Introduction to Time Series Analysis OC Data Driven Insights

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Resources

- ► Forecasting: Principles and Practice by Rob J. Hyndman and George Athanasopoulos
- ▶ Time Series Analysis and Its Applications with R Examples by Robert H. Shumway and David S. Stoffer
- astsa
 - Includes datasets and scripts to accompany Time Series Analysis and Its Applications with R Examples by Robert H. Shumway and David S. Stoffer

Outline

- ▶ What is a time series? Time series model?
 - ► Time Series Examples
- Fundamental building blocks of time series
 - White noise
 - Moving average
 - ► Random walk
- Second order properties and stationarity
- ► When/why linear regression fails

What is a Time Series: Definition

A sequence of values of a variable at **equally spaced** intervals with a **natural temporal ordering**.

What is a Time Series: Johnson and Johnson

Let's look at a dataset provided in the astsa package which shows **quarterly** earnings per share for Johnson and Johnson from the first quarter of 1960 to the last quarter of 1980.

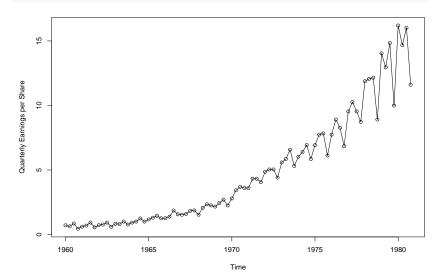
```
> require(astsa)
> window(jj, 1960, c(1965,4))
```

```
## Qtr1 Qtr2 Qtr3 Qtr4
## 1960 0.71 0.63 0.85 0.44
## 1961 0.61 0.69 0.92 0.55
## 1962 0.72 0.77 0.92 0.60
## 1963 0.83 0.80 1.00 0.77
## 1964 0.92 1.00 1.24 1.00
## 1965 1.16 1.30 1.45 1.25
```

What is a Time Series: Johnson and Johnson

If we plot the time series, we get:

> plot(jj, type="o", ylab="Quarterly Earnings per Share")



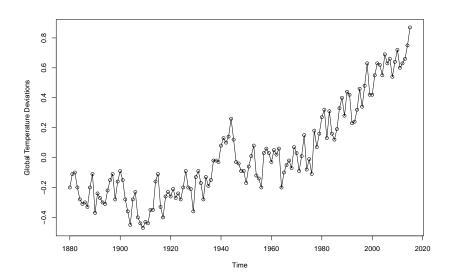
What is a Time Series: Global Temperature

Let's look at another dataset which shows the **yearly** global mean land-ocean temperature deviations (from 1951-1980 average), measured in degrees centigrade, for the years 1880-2015.

```
> class(globtemp)
## [1] "ts"
> window(globtemp, 1880, 1885)
## Time Series:
## Start = 1880
## End = 1885
## Frequency = 1
## [1] -0.20 -0.11 -0.10 -0.20 -0.28 -0.31
```

What is a Time Series: Global Temperature

plot(globtemp, type="o", ylab="Global Temperature Deviations")



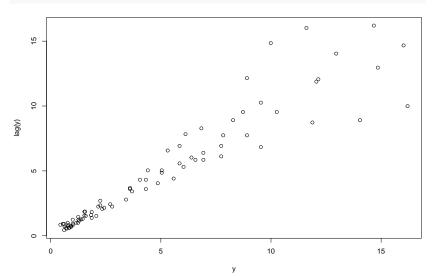
Key Assumption

Can't assume consecutive observations are independent

```
> jj_v <- unclass(jj)</pre>
> lag <- lag(jj_v, 1)</pre>
> lm <- lm(jj_v ~ lag)
> cor.test( ~ jj_v + lag,
+ method = "pearson",
+ conf.level = 0.95)
##
    Pearson's product-moment correlation
##
##
## data: jj_v and lag
## t = 25.987, df = 81, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.9159239 0.9641248
## sample estimates:
##
         cor
## 0.9449363
```

Key Assumption

Can't assume consecutive observations are independent



Why Care about Time Series?

- ► They are everywhere!
- ► Economics (stock market, unemployment, etc)
- Social sciences (population series like birth rates or school enrollments)
- ► Epidemiology (influenza outbreaks)
- Medicine (blood pressure measurements)

Working with Time Series

- ► Time series analysis
 - Analyzing observed data
 - Focus on characteristics of the data
 - Explanatory focus
- ► Time series forecasting
 - Generating a model
 - Predictive focus

Some Additional Terminology

Stochastic process

- A sequence of random variables
- Example: flipping a coin

Sample path

- Sample path of a stochastic process
- One sample from a stochastic process
- Example: HTHHTT (six coin flips)
- A stochastic process can generate MANY sample paths (infinitely many)

Fundamental Stochastic Processes

- ► White noise
- Moving average
- ► Random walks

White Noise

- ► Fundamental building block of other stochastic processes
- Y_t is called white noise (process looks like white light of spectrometers)

Key: No correlation between observations

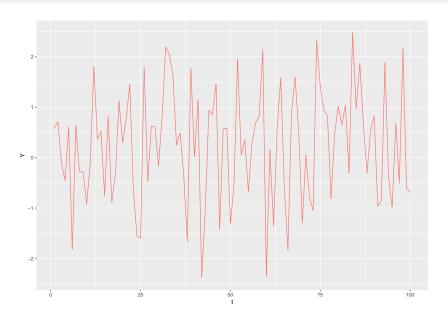
White Noise: Definition

- Assume $e_t \sim N(0,1)$ is a collection of IID random variables all following a normal distribution
- Normal distribution? Anyone? Anyone at all?

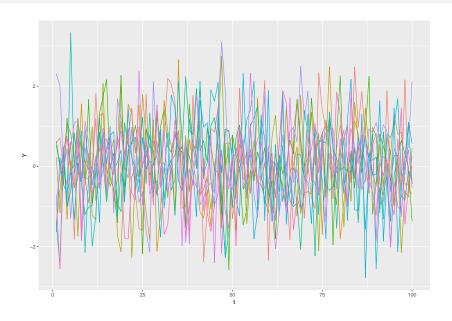
Define
$$Y_t = e_t$$
 for all t

- ► A white noise process satisfies:
 - ightharpoonup Mean(Y_t) = constant
 - $ightharpoonup Var(Y_t) = constant$
 - ► Auto-covariance = 0

White Noise: 1 Sample Path



White Noise: Many Sample Paths



White Noise: Implications

Since the data is random there is no point in modeling



Moving Average

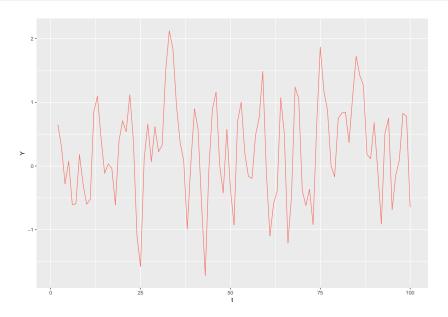
Assume $e_t \sim N(0,1)$ is a collection of IID random variables all following a normal distribution

Define
$$Y_t = \frac{e_t + e_{t+1}}{2}$$
 for all t

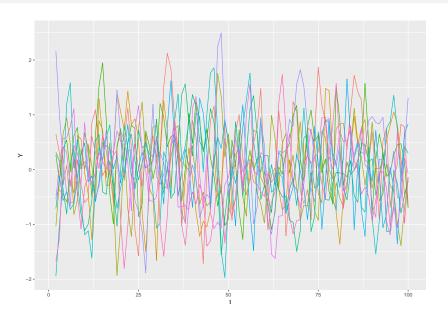
 \triangleright Y_t is a type of moving average stochastic process

Big difference: there IS correlation between observations

Moving Average: 1 Sample Path



Moving Average: Many Sample Paths



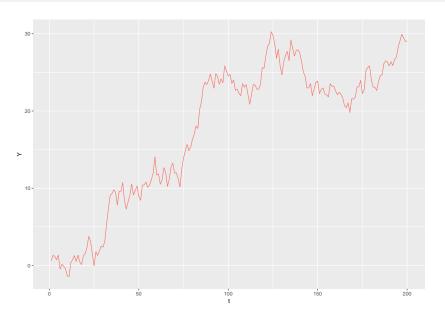
Random Walk

Assume $e_t \sim N(0,1)$ is a collection of IID random variables all following a normal distribution

Define
$$Y_t = e_1 + e_2 + ... + e_t$$
 for all t

 \triangleright Y_t is a random walk

Random Walk: 1 Sample Path



Random Walk: Many Sample Paths



Why Care about Stochastic Processes?

- ► We have ONE sample path of observations
- ► From this sample path we intend to infer the stochastic process that generated it
- ▶ We are NOT fitting lines to the data
- ▶ We are understanding the sample paths the process could take

Typical Process

- ▶ We observe a process to a specific point
- ▶ We determine what sample paths are likely as we look later in time

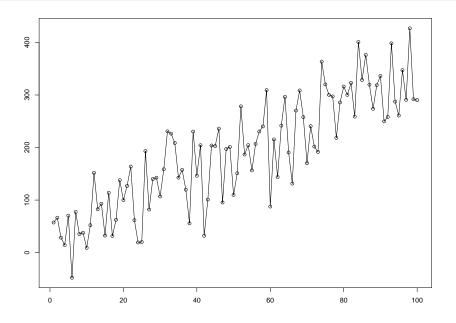
Stationarity

- ► Heavily mathematical definition
- Essence of it:
 - ▶ The mean (or expected value of the stochastic process) is constant
 - Covariance (or corelation between points only depends on distance between points and not on the value of t)
- Stationarity is an extremely common assumption in time series modeling!

Time Series: Linear Trend

- ▶ Assume stochastic process is $Y_t = \beta_0 + \beta_1 t + X_t$
- ▶ Assume $E[X_t] = 0$ and X_t is stationary
- ▶ This is just like linear regression so let's use OLS

Time Series: Linear Trend



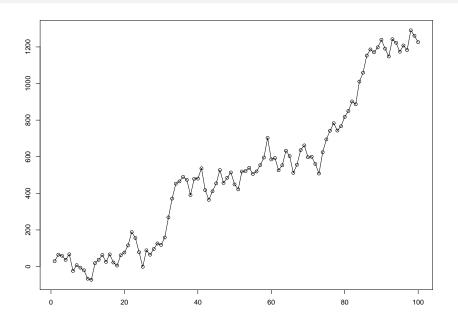
Time Series

► Linear regression recovers the true parameters (close)

Time Series: Linear Model for Random Walk

- \blacktriangleright What if we fit a random walk Y_t with a linear model?
- Put another way: what if we regress a random walk on time with a linear model?
- ► Good choice, bad choice?

Linear Model for a Random Walk



Linear Model for a Random Walk

- Linear regression fails. Hard
- Significant trend when there is NOT one
- ightharpoonup Reminder: there is not a trend because the expected value of Y_t is 0

What Happened?

- Spurious regression
- ▶ DO NOT regress non-stationary time series
- Our mistake
 - Coefficient is okay
 - Standard error is underestimated
- ► A random walk does not follow required linear regression assumptions!
 - Which key assumption does it not follow?