# lab8\_DerekSitu

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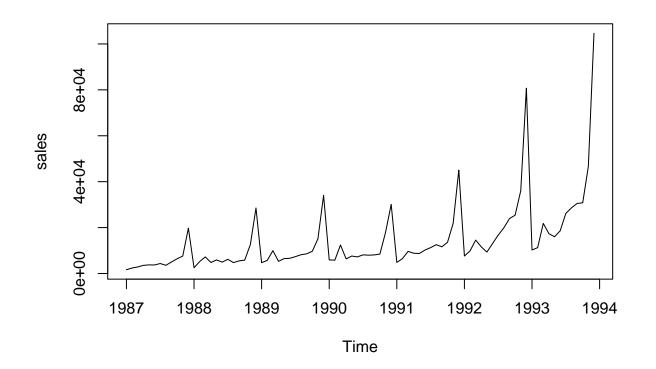
3/13/2022

## Data

```
dat <- read.delim("souvenir.txt", header = FALSE)
sales <- ts(dat$V1, start = c(1987, 1), frequency = 12)</pre>
```

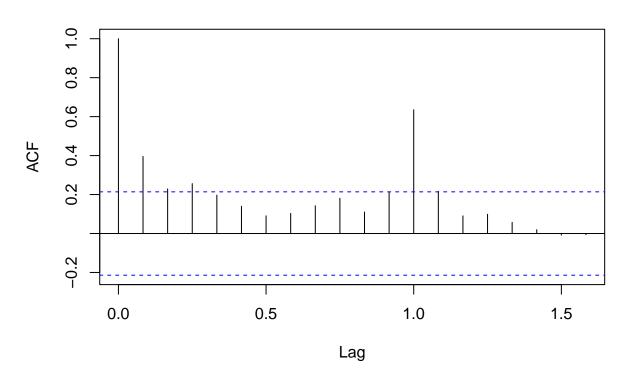
## Question 1

```
plot(sales)
```



acf(sales)

## Series sales



From the plot of the series, we notice a seasonal pattern, and we notice that the variation increases over time. This fits a multiplicative model. The acf also displays the seasonal pattern. There also may be a trend.

#### Question 2

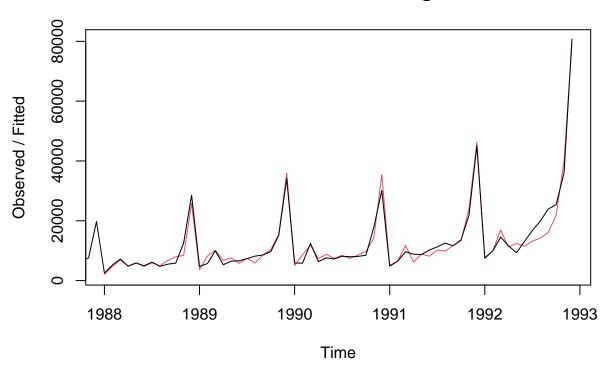
```
training <- window(sales, start = c(1987, 1), end = c(1992, 12))
hw_fit <- HoltWinters(training, seasonal = "multiplicative")
c(hw_fit$alpha, hw_fit$beta, hw_fit$gamma)</pre>
```

```
## alpha beta gamma
## 0.34698417 0.07501578 0.57114780
```

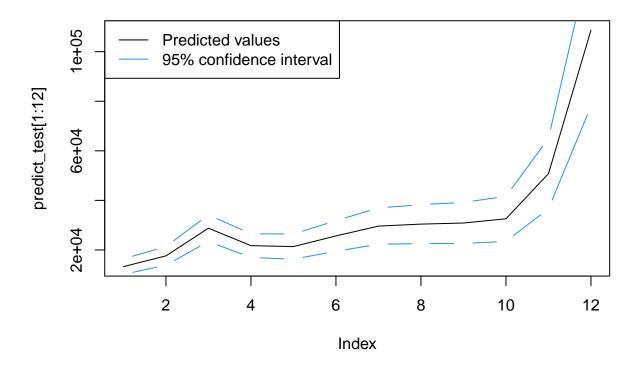
The parameters are  $\alpha = 0.347, \beta = 0.075, \gamma = 0.571, p = 12.$ 

```
plot(hw_fit)
```

## **Holt-Winters filtering**



## Question 3



```
predict_test[1:3] # predicted values for first 3 months
```

## [1] 13277.67 17609.17 28784.94

The forecast values for the first 3 months of 1993 are  $1.3277667 \times 10^4$ ,  $1.7609172 \times 10^4$ ,  $2.8784935 \times 10^4$ .

#### Question 4

Only the observed value for the first month of 1993 fits inside the 95% confidence interval.

## ${\bf Question}~{\bf 5}$

I would consider a log transformation since the series seems to be increasing exponentially. One useful property of logs is that the log of a product is the sum of the logs of the factors, so using this property we can change this multiplicative model into an additive model.