par(mar = c(3.8, 4, 0.8, 0.6))
plot(co2, las = 1, xlab = "", ylab = "")
mtext("Time (year)", side = 1, line = 2)
mtext(expression(paste("CO"[2], " Concentration (ppm)")), side = 2, line = 2.5)
```



```
# Seasonal and Trend decomposition using Loess (STL)
par(mar = c(4, 3.6, 0.8, 0.6))
stl <- stl(co2, s.window = "periodic")
plot(stl, las = 1)
```



U.S. monthly unemployment rates

```
library(quantmod)
getSymbols("UNRATE", src = "FRED")
```

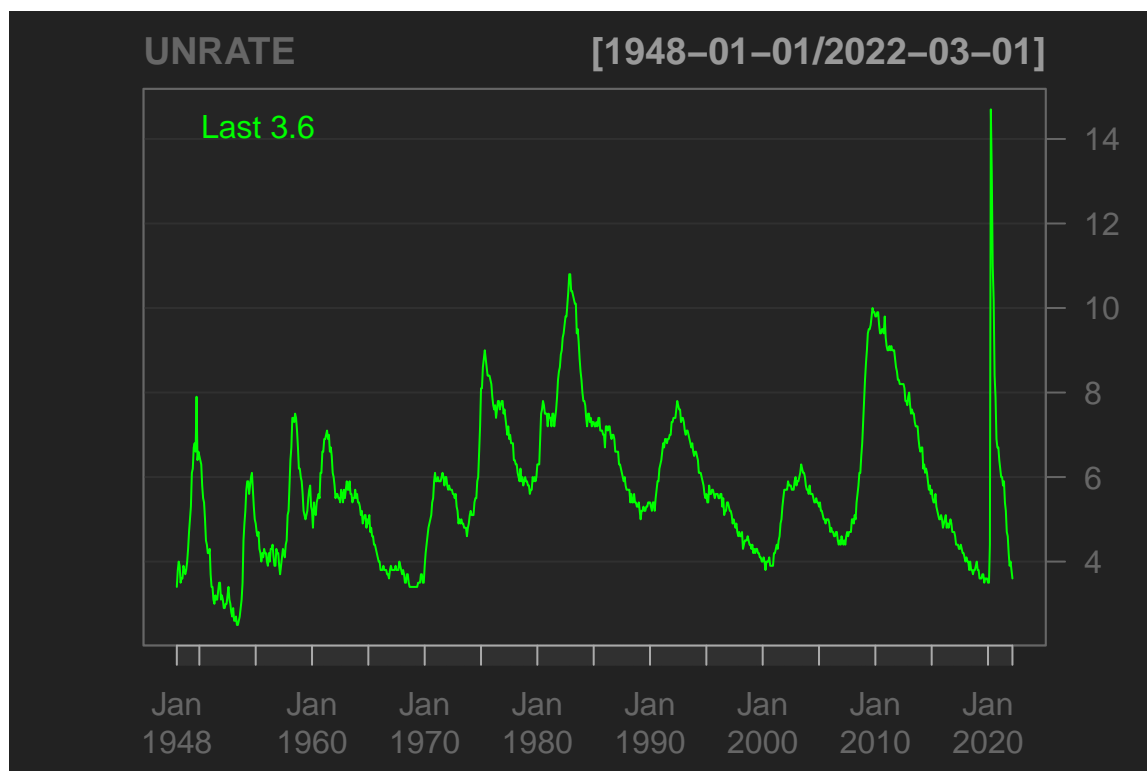
```
## [1] "UNRATE"
```

```
head(UNRATE); tail(UNRATE)
```

```
##           UNRATE
## 1948-01-01    3.4
## 1948-02-01    3.8
## 1948-03-01    4.0
## 1948-04-01    3.9
## 1948-05-01    3.5
## 1948-06-01    3.6
```

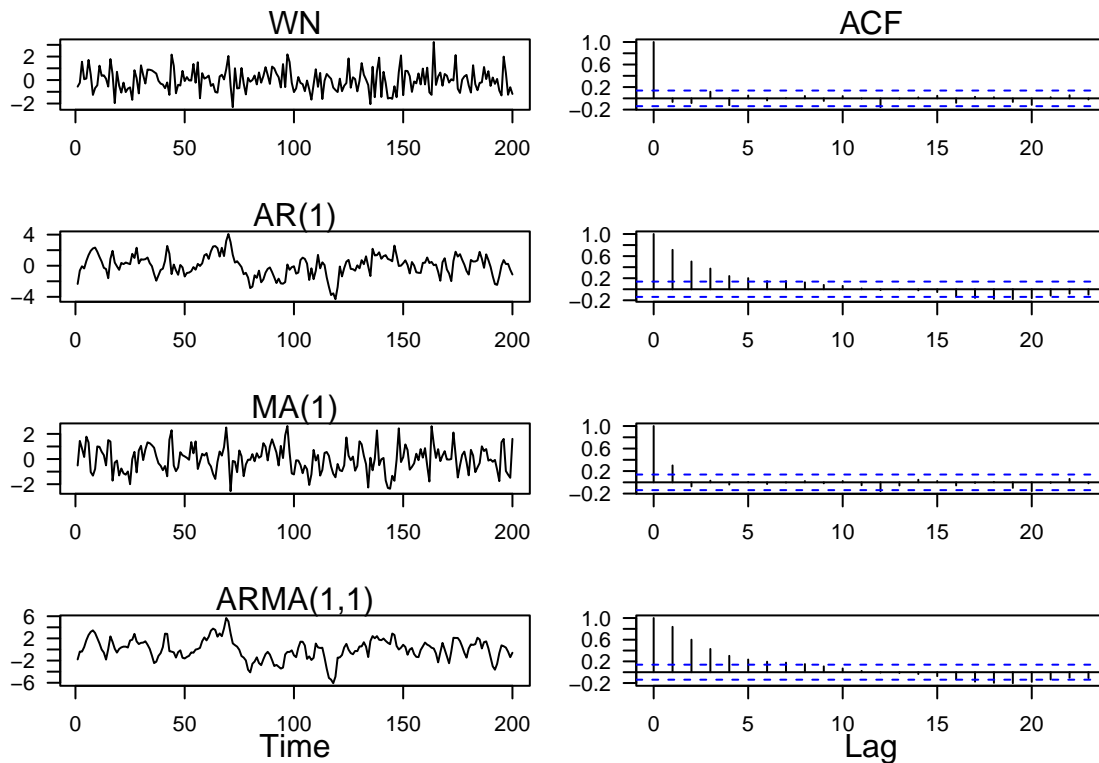
```
##           UNRATE
## 2021-10-01    4.6
## 2021-11-01    4.2
## 2021-12-01    3.9
## 2022-01-01    4.0
## 2022-02-01    3.8
## 2022-03-01    3.6
```

```
chartSeries(UNRATE)
```



ARMA: ACF and PACF

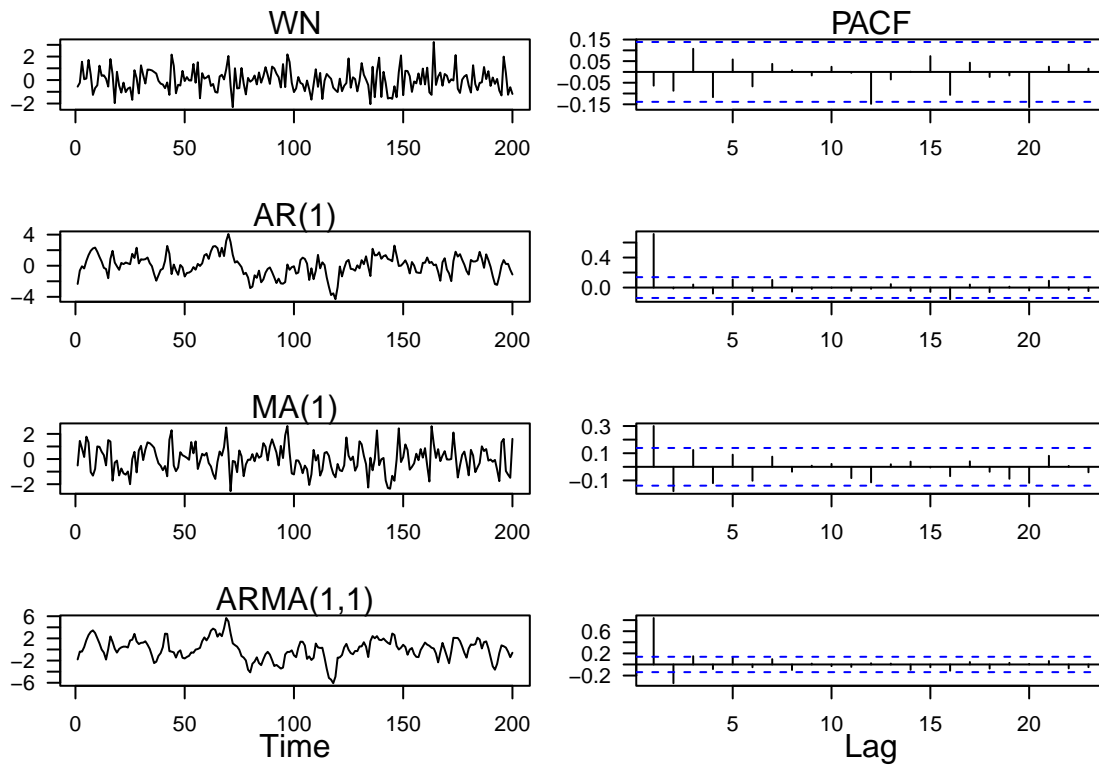
```
set.seed(123)
n = 200
WN <- rnorm(n)
par(mfrow = c(4, 2), mar = c(3.6, 3.6, 1.2, 0.6))
plot(1:n, WN, type = "l", las = 1, xlab = "", ylab = "")
mtext("WN")
acf(WN, xlab = "", ylab = "", main = "", las = 1)
mtext("ACF")
# AR(1) phi = 0.8
set.seed(123)
AR <- arima.sim(n = n, model = list(ar = 0.8))
plot(1:n, AR, type = "l", las = 1, xlab = "", ylab = "")
mtext("AR(1)")
acf(AR, xlab = "", ylab = "", main = "", las = 1)
# MA(1) theta = 0.5
set.seed(123)
MA <- arima.sim(n = n, model = list(ma = 0.5))
plot(1:n, MA, type = "l", las = 1, xlab = "", ylab = "")
mtext("MA(1)")
acf(MA, xlab = "", ylab = "", main = "", las = 1)
# ARMA(1, 1) phi = 0.8, theta = 0.5
set.seed(123)
ARMA <- arima.sim(n = n, model = list(ar = 0.8, ma = 0.5))
plot(1:n, ARMA, type = "l", las = 1, xlab = "", ylab = "")
mtext("ARMA(1,1)")
mtext("Time", side = 1, line = 2)
acf(ARMA, xlab = "", ylab = "", main = "", las = 1)
mtext("Lag", side = 1, line = 2)
```



```

par(mfrow = c(4, 2), mar = c(3.6, 3.6, 1.2, 0.6))
plot(1:n, WN, type = "l", las = 1, xlab = "", ylab = "")
mtext("WN")
pacf(WN, xlab = "", ylab = "", main = "", las = 1)
mtext("PACF")
# AR(1) phi = 0.8
set.seed(123)
AR <- arima.sim(n = n, model = list(ar = 0.8))
plot(1:n, AR, type = "l", las = 1, xlab = "", ylab = "")
mtext("AR(1)")
pacf(AR, xlab = "", ylab = "", main = "", las = 1)
# MA(1) theta = 0.5
set.seed(123)
MA <- arima.sim(n = n, model = list(ma = 0.5))
plot(1:n, MA, type = "l", las = 1, xlab = "", ylab = "")
mtext("MA(1)")
pacf(MA, xlab = "", ylab = "", main = "", las = 1)
# ARMA(1, 1) phi = 0.8, theta = 0.5
set.seed(123)
ARMA <- arima.sim(n = n, model = list(ar = 0.8, ma = 0.5))
plot(1:n, ARMA, type = "l", las = 1, xlab = "", ylab = "")
mtext("ARMA(1,1)")
mtext("Time", side = 1, line = 2)
pacf(ARMA, xlab = "", ylab = "", main = "", las = 1)
mtext("Lag", side = 1, line = 2)

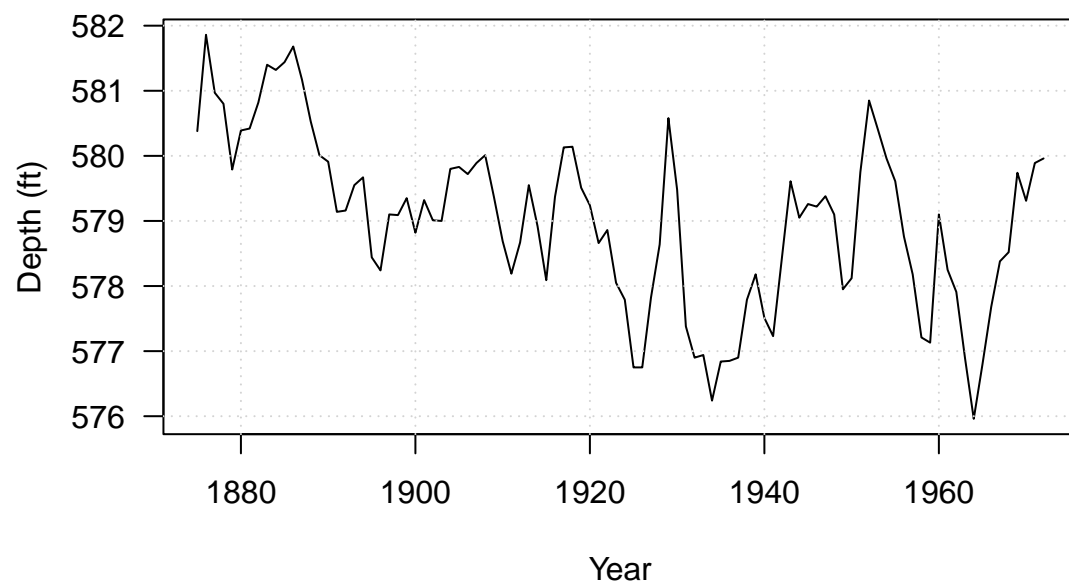
```



Lake Huron Case Study

Detrend

```
## Let us create a 'years' variable.
years <- time(LakeHuron)
## Plot time series
plot(LakeHuron, ylab = "Depth (ft)", xlab = "Year", las = 1)
grid()
```

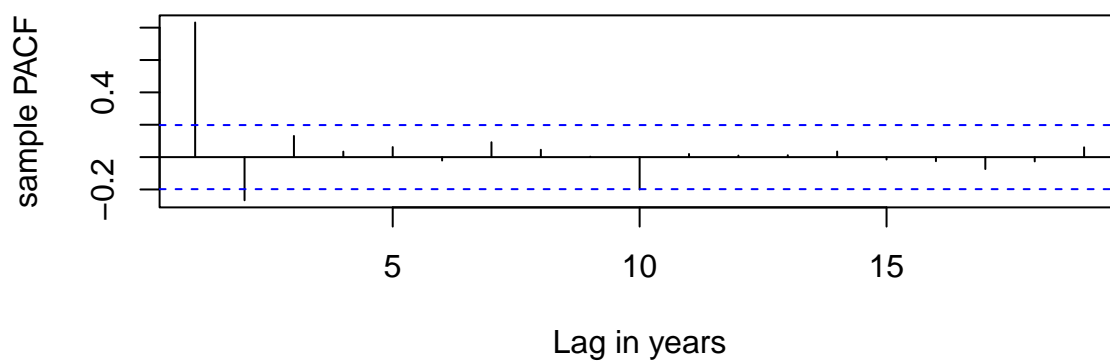
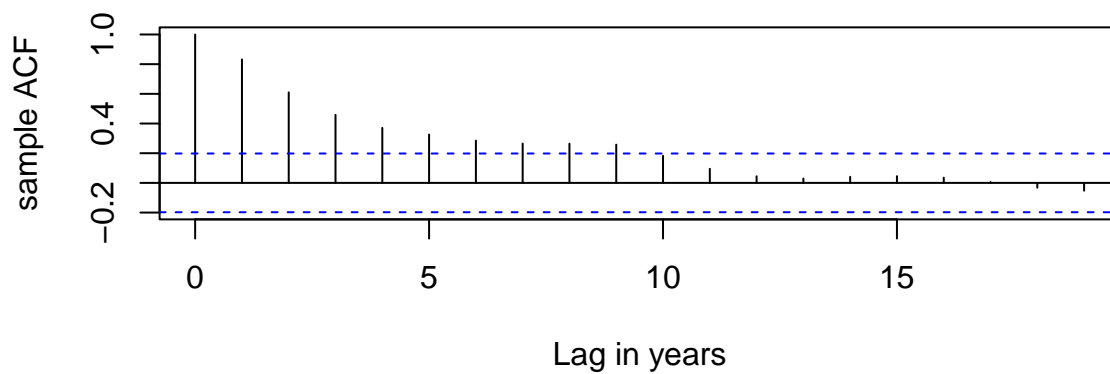


```
## ACF and PACF
```

```
par(mfrow = c(2, 1), mar = c(4, 4, 1, 1))
```

```
acf(LakeHuron, xlab="Lag in years", ylab = "sample ACF", main = "")
```

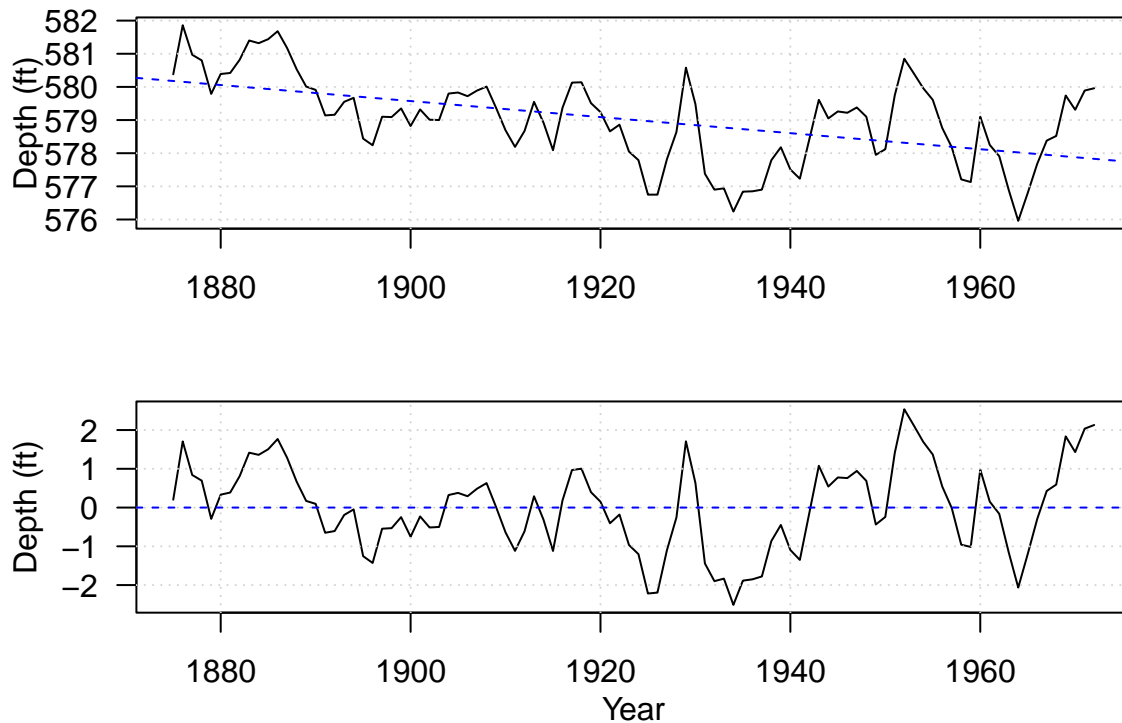
```
pacf(LakeHuron, xlab="Lag in years", ylab = "sample PACF", main = "")
```




```

# Estimate the linear trend
lm <- lm(LakeHuron ~ years)
par(mfrow = c(2, 1), mar = c(3.5, 3.5, 1, 0.6))
plot(LakeHuron, ylab = "", xlab = "", las = 1); grid()
abline(lm, col = "blue", lty = 2)
mtext("Depth (ft)", 2, line = 2.4)
deTrend <- resid(lm)
plot(1875:1972, deTrend, type = "l", ylab = "", xlab = "", las = 1); grid()
abline(h = 0, col = "blue", lty = 2)
mtext("Year", 1, line = 2)
mtext("Depth (ft)", 2, line = 2.4)

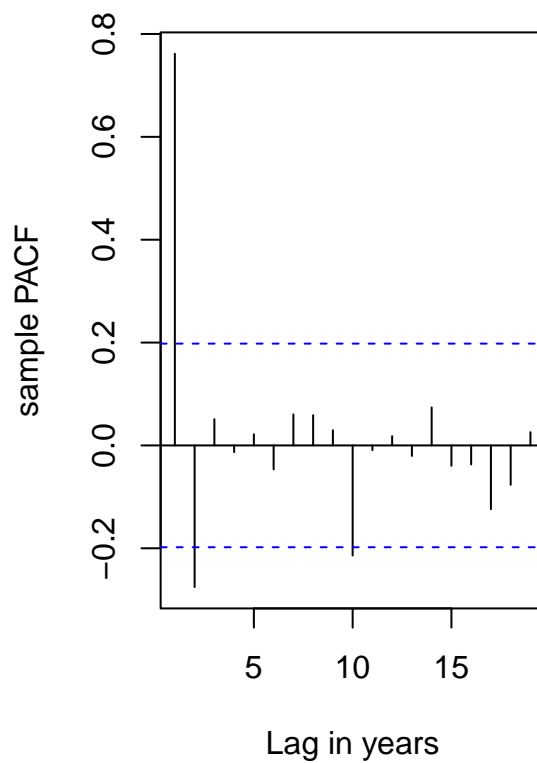
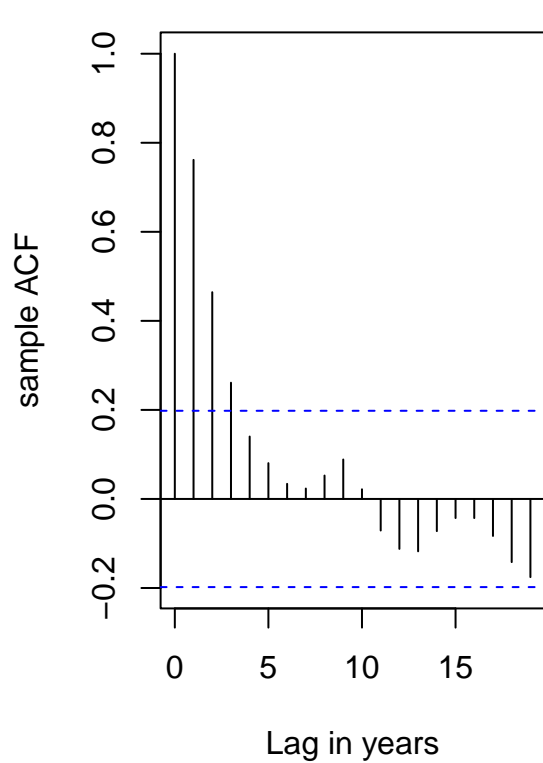
```



```

## ACF and PACF
par(mfrow = c(1, 2), mar = c(4, 4, 1, 1))
acf(deTrend, xlab="Lag in years", ylab = "sample ACF", main = "")
pacf(deTrend, xlab="Lag in years", ylab = "sample PACF", main = "")

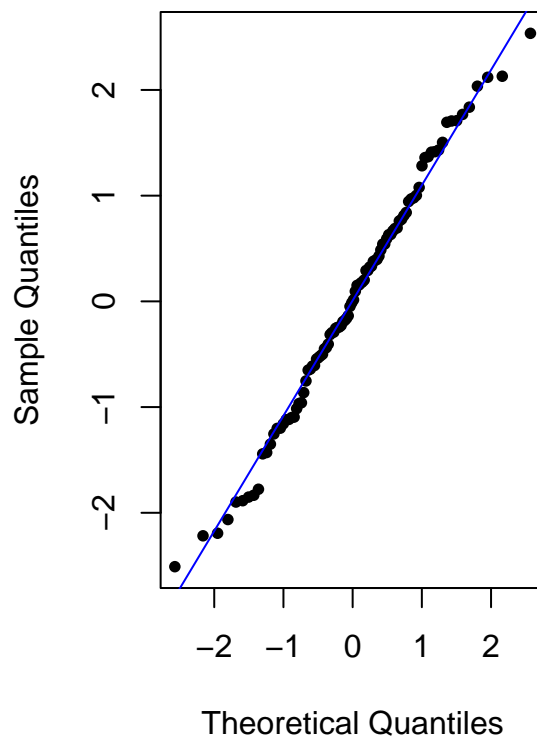
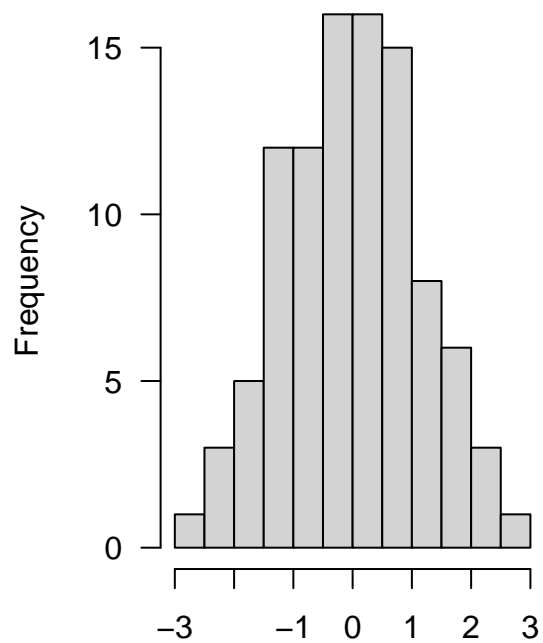
```



```
# Histogram and QQ plot
```

```
hist(deTrend, main = "", xlab = "", las = 1)
```

```
qqnorm(deTrend, main = "", pch = 16, cex = 0.8); qqline(deTrend, col = "blue")
```

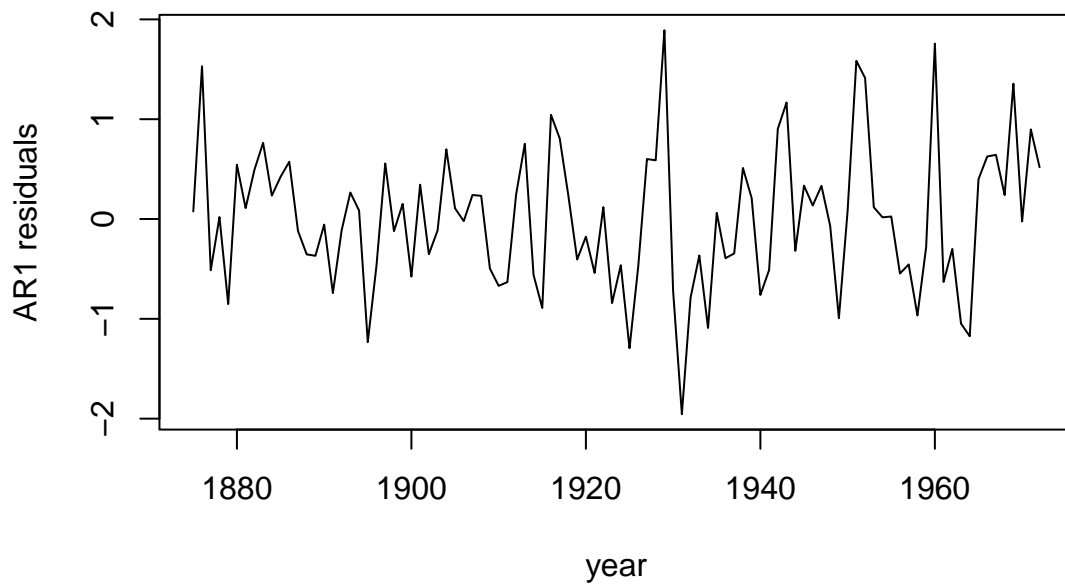


Model Selection/Fitting

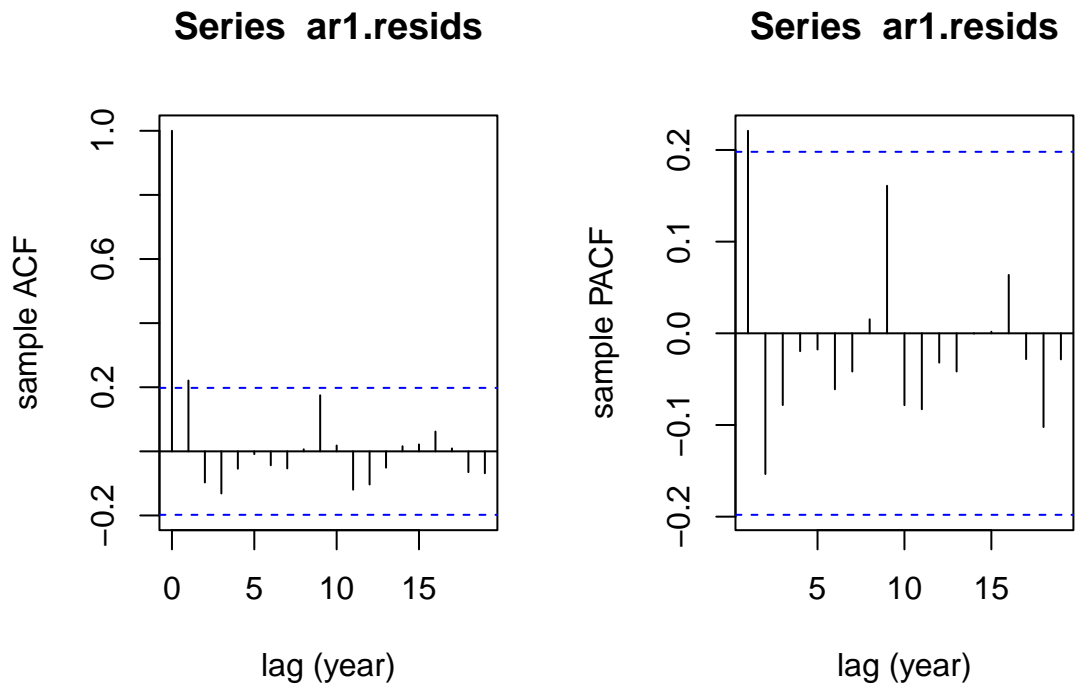
```
## AR(1)
(ar1.model <- arima(deTrend, order = c(1, 0, 0)))

##
## Call:
## arima(x = deTrend, order = c(1, 0, 0))
##
## Coefficients:
##          ar1  intercept
##       0.7829    0.0797
## s.e.  0.0634    0.3178
##
## sigma^2 estimated as 0.4972:  log likelihood = -105.29,  aic = 216.58

ar1.resids <- resid(ar1.model)
plot(1875:1972, ar1.resids, type = "l", xlab = "year", ylab = "AR1 residuals")
```



```
## Sample ACF and PACF of the residuals
par(mfrow = c(1, 2))
acf(ar1.resids, ylab = "sample ACF", xlab = "lag (year)")
pacf(ar1.resids, ylab = "sample PACF", xlab = "lag (year)")
```



```
## Normal Q-Q plot for the residuals
qqnorm(ar1.resids, main = ""); qqline(ar1.resids, col = "blue")
## Test for time dependence for the residuals
Box.test(ar1.resids, type = "Ljung-Box")
```

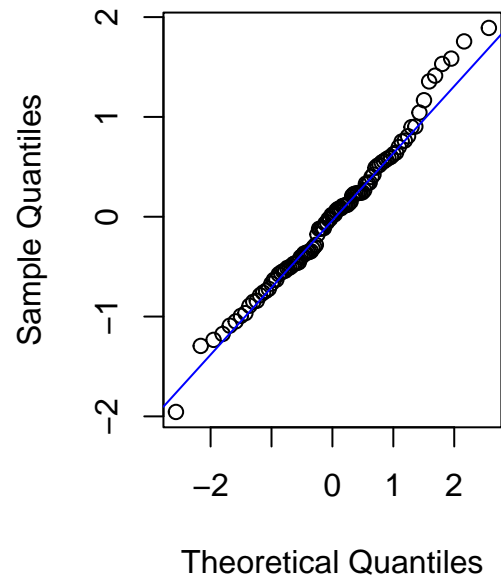
```
##
## Box-Ljung test
##
## data: ar1.resids
## X-squared = 4.93, df = 1, p-value = 0.02639
```

```
## AR(2)
(ar2.model <- arima(deTrend, order = c(2, 0, 0)))
```

```
##
## Call:
## arima(x = deTrend, order = c(2, 0, 0))
##
## Coefficients:
##          ar1          ar2 intercept
##          1.0047         -0.2919          0.0196
## s.e.  0.0977      0.1004      0.2351
##
## sigma^2 estimated as 0.4571: log likelihood = -101.25, aic = 210.5
```

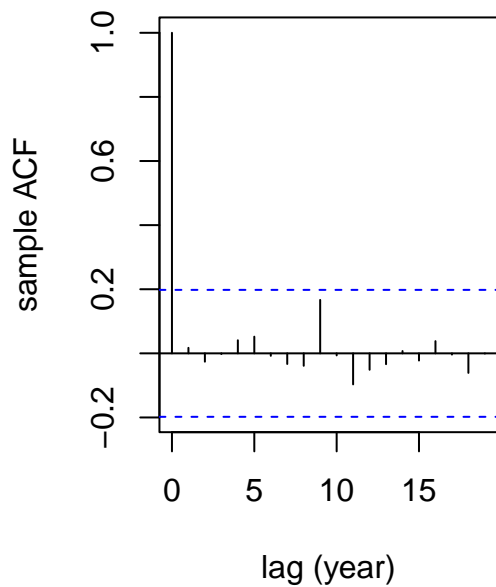
```
## calculate the residuals
ar2.resids <- resid(ar2.model)
```

```
## Sample ACF and PACF of the residuals  
par(mfrow = c(1, 2))
```

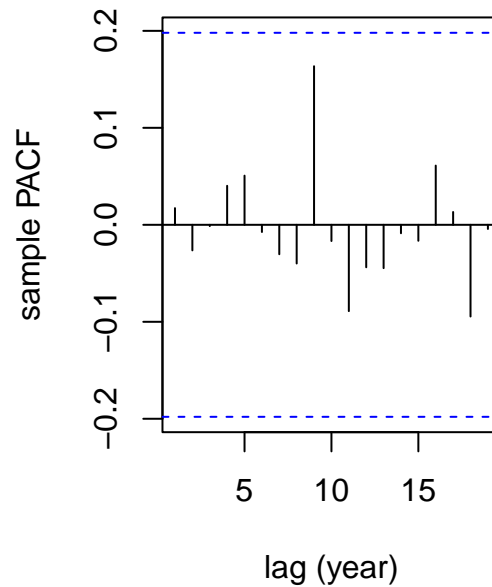


```
acf(ar2.resids, ylab = "sample ACF", xlab = "lag (year)")  
pacf(ar2.resids, ylab = "sample PACF", xlab = "lag (year)")
```

Series ar2.resids



Series ar2.resids



```
## Test for time dependence for the residuals
```

```
Box.test(ar2.resids, type = "Ljung-Box")
```

```
##
```

```
## Box-Ljung test
```

```
##
```

```
## data: ar2.resids
```

```
## X-squared = 0.029966, df = 1, p-value = 0.8626
```

```
## Normal Q-Q plot for the residuals
```

```
qqnorm(ar2.resids, main = ""); qqline(ar2.resids, col = "blue")
```

```
## Fit the ARMA(2, 1) model
```

```
(arma21.model <- arima(deTrend, order = c(2, 0, 1)))
```

```
##
```

```
## Call:
```

```
## arima(x = deTrend, order = c(2, 0, 1))
```

```
##
```

```
## Coefficients:
```

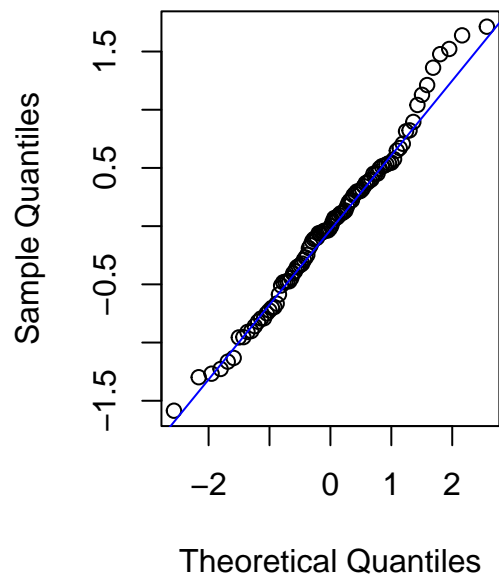
```
##          ar1          ar2          ma1  intercept
##          0.8374        -0.1622        0.1846         0.0245
```

```
## s.e.    0.3180         0.2621        0.3180         0.2452
```

```
##
```

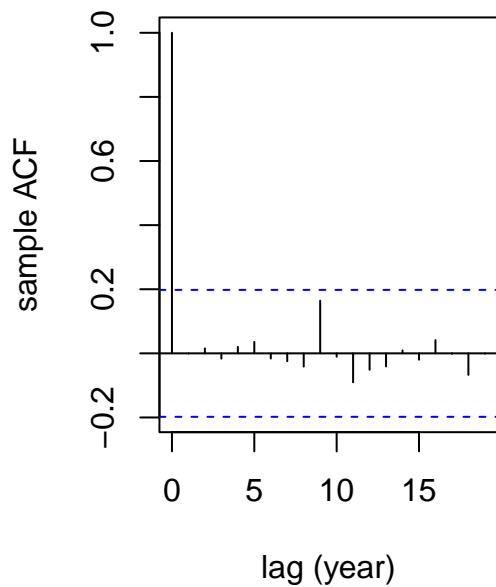
```
## sigma^2 estimated as 0.4556: log likelihood = -101.09, aic = 212.18
```

```
## calculate the residuals
arma21.resids <- resid(arma21.model)
## Sample ACF and PACF of the residuals
par(mfrow=c(1,2))
```

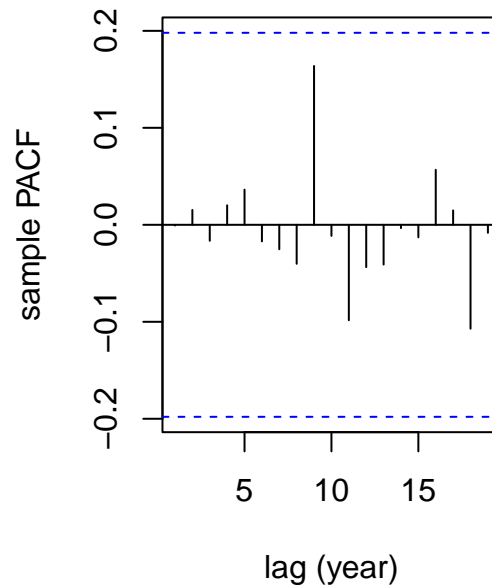


```
acf(arma21.resids, ylab = "sample ACF", xlab = "lag (year)")
pacf(arma21.resids, ylab = "sample PACF", xlab = "lag (year)")
```

Series arma21.resids



Series arma21.resids



```
## Normal Q-Q plot for the residuals
qqnorm(arma21.resids, main = ""); qqline(arma21.resids, col = "blue")
## Test
Box.test(arma21.resids, type = "Ljung-Box")
```

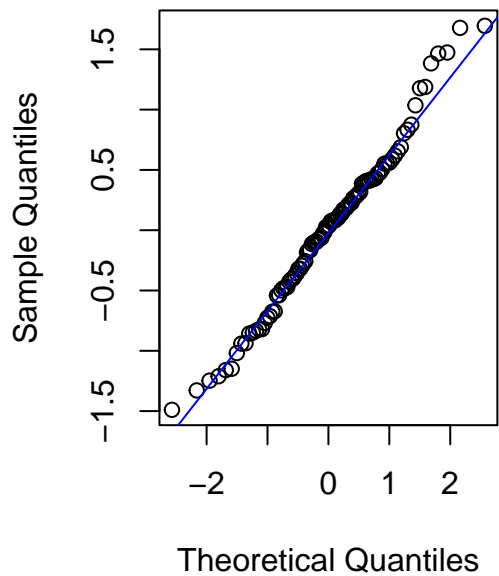
```
##
## Box-Ljung test
##
## data: arma21.resids
## X-squared = 5.5105e-05, df = 1, p-value = 0.9941
```

```
# Model selection using AIC
AIC(ar1.model); AIC(ar2.model); AIC(arma21.model)
```

```
## [1] 216.5835
```

```
## [1] 210.5032
```

```
## [1] 212.1784
```

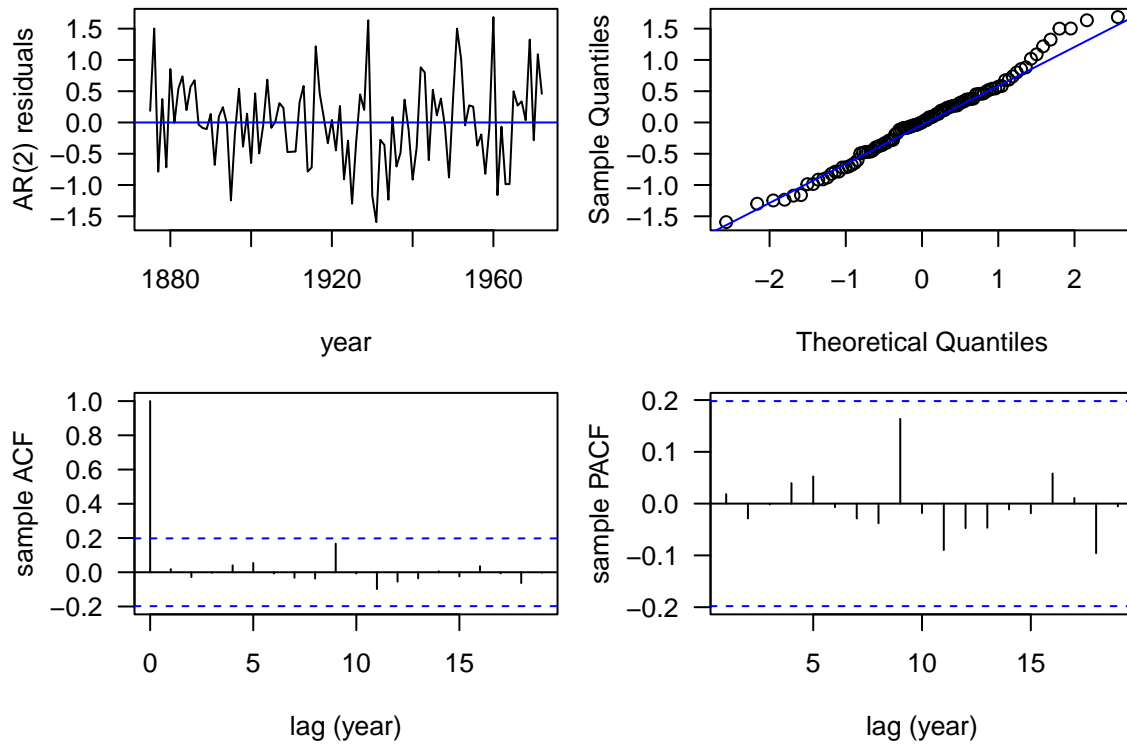



AR(2) Fitting and Forecasting

```
library(forecast)
(fit <- Arima(LakeHuron, order = c(2, 0, 0), include.drift = T))
```

```
## Series: LakeHuron
## ARIMA(2,0,0) with drift
##
## Coefficients:
##          ar1      ar2  intercept      drift
##          1.0048 -0.2913   580.0915  -0.0216
## s.e.    0.0976   0.1004     0.4636   0.0081
##
## sigma^2 = 0.476:  log likelihood = -101.2
## AIC=212.4   AICc=213.05   BIC=225.32
```

```
par(mfrow = c(2, 2), mar = c(4.1, 4, 1, 0.8), las = 1)
res <- fit$residuals
plot(res, type = "l", xlab = "year", ylab = "AR(2) residuals", las = 1)
abline(h = 0, col = "blue")
qqnorm(res, main = ""); qqline(res, col = "blue")
acf(res, ylab = "sample ACF", xlab = "lag (year)")
pacf(res, ylab = "sample PACF", xlab = "lag (year)")
```



```
# 10-year-ahead forecasts
autoplot(forecast(fit, h = 10, level = c(50, 95)))
```

