

# Australian beer production forecasting

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## Case Study

Analyze the historical beer production data and use time series forecasting techniques to forecast future beer production.

## load packages

```
library(fpp2)
library(astsa)
library(DT)
library(dygraphs)
```

## load data

```
beer <- read.csv("data/monthly-beer-production-australia.csv")
head(beer)
```

```
##      Month Monthly.beer.production.in.Australia
## 1 1956-01                                93.2
## 2 1956-02                                96.0
## 3 1956-03                                95.2
## 4 1956-04                                77.1
## 5 1956-05                                70.9
## 6 1956-06                                64.8
```

## basic analysis

```
tail(beer)
```

```
##      Month Monthly.beer.production.in.Australia
## 471 1995-03                                152
## 472 1995-04                                127
## 473 1995-05                                151
## 474 1995-06                                130
## 475 1995-07                                119
## 476 1995-08                                153
```

```
summary(beer)
```

```
##      Month      Monthly.beer.production.in.Australia
## 1956-01: 1      Min.      : 64.8
## 1956-02: 1      1st Qu.:112.9
## 1956-03: 1      Median :139.2
```

```
## 1956-04: 1 Mean :136.4
## 1956-05: 1 3rd Qu.:158.8
## 1956-06: 1 Max. :217.8
## (Other):470
```

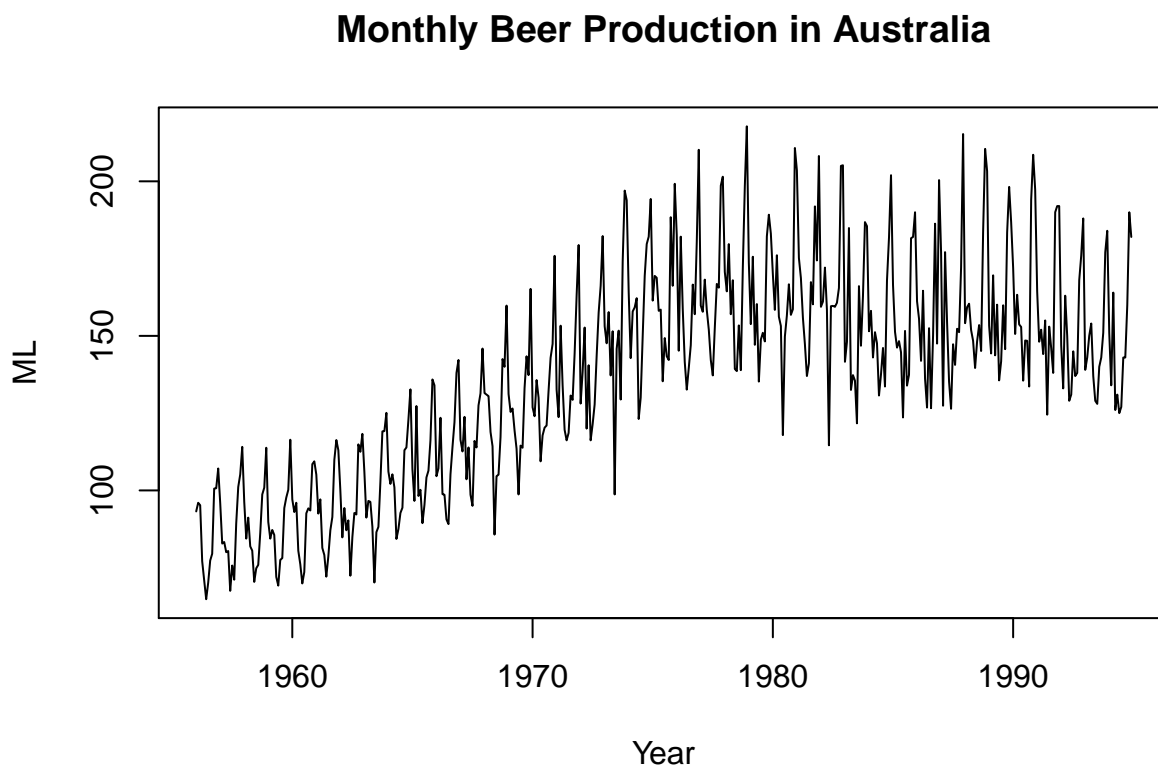
```
beer.ts <- ts(beer, frequency = 12, start = c(1956,1), end = c(1994,12))
```

```
head(beer.ts)
```

```
##      Month Monthly.beer.production.in.Australia
## Jan 1956      1                      93.2
## Feb 1956      2                      96.0
## Mar 1956      3                      95.2
## Apr 1956      4                      77.1
## May 1956      5                      70.9
## Jun 1956      6                      64.8
```

## time series plots

```
plot.ts(beer.ts[,2], main = "Monthly Beer Production in Australia", xlab = "Year", ylab = "ML")
```

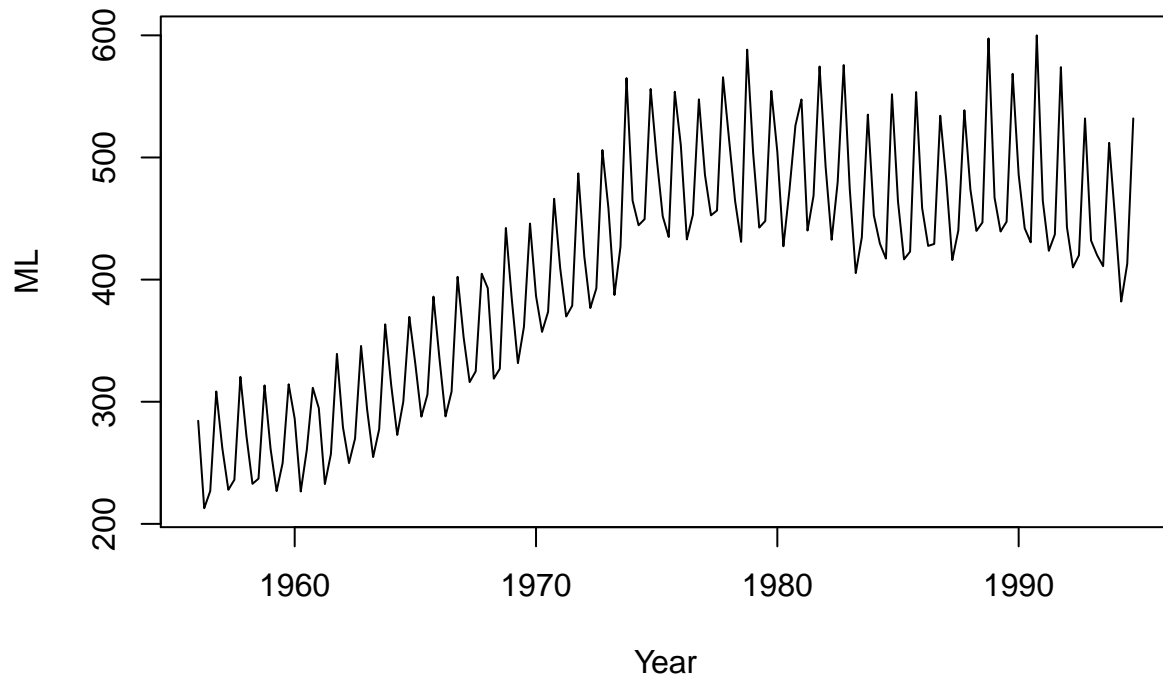


```
# aggregated on quater level
```

```
beer.ts.qtr <- aggregate(beer.ts, nfrequency=4)
```

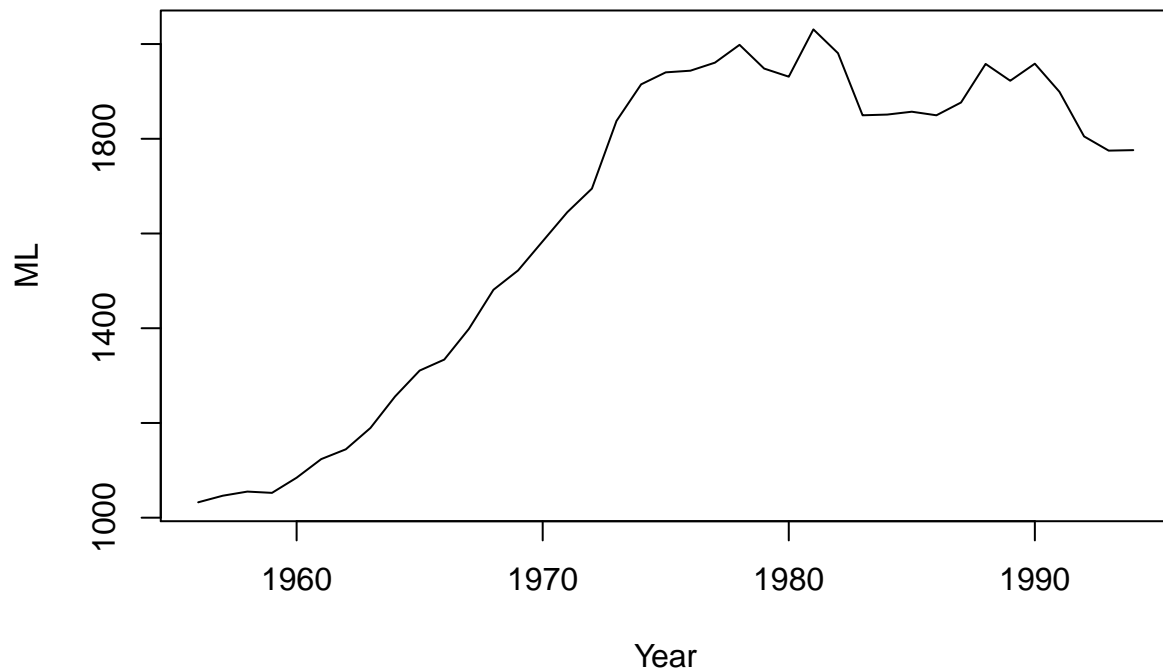
```
plot.ts(beer.ts.qtr[,2], main = "Quarterly Beer Production in Australia", xlab = "Year", ylab = "ML")
```

## Quaterly Beer Production in Australia



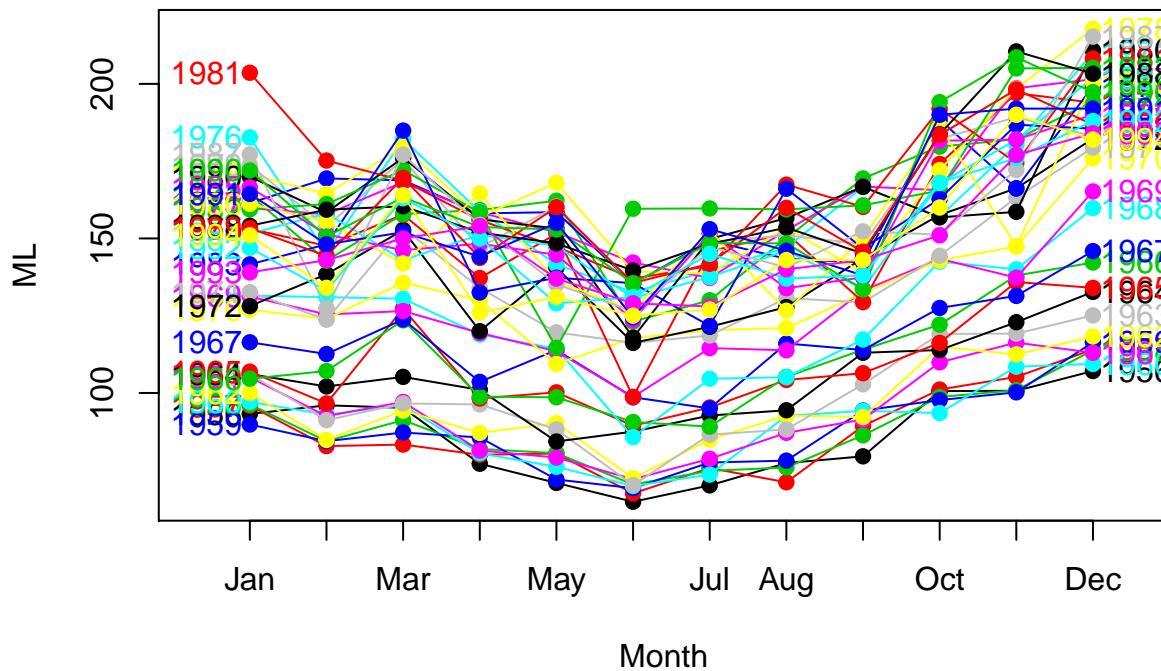
```
# aggregated on year level  
beer.ts.yr <- aggregate(beer.ts, nfrequency=1)  
plot.ts(beer.ts.yr[,2], main = "Yearly Beer Production in Australia", xlab = "Year", ylab = "ML")
```

## Yearly Beer Production in Australia



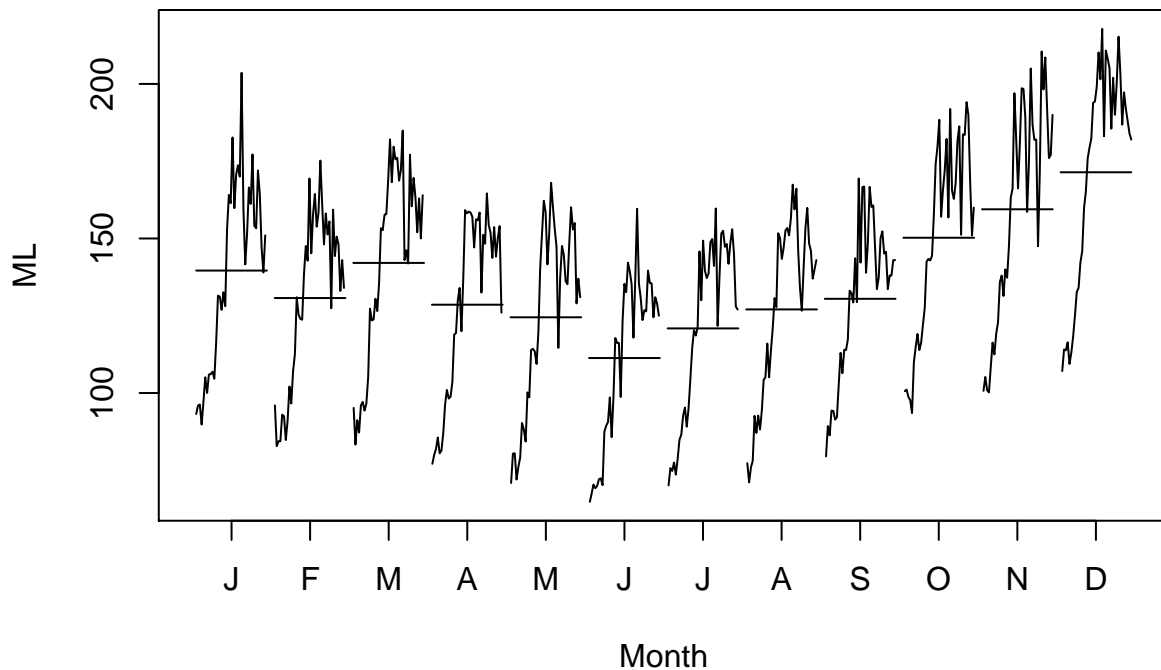
```
seasonplot(beer.ts[,2], year.labels = TRUE, year.labels.left=TRUE, col=1:40, pch=19, main = "Monthly Beer Production in Australia – seasonplot")
```

### Monthly Beer Production in Australia – seasonplot



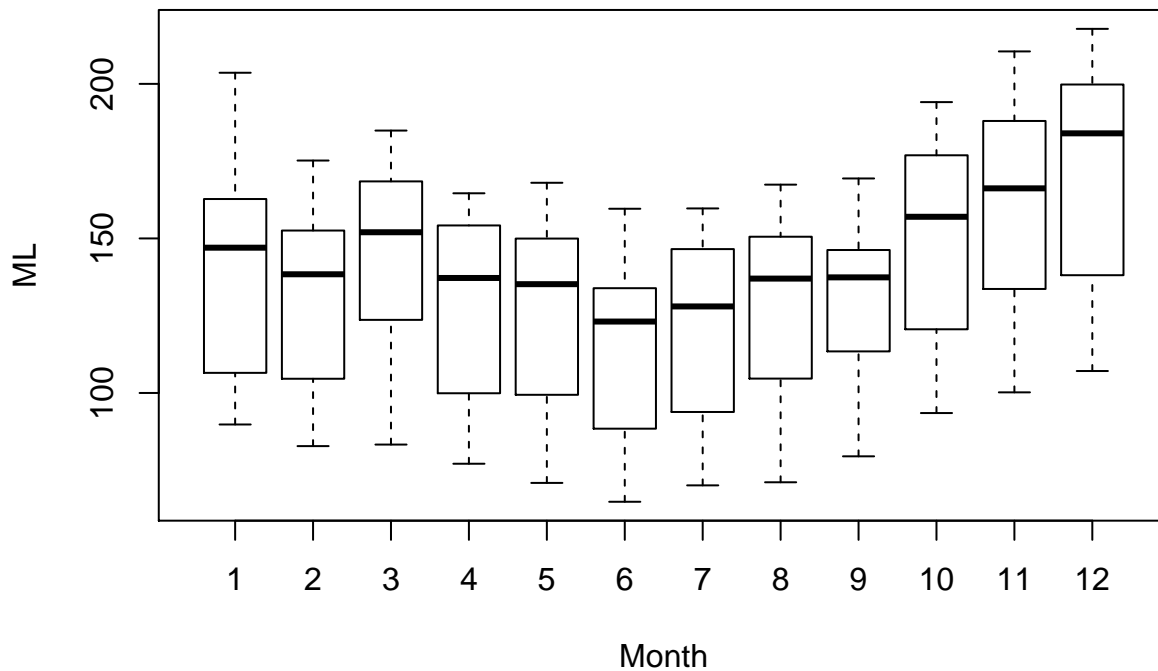
```
# monthly plot aggregated the data of all year for montly analysis
# each month plots show the variation for entire data in each month
monthplot(beer.ts[,2], main = "Monthly Beer Production in Australia - monthplot", xlab = "Month", ylab = "ML")
```

### Monthly Beer Production in Australia – monthplot



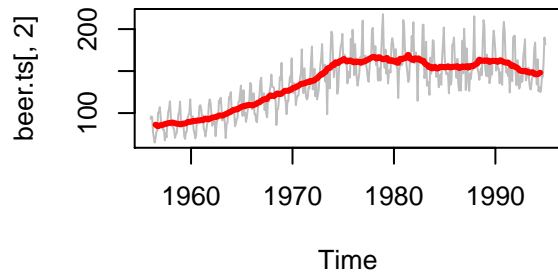
```
boxplot(beer.ts[,2] ~ cycle(beer.ts[,2]), xlab = "Month", ylab = "ML", main = "Monthly Beer Production in Australia")
```

## Monthly Beer Production in Australia – Boxplot

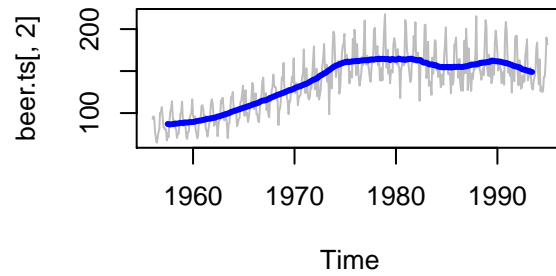


```
# moving average is useful when we need to analyse trend for the underlying data
# here, we see the moving average for 1 year, 3 year 5 year and 10 year
# Note : If there is not trend, average is good enough for the analysis
par(mfrow = c(2,2))
plot(beer.ts[,2], col="gray", main = "1 Year Moving Average Smoothing")
lines(ma(beer.ts[,2], order = 12), col = "red", lwd=3)
plot(beer.ts[,2], col="gray", main = "3 Year Moving Average Smoothing")
lines(ma(beer.ts[,2], order = 36), col = "blue", lwd=3)
plot(beer.ts[,2], col="gray", main = "5 Year Moving Average Smoothing")
lines(ma(beer.ts[,2], order = 60), col = "green", lwd=3)
plot(beer.ts[,2], col="gray", main = "10 Year Moving Average Smoothing")
lines(ma(beer.ts[,2], order = 120), col = "yellow4", lwd=3)
```

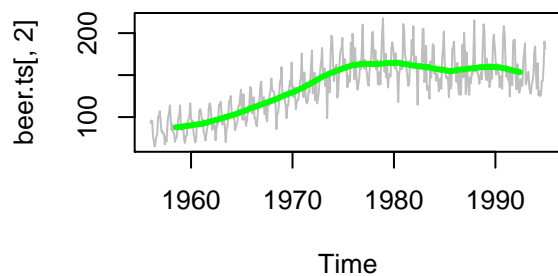
**1 Year Moving Average Smoothing**



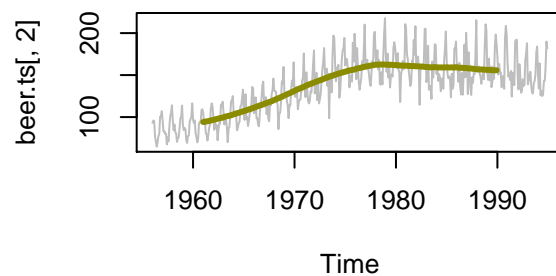
**3 Year Moving Average Smoothing**



**5 Year Moving Average Smoothing**



**10 Year Moving Average Smoothing**

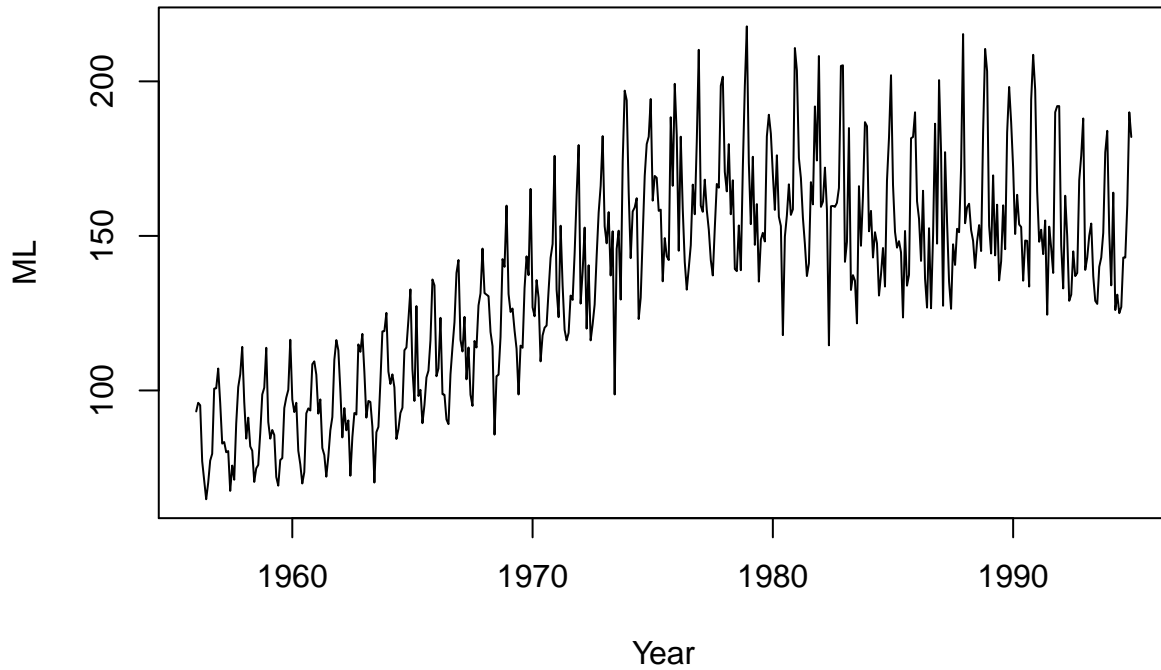


# data transformation and adjustments

calendar adjustment

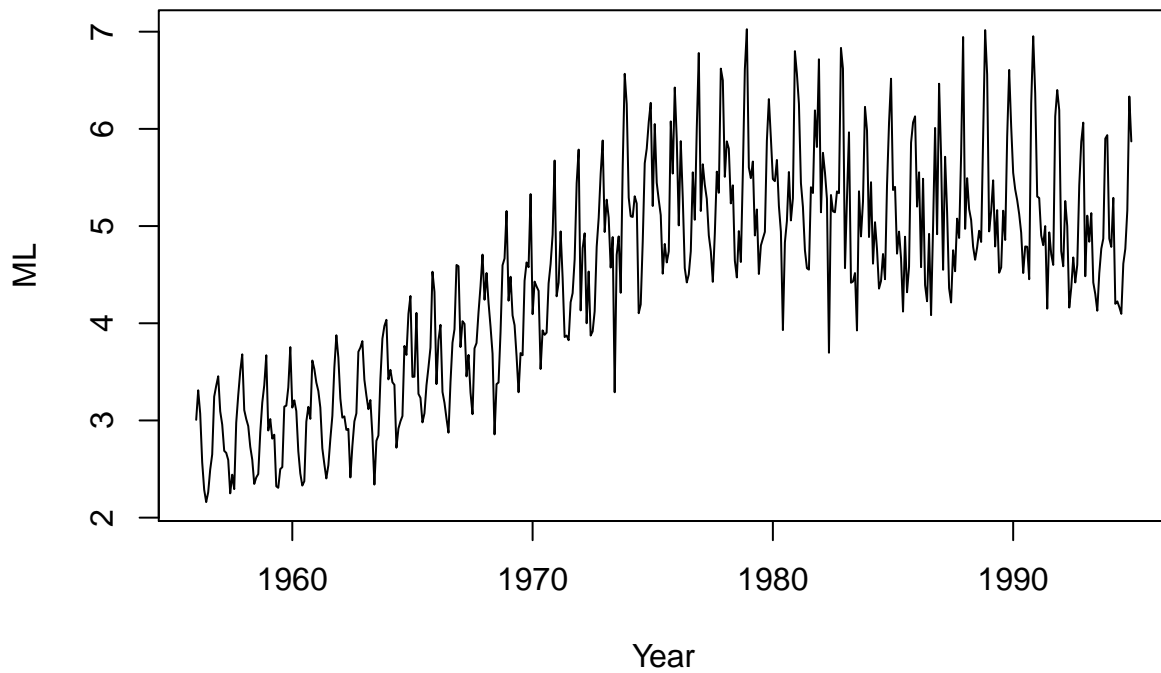
```
plot.ts(beer.ts[,2], main = "Monthly Beer Production in Australia", xlab = "Year", ylab = "ML")
```

## Monthly Beer Production in Australia



```
plot.ts(beer.ts[,2]/monthdays(beer.ts[,2]), main = "Monthly Beer Production in Australia - Adjusted By Calendar Days")
```

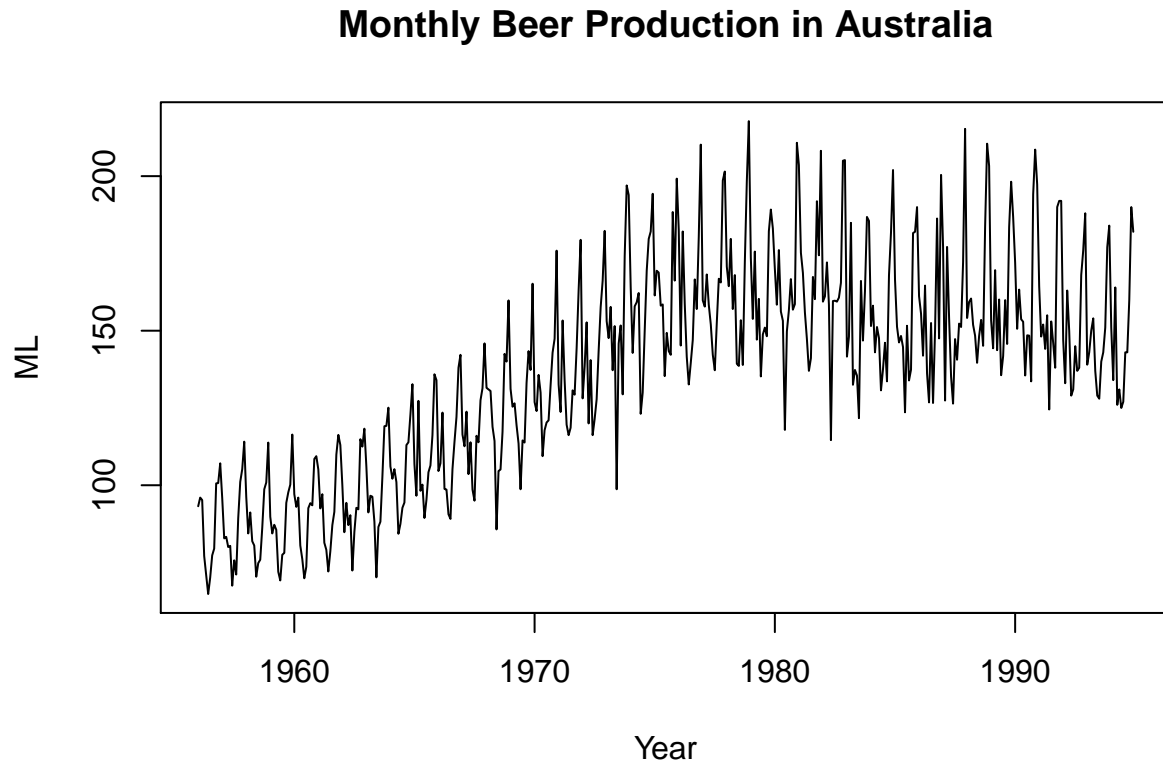
## Monthly Beer Production in Australia – Adjusted By Calendar Days



population adjustment

logarithmic transformation

```
plot.ts(beer.ts[,2], main = "Monthly Beer Production in Australia", xlab = "Year", ylab = "ML")
```



```
plot.ts(log(beer.ts[,2]), main = "Log Transformed Monthly Beer Production in Australia", xlab = "Year")
```



## Log Transformed Monthly Beer Production in Australia

