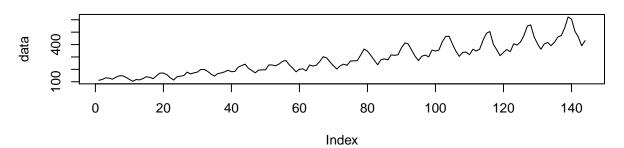
BoxCox Transformation

Here we illustrate how to apply Box-Cox Transformation. The data set is a airpassenger data in US.

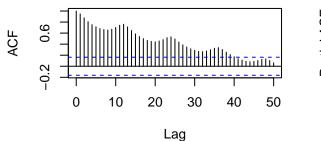
```
data = scan("airpass.txt");
layout(matrix(c(1,1,2,3), 2, 2, byrow = TRUE))
plot(data, type="l")
title("Airpass")
acf(data, lag=50);
pacf(data, lag=50);
```

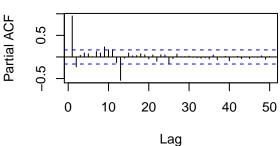
Airpass



Series data

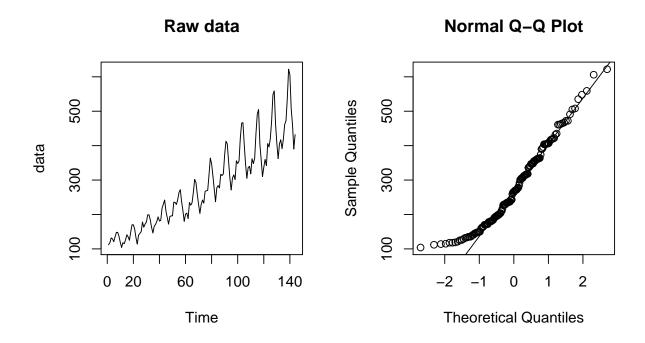
Series data



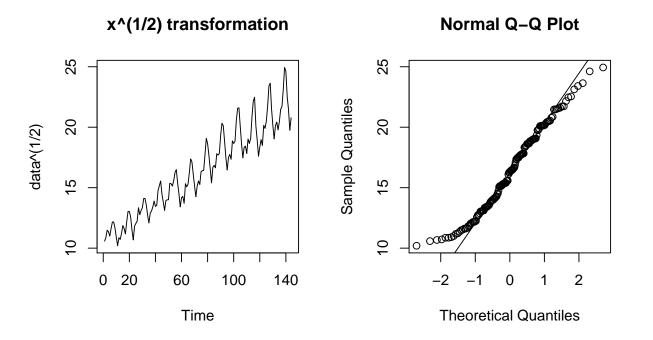


Albeit clear trend and seasonal component, we can observe some increasing variance, so we may need to do some transformation. The first method is ad-hoc graphical method. If we assume Gaussianity on the innovations, it is equivalent to find the transformation makes the data closest to Normal. Hence, we can use QQplot to detect it.

```
par(mfrow=c(1,2))
plot.ts(data); title("Raw data")
qqnorm(data);qqline(data);
```

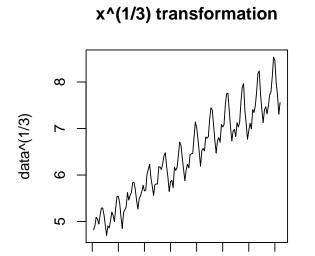


```
par(mfrow=c(1,2))
par(mfrow=c(1,2))
plot.ts(data^(1/2)); title("x^(1/2) transformation")
qqnorm(data^(1/2)); qqline(data^(1/2));
```

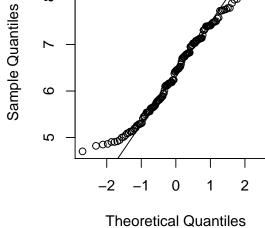


```
plot.ts(data^(1/3)); title("x^(1/3) transformation")
qqnorm(data^(1/3)); qqline(data^(1/3));
```

 ∞



Normal Q-Q Plot



par(mfrow=c(1,2))
plot.ts(log(data)); title("Log-transformation")
qqnorm(log(data)); qqline(log(data));

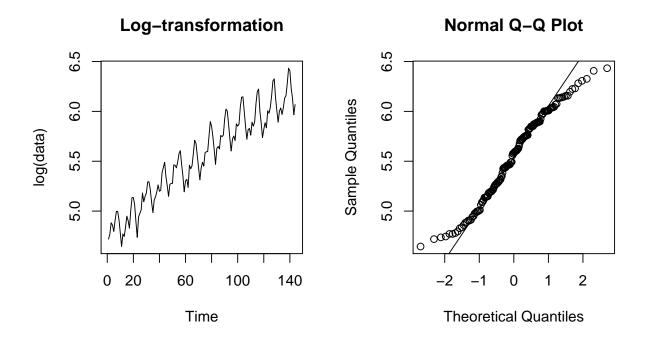
60

Time

0 20

100

140



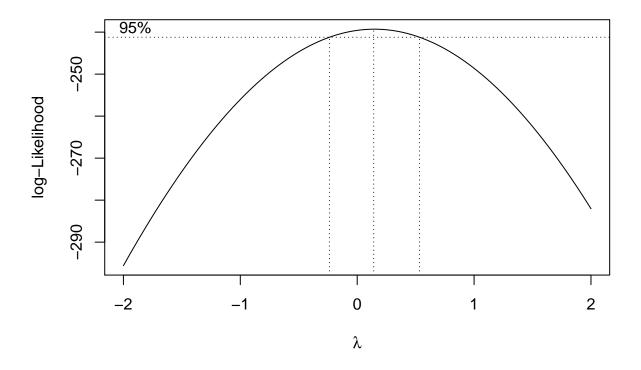
The second method is to estimate lambda by MLE. It is implemented in MASS() library.

```
library(MASS)
library(itsmr)

##
## Attaching package: 'itsmr'

## The following object is masked from 'package:MASS':
##
## deaths

x = 1:length(data)
fit = boxcox(data~1, plotit=TRUE);
```

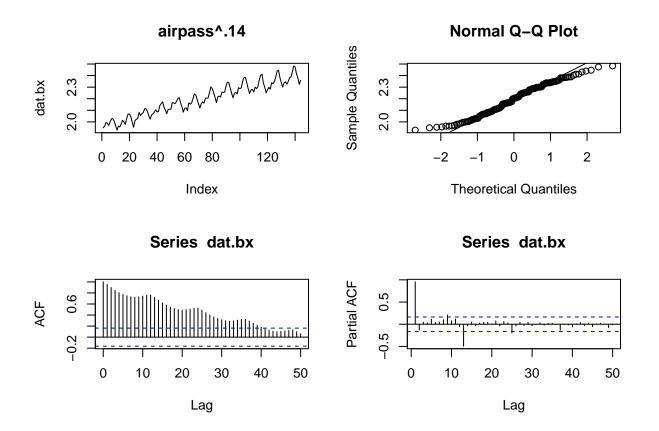


```
lambda = fit$x[which.max(fit$y)]
lambda
```

[1] 0.1414141

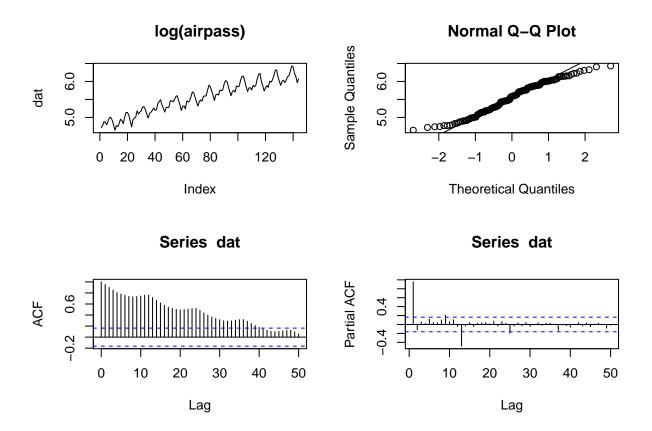
It suggests to take $data^{.14}$ transformation. Then, it gives the following results.

```
dat.bx = data^lambda;
layout(matrix(c(1,2,3,4), 2, 2, byrow = TRUE))
plot(dat.bx, type="l")
title("airpass^.14")
qqnorm(dat.bx); qqline(dat.bx);
acf(dat.bx, lag=50);
pacf(dat.bx, lag=50);
```



Box-Cox transforantion makes the variance constant over time. However, in practice, taking the power transformation makes the model interpretation hard. In my own experiences, we prefer to take either $\log(y)$ or \sqrt{y} transformation if needed.

```
dat = log(data);
layout(matrix(c(1,2,3,4), 2, 2, byrow = TRUE))
plot(dat, type="l")
title("log(airpass)")
qqnorm(dat); qqline(dat);
acf(dat, lag=50);
pacf(dat, lag=50);
```



Practice: See HW problem with wind data.