

Lecture 1

An Overview of Time Series Analysis

Readings: CC08 Chapter 1; SS17 Chapter 1; BD16 Chapter
1.1 - 1.3

MATH 8090 Time Series Analysis
Week 1

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Agenda

Time Series Data

Time Series Models

Objectives of Time
Series Analysis

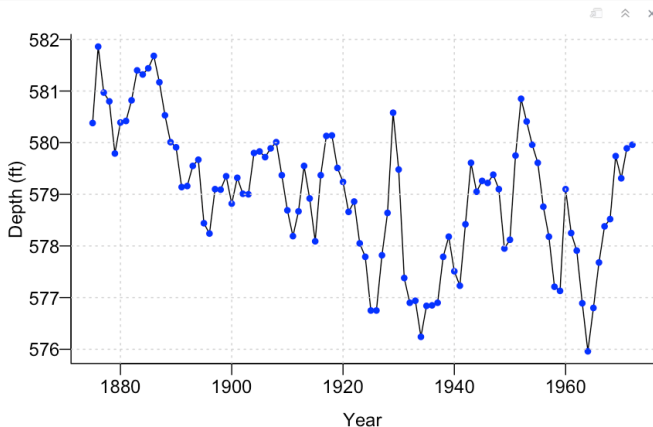
- 1 Time Series Data
- 2 Time Series Models
- 3 Objectives of Time Series Analysis

Level of Lake Huron 1875–1972

Annual measurements of the level of Lake Huron in feet.

[Source: Brockwell & Davis, 1991]

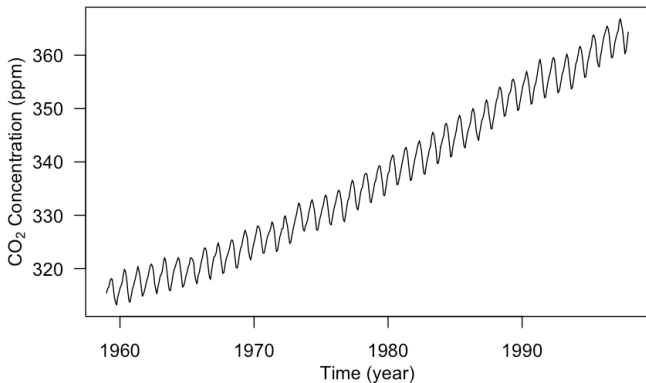
```
```{r}
par(mar = c(3.2, 3.2, 0.5, 0.5), mgp = c(2, 0.5, 0), bty = "L")
data(LakeHuron)
plot(LakeHuron, ylab = "Depth (ft)", xlab = "Year", las = 1)
points(LakeHuron, cex = 0.8, col = "blue", pch = 16)
grid()
```
```



Mauna Loa Atmospheric CO₂ Concentration

Monthly atmospheric concentrations of CO₂ at the Mauna Loa Observatory [Source: Keeling & Whorf, Scripps Institution of Oceanography]

```
````{r}```  
data(co2)
par(mar = c(3.8, 4, 0.8, 0.6))
plot(co2, las = 1, xlab = "", ylab = "")
mtext("Time (year)", side = 1, line = 2)
mtext(expression(paste("CO"[2], " Concentration (ppm)")), side = 2, line = 2.5)
````
```



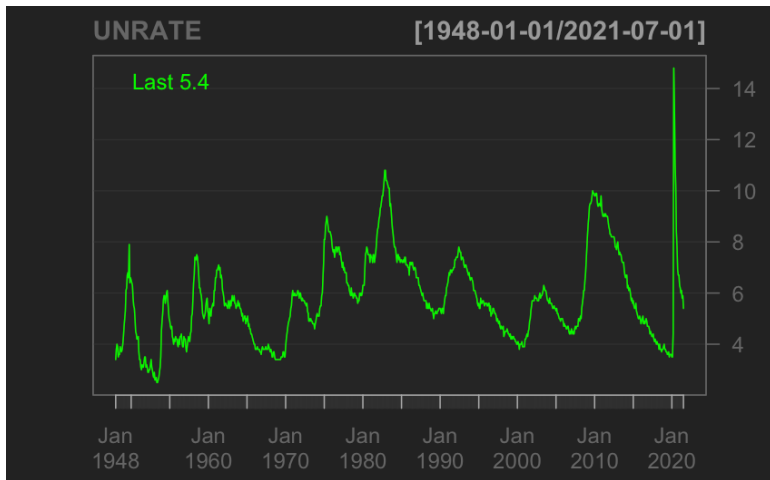
US Unemployment Rate 1948 Jan. – 2021 July

[Source: St. Louis Federal Reserve Bank's FRED system]

Time Series Data

Time Series Models

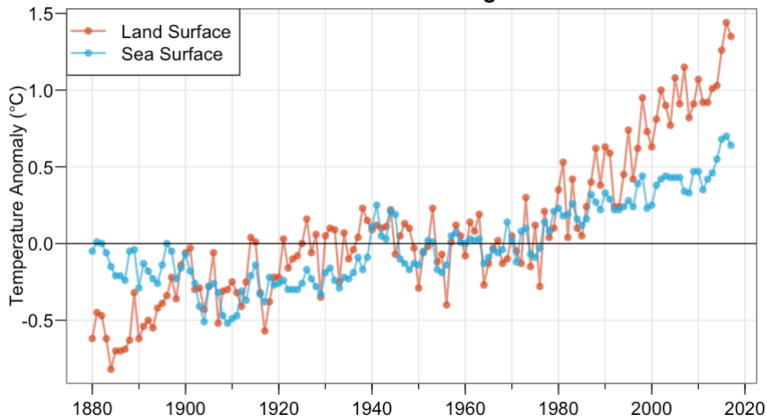
Objectives of Time Series Analysis



Global Annual Temperature Anomalies

[Source: NASA GISS Surface Temperature Analysis]

Global Warming



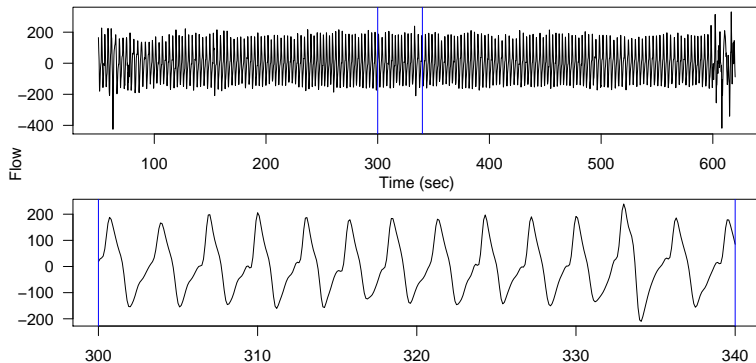
Time Series Data

Time Series Models

Objectives of Time Series Analysis

Sleep Airflow Signal

A “normal” patient’s 10 Hz sleep airflow signal [Source: [H. et al. 2022](#)]



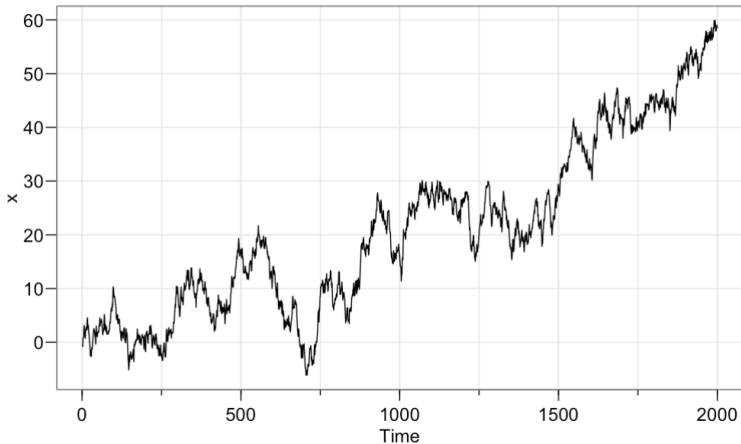
Time Series Data

Time Series Models

Objectives of Time Series Analysis

A Simulated Time Series

```
## {r}  
set.seed(123)  
w <- rnorm(2000); x <- cumsum(w); tsplot(x, las = 1)  
##
```




Time Series Data

Time Series Models

Objectives of Time Series Analysis

- A **time series** is a collection of observations $\{y_t, t \in T\}$ taken sequentially in time (t) with the index set T
 - $T = \{0, 1, 2, \dots, T\} \subset \mathbb{Z} \Rightarrow$ **discrete-time time series**
 - $T = [0, T] \subset \mathbb{R} \Rightarrow$ **continuous-time time series**
- A discrete-time time series might be intrinsically discrete or might arise from a underlying continuous-time time series via
 - sampling (e.g., instantaneous wind speed)
 - aggregation (e.g., daily accumulated precipitation amount)
 - extrema (e.g., daily maximum temperature)
- We will focus on dealing with **discrete-time real-valued** ($Y_t \in \mathbb{R}$) **time series** in this course

- Start with a **time series plot**, i.e., to plot y_t versus t 
- Look at the following:
 - Are there abrupt changes?
 - Are there “outliers”?
 - Is there a need to transform the data?
- Examine the **trend**, **seasonal components**, and the “noise” term

● Trends

- One can think of trend, μ_t , as continuous changes, usually in the mean, over longer time scales \Rightarrow *“the essential idea of trend is that it shall be smooth”* - [Kendall, 1973]
- Usually the form of the trend is unknown and needs to be estimated. When the trend is removed, we obtain a **detrended** series

● Seasonal or periodic components

- A seasonal component s_t constantly repeats itself in time, i.e., $s_t = s_{t+kd}$
- We need to estimate the form and/or the period d of the seasonal component to **deseasonalize** the series

● The “noise” process

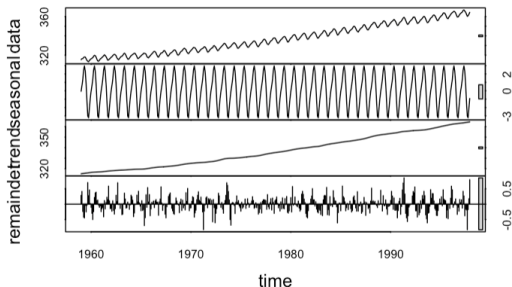
- The noise process, η_t , is the component that is neither trend nor seasonality
- We will focus on finding plausible (typically stationary) statistical models for this process

Combining Trend, Seasonality, and Noise Together

There are two commonly used approaches

- Additive model:

$$y_t = \mu_t + s_t + \eta_t, \quad t = 1, \dots, T$$



- Multiplicative model:

$$y_t = \mu_t s_t \eta_t, \quad t = 1, \dots, T$$

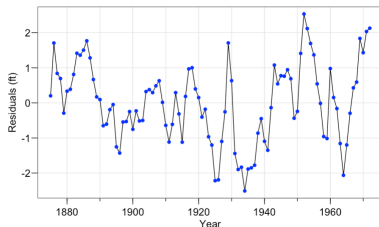
If all $\{y_t\}$ are positive then we obtain the additive model by taking logarithms:

$$\log y_t = \log \mu_t + \log s_t + \log \eta_t, \quad t = 1, \dots, T$$

Time Series Models

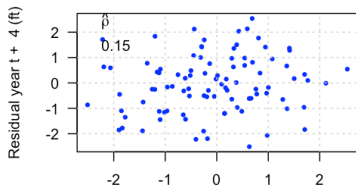
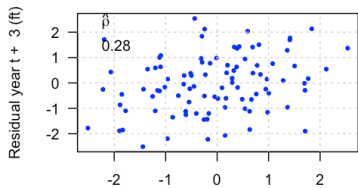
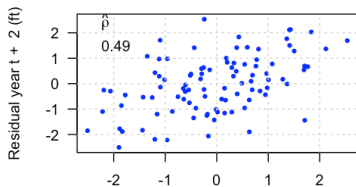
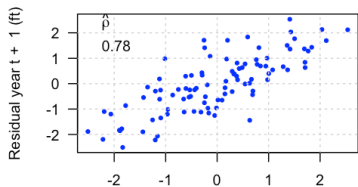
Lake Huron Time Series

- Time series analysis is the area of statistics which deals with the analysis of dependency between different observations (typically $\{\eta_t\}$)
- Some key features of the Lake Huron time series:
 - decreasing trend
 - some “random” fluctuations around the decreasing trend
- We extract the “noise” component by assuming a linear trend



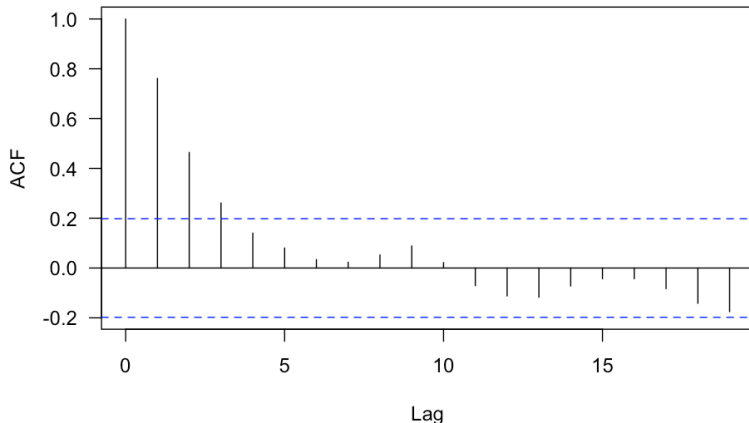
Exploring the Temporal Dependence Structure of $\{\eta_t\}$

$\{\eta_t\}$ exhibit a temporal dependence structure, meaning that the nearby (in time) values tend to be more alike than those that are far part. To observe this, let's create a few time lag plots



Further Exploration of the Temporal Dependence Structure

Let's plot the correlation as a function of the time lag



Time Series Data

Time Series Models

Objectives of Time Series Analysis

In a few weeks we will learn how to use this information to suggest an appropriate model

- A **time series model** is a probabilistic model that describes ways that the series data $\{y_t\}$ could have been generated
- More specifically, a time series model is usually a probability model for $\{Y_t : t \in T\}$, **a collection of random variables indexed in time**
- We will try to keep our models for $\{Y_t\}$ as simple as possible by assuming **stationarity** \Rightarrow some characteristic of the distribution of $\{Y_t\}$ does not depend on the time points, only on the “time lag”
- While most time series are not stationary, one either remove or model the non-stationary parts (e.g., de-trend or de-seasonalization) so that we are only left with a stationary component $\{\eta_t\}$. We typically further assume that the process is **second order stationary** \Rightarrow
 $E[\eta_t] = 0, \quad \forall t \in T$ and
 $\text{Cov}(\eta_t, \eta_{t'}) = \gamma(t' - t) = \text{Cov}(\eta_{t+s}, \eta_{t'+s})$

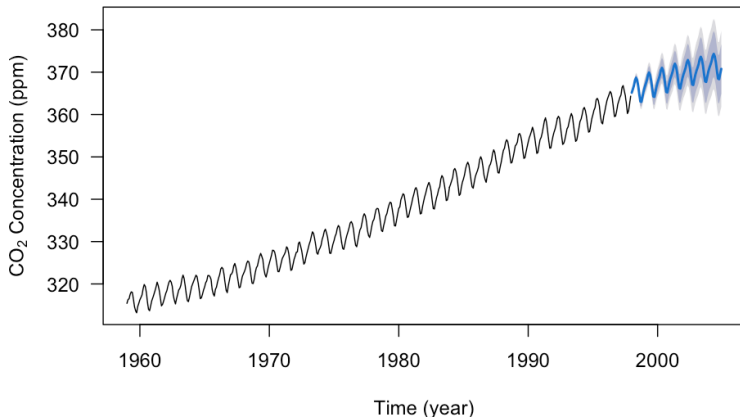
Objectives of Time Series Analysis

- **Modeling:** Find a **statistical model** that adequately explains the observed time series
- For example, identify a model which can account for the fact that the depths of Lake Huron are correlated with different years and with a decreasing long-term trend
- The fitted model can be used for further **statistical inference**, for instance, to answer the question like: **Is there evidence of decreasing trend in the Lake Huron depths?**

Some Objectives of Time Series Analysis, Cont'd

Forecasting is perhaps the most common objective. One observe a time series of given length and wish to **predict** or **forecast** future values of the time series based on those already observed.

Forecasts from TBATS(1, {3,1}, -, {<12,5>})



- **Adjustment:** an example would be **seasonal adjustment**, where the seasonal component is estimated and then removed to better understand the underlying trend
- **Simulation:** use a time series model (which adequately describes a physical process) as a surrogate to *simulate repeatedly in order to approximate how the physical process behaves*
- **Control:** adjust various **input (control)** parameters to make the time series fit more closely to a given standard (many examples from statistical quality control)