

# Time Series Analysis

## Basics of Time Series Analysis: Data Example

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Stationarity of ED Volume  
Time Series

# About This Lesson



# Residual Process

## ## Residual Process: Trend Removal

```
resid.1 = Volume.tr-vol.fit.gam
```

## ## Residual Process: Seasonality Removal

```
resid.2 = Volume.tr-vol.fit.lm.seastr.2
```

## ## Residual Process: Trend and Seasonality Removal

```
resid.3 = Volume.tr-vol.fit.gam.seastr.2
```

```
y.min = min(c(resid.1,resid.2,resid.3))
```

```
y.max = max(c(resid.1,resid.2,resid.3))
```

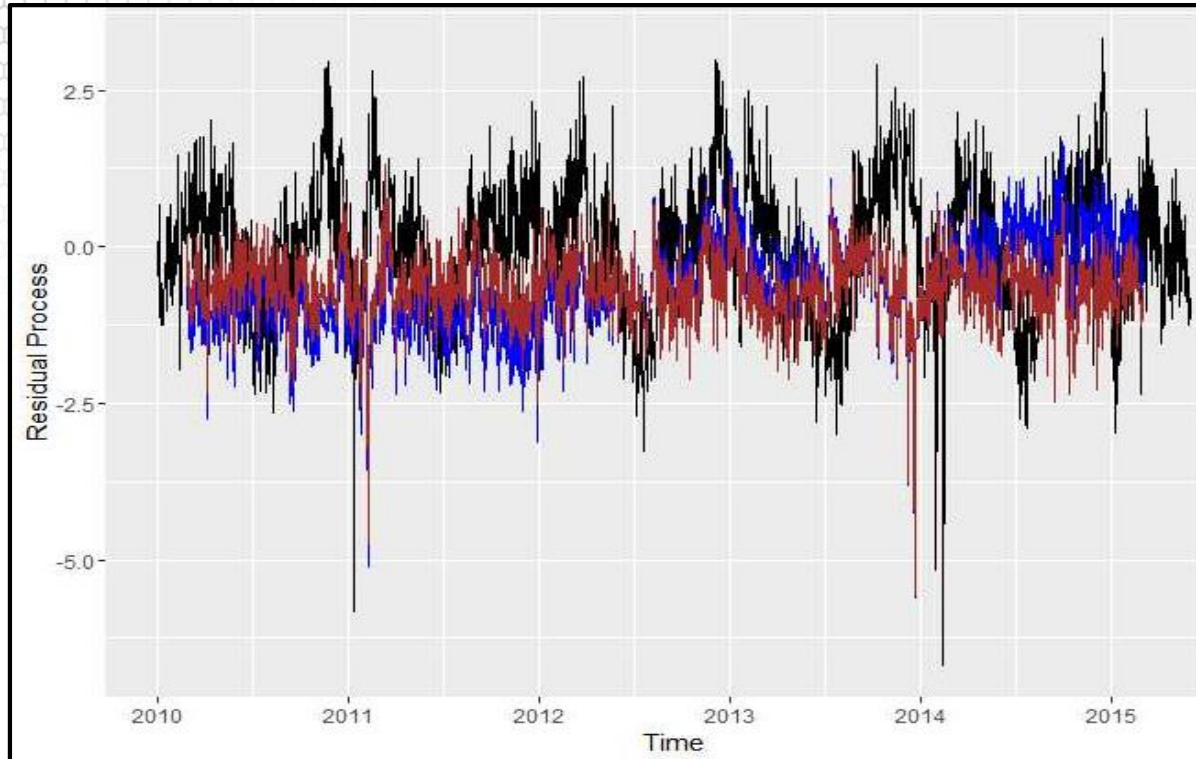
```
ggplot(edvoldata, aes(dates, resid.1), ymin=y.min, ymax=y.max) + geom_line() +
```

```
xlab("Time") + ylab("Residual Process")
```

```
lines(dates,resid.2,col="blue")
```

```
lines(dates,resid.3,col="brown")
```

# Residual Process

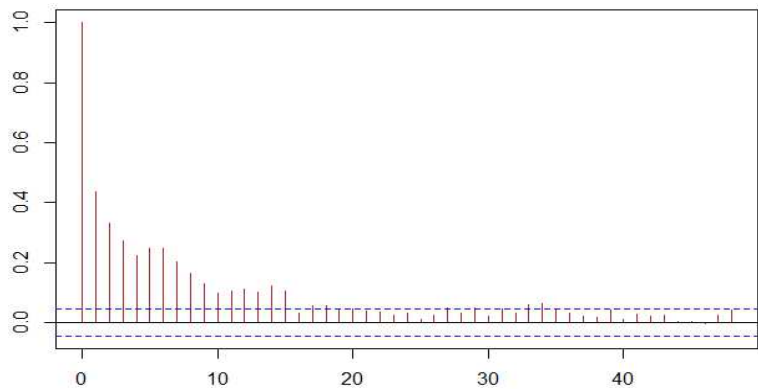
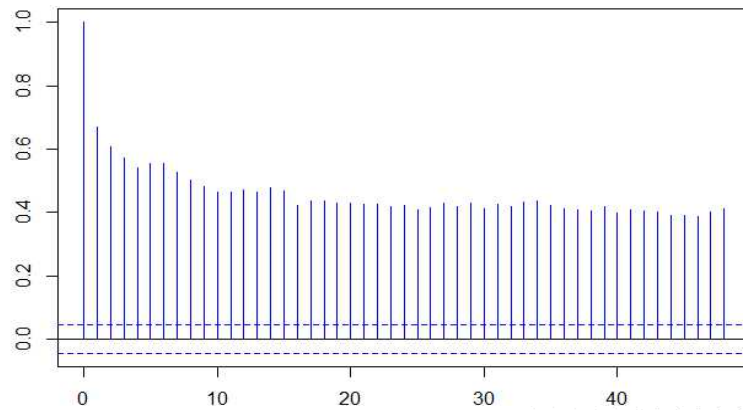
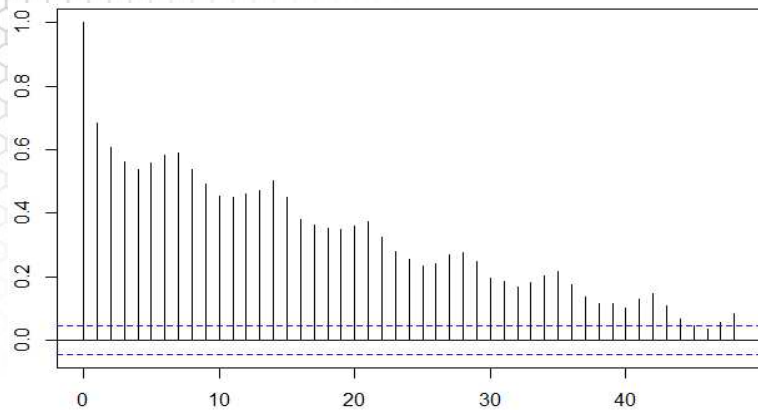


# Residual Process: ACF

## **## Compare Auto-correlation plots**

```
acf(resid.1,lag.max=12*4,main="")  
acf(resid.2,lag.max=12*4,main="",col="blue")  
acf(resid.3,lag.max=12*4,main="",col="brown")
```

# Residual Process: ACF



ACF is outside of the band for the first 15 lags, an indication of stationarity

# Findings

- There is a significant increasing trend in the Emergency Department (ED) patient volume over the past five years
- Seasonality is more complex; both monthly and day-of-the-week are statistically significant seasonality
- There are cyclical patterns that may not be fully captured by seasonality; other cyclical factors such as flu season or school season may explain the cyclical pattern

# Summary

