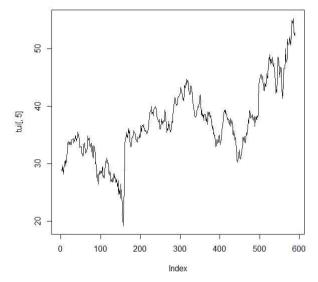
데이터 사이언스 과제8

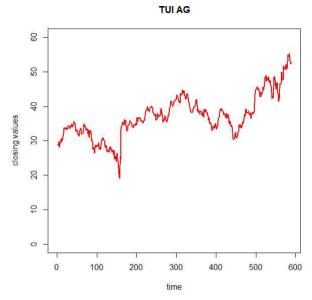
< 1. Getting started >

```
> tui <- read.csv("C:/Users/김황규/Documents/tui.csv", header=T, dec=",",
sep=";")
```

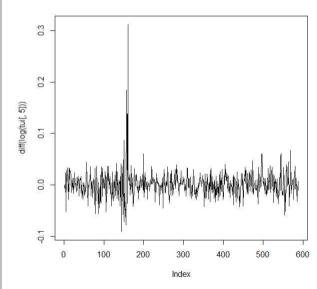
> plot(tui[,5],type="l")



- > plot(tui[,5], type="l",
 + lwd=2, col="red", xlab="time", ylab="closing values",
 + main="TUI AG", ylim=c(0,60))

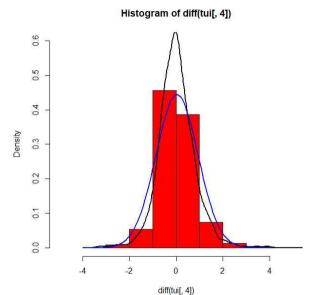


> plot(diff(log(tui[,5])),type="l")



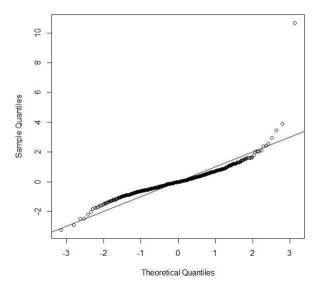
- > hist(diff(tui[,4]),prob=T,ylim=c(0,0.6),xlim=c(-5,5),col="red")
 > lines(density(diff(tui[,4])),lwd=2)
- > mu<-mean(diff(tui[,4]))
 > sigma<-sd(diff(tui[,4]))</pre>

- > x < -seq(-4.4, length = 100)
- > y<-dnorm(x,mu,sigma)
- > lines(x,y,lwd=2,col="blue")



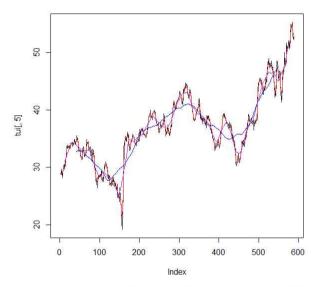
- > qqnorm(diff(tui[,4]))
 > abline(0,1)

Normal Q-Q Plot

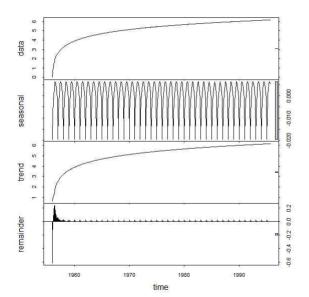


> x < -diff(log(tui[,5]))

```
> ks.test(x,"pnorm",mean(x),sd(x))
          One-sample Kolmogorov-Smirnov test
data: x
D = 0.098003, p-value = 2.487e-05
alternative hypothesis: two-sided
경고메시지(들):
In ks.test(x, "pnorm", mean(x), sd(x)) :
  Kolmogorov-Smirnov 테스트를 이용할 때는 ties가 있으면 안됩니다
>
> shapiro.test(x)
         Shapiro-Wilk normality test
data: x
W = 0.80962, p-value < 2.2e-16
< 2. Simple Component Analysis >
> #library(ts)
> #install.packages("timeSeries") 로 바뀜
> library(timeSeries)
필요한 패키지를 로딩중입니다: timeDate
  plot(tui[,5],type="l")
> tui.1 <- filter(tui[,5],filter=rep(1/5,5))
> tui.2 <- filter(tui[,5],filter=rep(1/25,25))
> tui.3 <- filter(tui[,5],filter=rep(1/81,81))
> lines(tui.1,col="red")
> lines(tui.1,col="red")
> lines(tui.2,col="purple")
> lines(tui.3,col="blue")
```



- > beer<-read.csv("beer.csv",header=T,dec=",",sep=";")
 > beer<-ts(beer[,1],start=1956,freq=12)
 > plot(stl(log(beer),s.window="periodic"))



- > lbeer<-log(beer) > t<-seq(1956,1995.2,length=length(beer)) > t2<-t^2
- > plot(lbeer)

```
> lm(lbeer~t+t2)
```

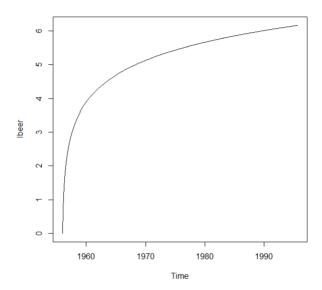
Call:

 $lm(formula = lbeer \sim t + t2)$

Coefficients:

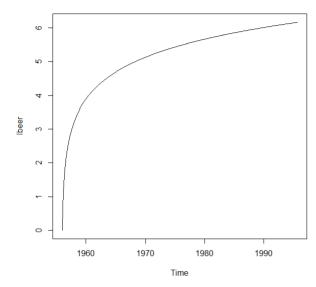
(Intercept) -1.239e+04 1.247e+01 -3.138e-03

> lines(lm(lbeer~t+t2)\$fit,col=2,lwd=2)



- > lbeer<-log(beer)
- > t<-seq(1956,1995.2,length=length(beer)) > t2<-t^2

- > sin.t<-sin(2*pi*t)
 > cos.t<-cos(2*pi*t)
 > plot(lbeer)
 > lines(lm(lbeer~t+t2+sin.t+cos.t)\$fit,col=4)



> summary(lm(lbeer~t+t2+sin.t+cos.t))

Call:

 $lm(formula = lbeer \sim t + t2 + sin.t + cos.t)$

Residuals:

Min 1Q Median 3Q Max -2.88651 -0.13678 -0.00285 0.19673 0.34909

Coefficients:

Estimate Std. Error t value Pr(>|t|) (Intercept) -1.238e+04 4.614e+02 -26.835 <2e-16 *** 1.246e+01 4.671e-01 26.685 <2e-16 *** t2 -3.135e-03 1.182e-04 -26.524 <2e-16 *** sin.t -1.770e-02 1.924e-02 -0.920 0.358 -1.035e-02 1.922e-02 -0.538 0.591 cos.t Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1

Residual standard error: 0.2969 on 472 degrees of freedom Multiple R-squared: 0.9081, Adjusted R-squared: 0.9073 F-statistic: 1166 on 4 and 472 DF, p-value: < 2.2e-16

< 3. Exponential Smoothing >

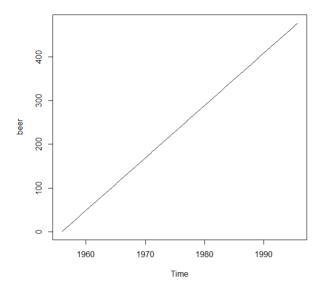
```
> beer<-read.csv("beer.csv",header=T,dec=",",sep=";")
> beer<-ts(beer[,1],start=1956,freq=12)
>
```

```
> HoltWinters(beer)
Holt-Winters exponential smoothing with trend and additive seasonal
component.
Call:
HoltWinters(x = beer)
Smoothing parameters:
alpha: 1
beta: 0
gamma: 0
Coefficients:
    4.770000e+02
b
    1.000000e+00
    9.621933e-16
s1
   -8.141636e-16
   -8.141636e-16
    9.621933e-16
   -8.141636e-16
   -8.141636e-16
    9.621933e-16
s7
    2.738550e-15
   -8.141636e-16
s10 -8.141636e-16
s11 7.401487e-17
s12 -8.141636e-16
```

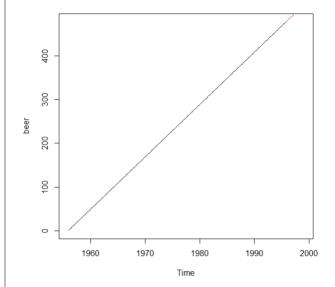
> lines(HoltWinters(beer)\$fitted,col="red")

Error in xy.coords(x, y): 'x' and 'y' lengths differ

> plot(beer)



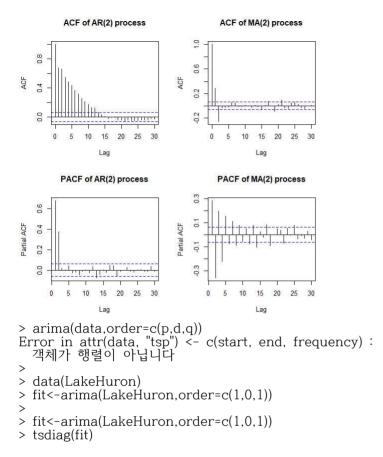
- > beer.hw<-HoltWinters(beer)
- > predict(beer.hw,n.ahead=12) Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec 1995 478 479 480 1996 481 482 483 484 485 486 487 488 489
- > plot(beer,xlim=c(1956,1999))
 > lines(predict(beer.hw,n.ahead=48),col=2)

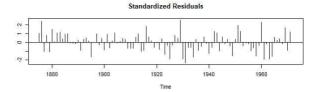


< 4. ARIMA-Models >

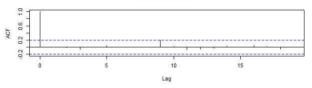
- > sim.ar<-arima.sim(list(ar=c(0.4,0.4)),n=1000) > sim.ma<-arima.sim(list(ma=c(0.6,-0.4)),n=1000)

- > par(mfrow=c(2,2))
 > acf(sim.ar,main="ACF of AR(2) process")
 > acf(sim.ma,main="ACF of MA(2) process")
 > pacf(sim.ar,main="PACF of AR(2) process")
 > pacf(sim.ar,main="PACF of MA(2) process")
 > pacf(sim.ma,main="PACF of MA(2) process")

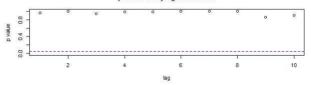




ACF of Residuals



p values for Ljung-Box statistic



> Box.test(fit\$residuals,lag=1)

Box-Pierce test

data: fit\$residuals X-squared = 0.0021379, df = 1, p-value = 0.9631

- > > fit<-arima(LakeHuron.order=c(1,0,1))</pre>
- > LH.pred<-predict(fit,n.ahead=8)
- > plot(LakeHuron,xlim=c(1875,1980),ylim=c(575,584))
 > LH.pred<-predict(fit,n.ahead=8)</pre>

- > lines(LH.pred\$pred,col="red")
 > lines(LH.pred\$pred+2*LH.pred\$se,col="red",lty=3)
 > lines(LH.pred\$pred-2*LH.pred\$se,col="red",lty=3)

