

Time Series and Dynamic Learning

Concepts, Notation, Visualization

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Models and Inference

A model is a simplification of reality
(and some are useful)

Observable quantities
(can be measured) Unobservable quantities
(parameters / latent variables)

Approaches: Classical / Bayesian

Data: the observed values of the observable quantities.

Computation

"A big computer, a complex algorithm and a long time does not equal science." Robert Gentleman.

Approximation

"Far better an approximate answer to the right question than the exact answer to the wrong question." John Tukey.

Models

"All models are wrong, but some are useful." George Box.

Forecasts

"Forecasts are always wrong (but we need them anyway)." Rob Hyndman

Introduction

- ▶ What is a time series?
- ▶ What is time series analysis?
- ▶ What are the objectives in time series analysis?

Time series data

A *time series* is a collection of observations taken sequentially along time.

Time series process

A *stochastic process* is a collection of random variables indexed by time.

- ▶ The most important feature of this type of data is that neighbouring observations are dependent and we are interested in analysing and modeling these dependencies.
- ▶ The chronological order of the observations is crucial for the analysis.
- ▶ Time can be replaced by other variable like space, depth, etc.

Particular features

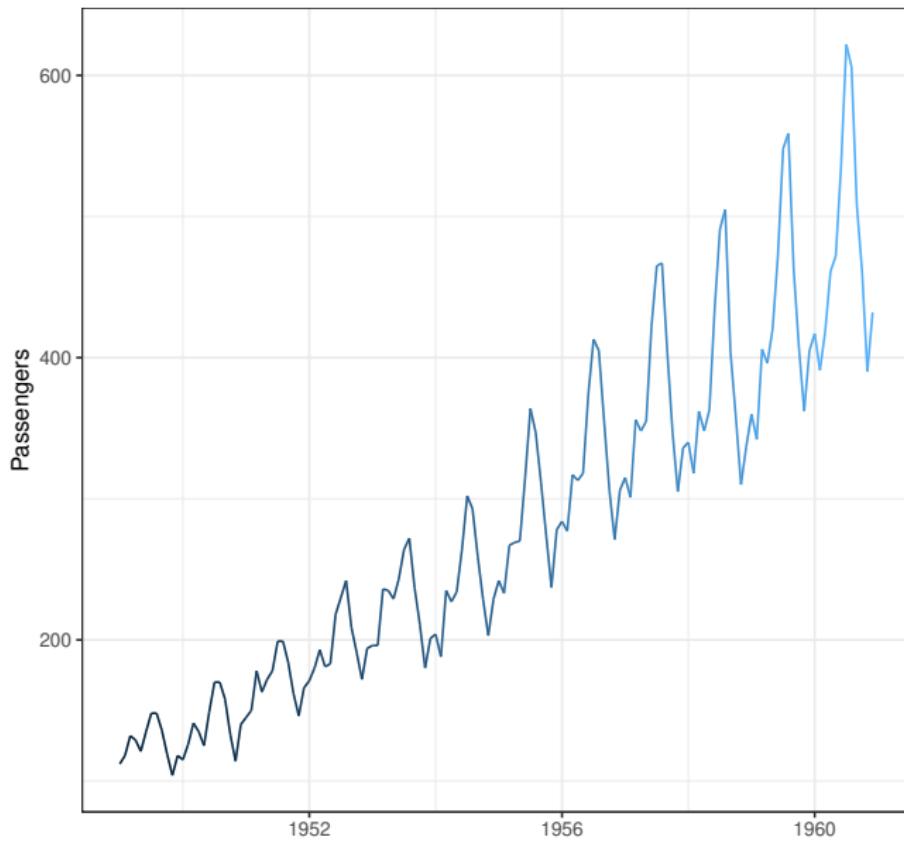
- ▶ Correlated data are in general more difficult to analyse and require specific techniques.
- ▶ We need to take into account the chronological order of the observations.
- ▶ Complicating factors such as presence of trends and seasonal or cyclical variation can be difficult to estimate or remove.
- ▶ Model selection and comparison can be more challenging.
- ▶ It is in general more difficult to deal with missing values and outliers due to the sequential nature of the observations.
- ▶ A graphical representation of the time series is always a good starting point.

Time Series Vizualization

The `ggplot2` package provides great features for time series visualization

```
> library(ggplot2)
```

Monthly totals of international airline passengers



```
> help(AirPassengers)
```

AirPassengers

package:datasets

R Documentation

Monthly Airline Passenger Numbers 1949-1960

Description:

The classic Box & Jenkins airline data. Monthly totals of international airline passengers, 1949 to 1960.

Usage:

```
AirPassengers
```

Format:

A monthly time series, in thousands.

Source:

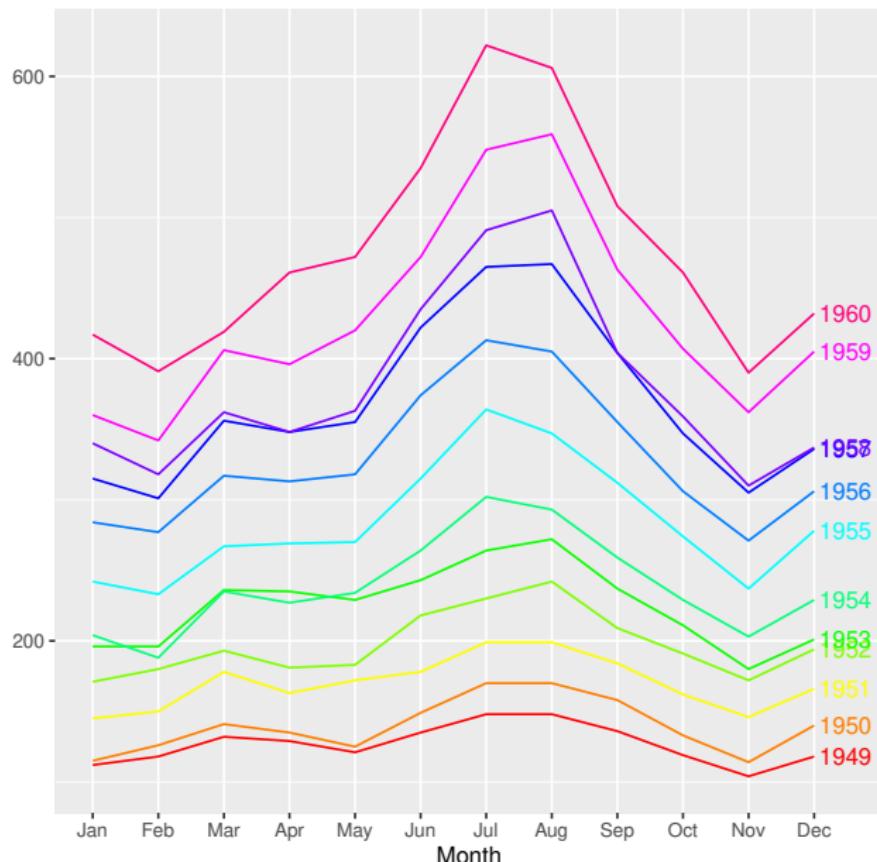
Box, G. E. P., Jenkins, G. M. and Reinsel, G. C. (1976) *_Time Series Analysis, Forecasting and Control._* Third Edition. Holden-Day. Series G.

> *AirPassengers*

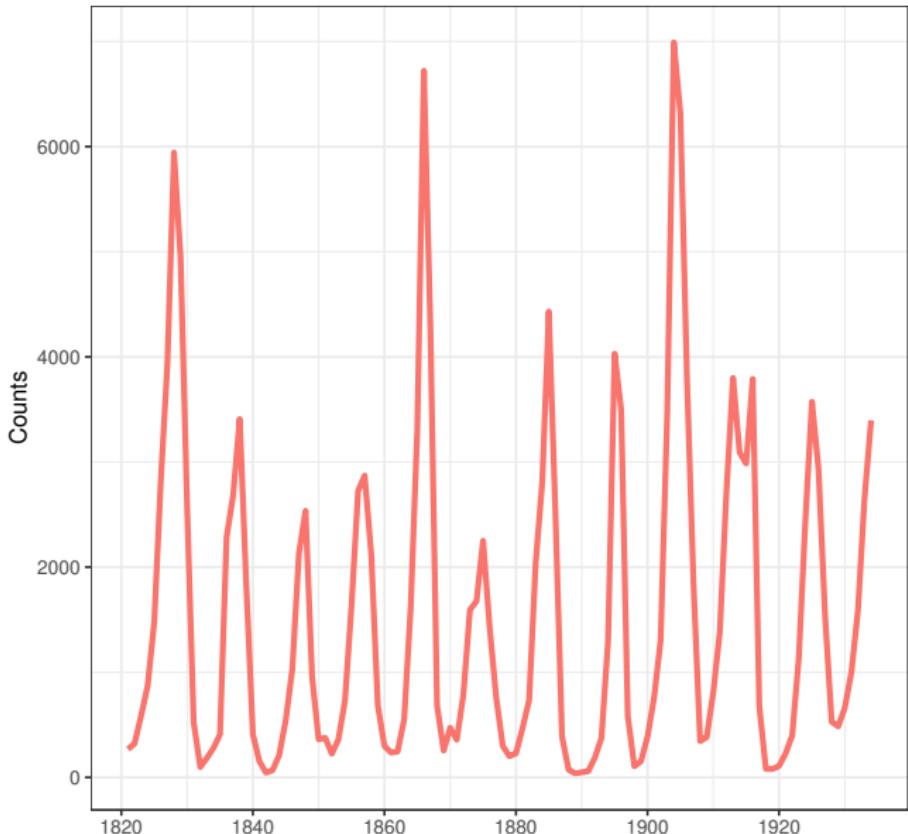
	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
1949	112	118	132	129	121	135	148	148	136	119	104	118
1950	115	126	141	135	125	149	170	170	158	133	114	140
1951	145	150	178	163	172	178	199	199	184	162	146	166
1952	171	180	193	181	183	218	230	242	209	191	172	194
1953	196	196	236	235	229	243	264	272	237	211	180	201
1954	204	188	235	227	234	264	302	293	259	229	203	229
1955	242	233	267	269	270	315	364	347	312	274	237	278
1956	284	277	317	313	318	374	413	405	355	306	271	306
1957	315	301	356	348	355	422	465	467	404	347	305	336
1958	340	318	362	348	363	435	491	505	404	359	310	337
1959	360	342	406	396	420	472	548	559	463	407	362	405
1960	417	391	419	461	472	535	622	606	508	461	390	432

```
> library(ggplot2)
> Year= as.integer(time(AirPassengers))
> Time= as.numeric(time(AirPassengers))
> data= data.frame(Year=Year,Time=Time,
+                   Passengers=as.numeric(AirPassengers))
> ggplot(data,aes(x=Time,y=Passengers,color=Year)) +
+ geom_line() +
+ theme_bw() +
+ theme(legend.position = "none") +
+ ggtitle("Monthly totals of international airline passengers") +
+ labs(x="")
```

Seasonal plot: AirPassengers



Annual numbers of lynx trappings in Canada



```
> help(lynx)
```

lynx

package:datasets

R Documentation

Annual Canadian Lynx trappings 1821-1934

Description:

Annual numbers of lynx trappings for 1821-1934 in Canada. Taken from Brockwell & Davis (1991), this appears to be the series considered by Campbell & Walker (1977).

Usage:

```
lynx
```

Source:

Brockwell, P. J. and Davis, R. A. (1991). *_Time Series and Forecasting Methods_*. Second edition. Springer. Series G (page 557).

References:

Becker, R. A., Chambers, J. M. and Wilks, A. R. (1988). *_The New S Language_*. Wadsworth & Brooks/Cole.

Campbell, M. J. and Walker, A. M. (1977). A Survey of statistical work on the Mackenzie River series of annual Canadian lynx trappings for the years 1821-1934 and a new analysis. *Journal of the Royal Statistical Society series A*, *140*, 411-431. doi: 10.2307/2345277 (<https://doi.org/10.2307/2345277>).

```
> lynx
```

Time Series:

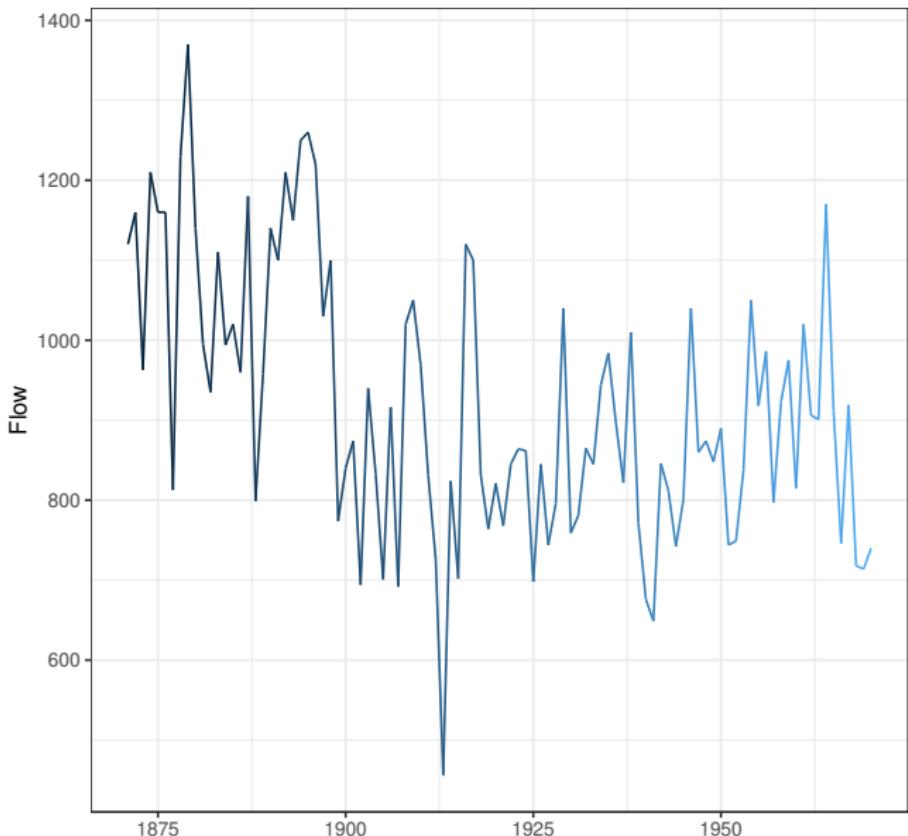
Start = 1821

End = 1934

Frequency = 1

```
[1] 269 321 585 871 1475 2821 3928 5943 4950 2577 523  
[12] 98 184 279 409 2285 2685 3409 1824 409 151 45  
[23] 68 213 546 1033 2129 2536 957 361 377 225 360  
[34] 731 1638 2725 2871 2119 684 299 236 245 552 1623  
[45] 3311 6721 4254 687 255 473 358 784 1594 1676 2251  
[56] 1426 756 299 201 229 469 736 2042 2811 4431 2511  
[67] 389 73 39 49 59 188 377 1292 4031 3495 587  
[78] 105 153 387 758 1307 3465 6991 6313 3794 1836 345  
[89] 382 808 1388 2713 3800 3091 2985 3790 674 81 80  
[100] 108 229 399 1132 2432 3574 2935 1537 529 485 662  
[111] 1000 1590 2657 3396
```

Annual Flow of the River Nile at Aswan



```
> help(Nile)
```

Nile package:datasets

R Documentation

Flow of the River Nile

Description:

Measurements of the annual flow of the river Nile at Aswan (formerly 'Assuan'), 1871-1970, in 10^8 m^3 , "with apparent changepoint near 1898" (Cobb(1978), Table 1, p.249).

Usage:

```
Nile
```

Format:

A time series of length 100.

Source:

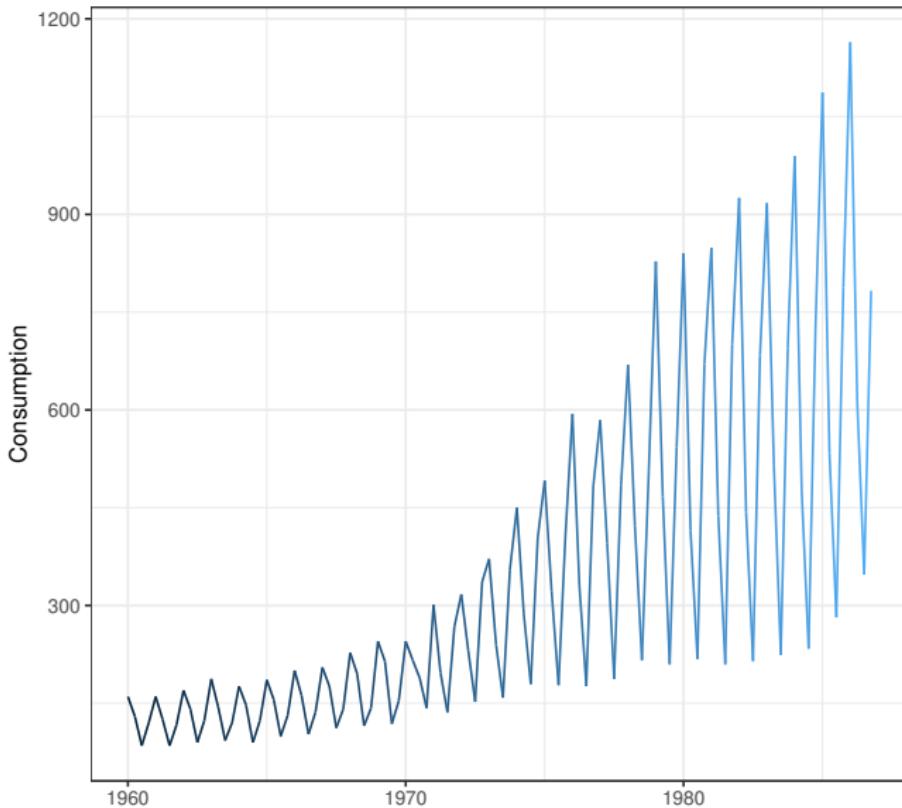
Durbin, J. and Koopman, S. J. (2001). *Time Series Analysis by State Space Methods*. Oxford University Press. <URL:
<http://www.ssftpack.com/DKbook.html>>

References:

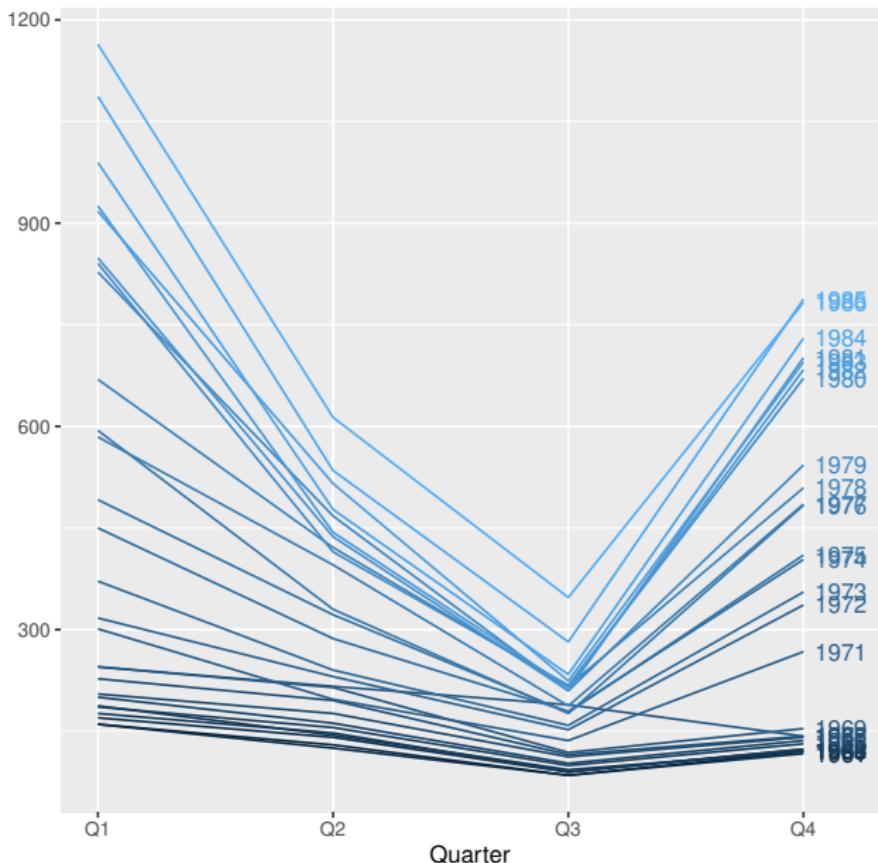
Balke, N. S. (1993). Detecting level shifts in time series. *Journal of Business and Economic Statistics*, *11*, 81-92. doi: 10.2307/1391308 (<https://doi.org/10.2307/1391308>).

Cobb, G. W. (1978). The problem of the Nile: conditional solution to a change-point problem. *Biometrika* *65*, 243-51. doi: 10.2307/2335202 (<https://doi.org/10.2307/2335202>).

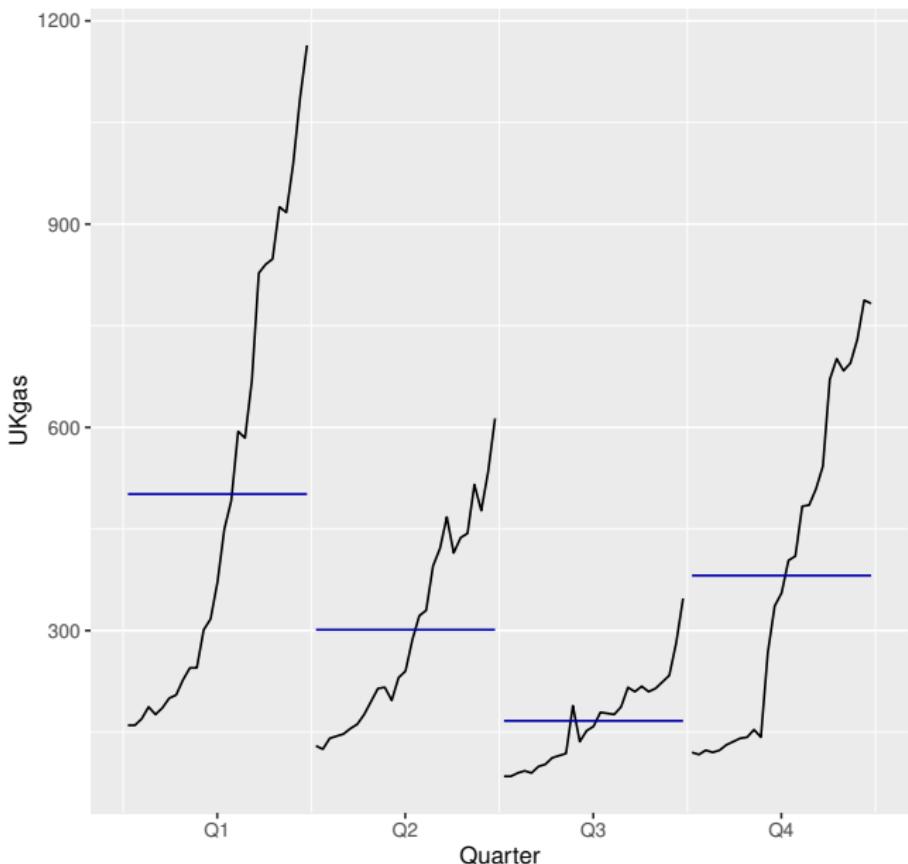
UK Quarterly Gas Consumption 1960Q1–1986Q4,
in millions of therms



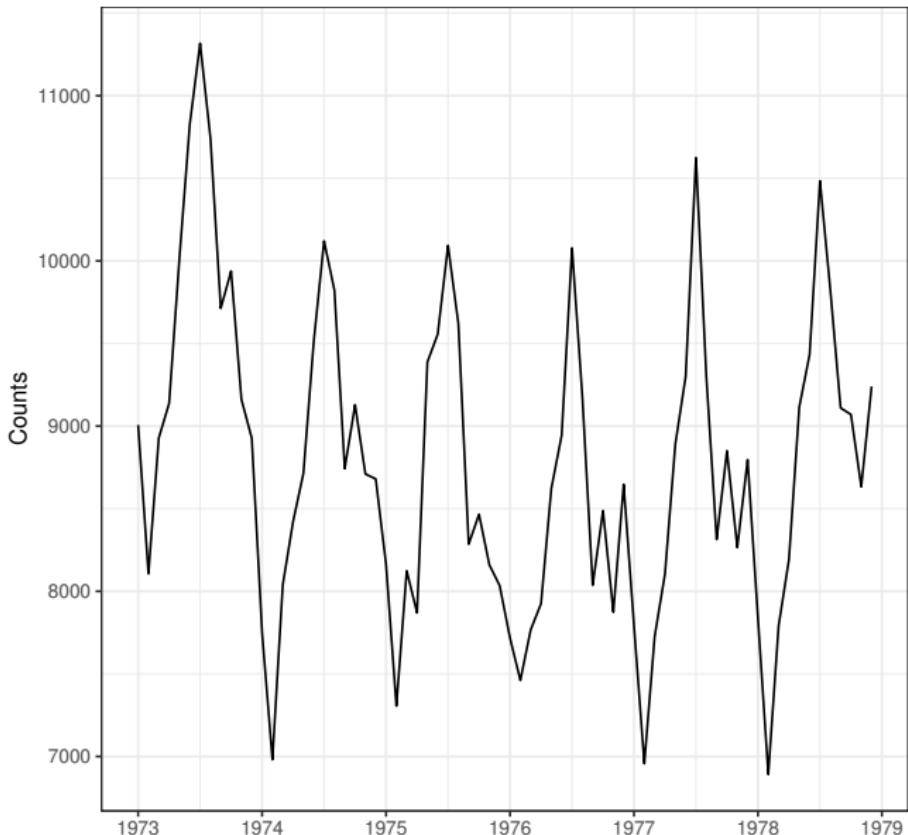
Seasonal plot: UKgas



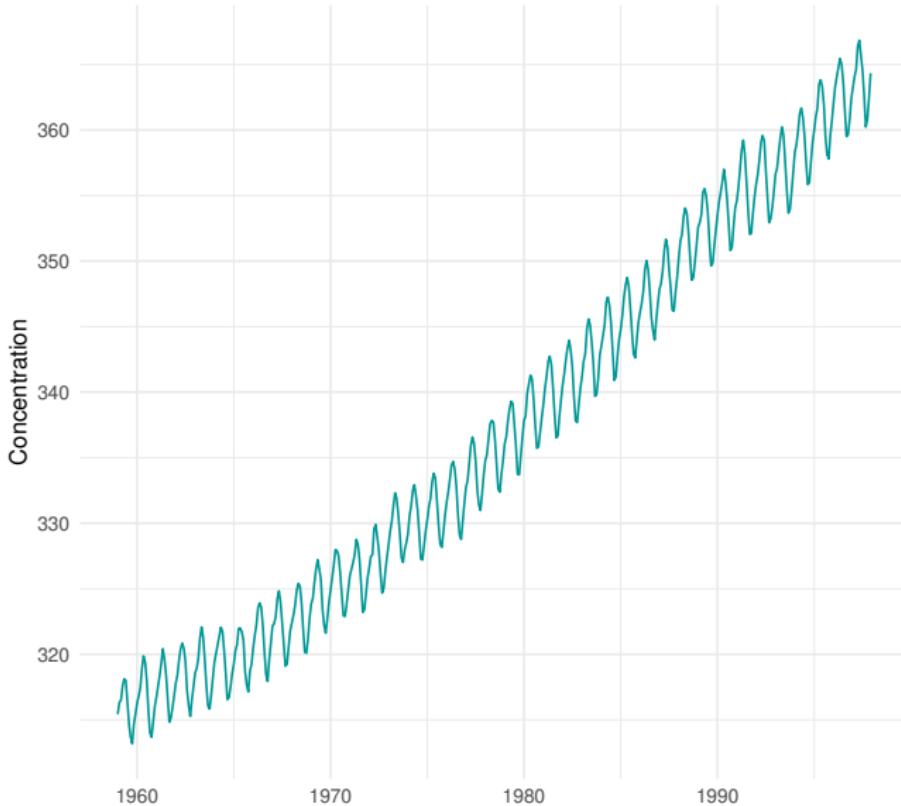
UKgas time series per season



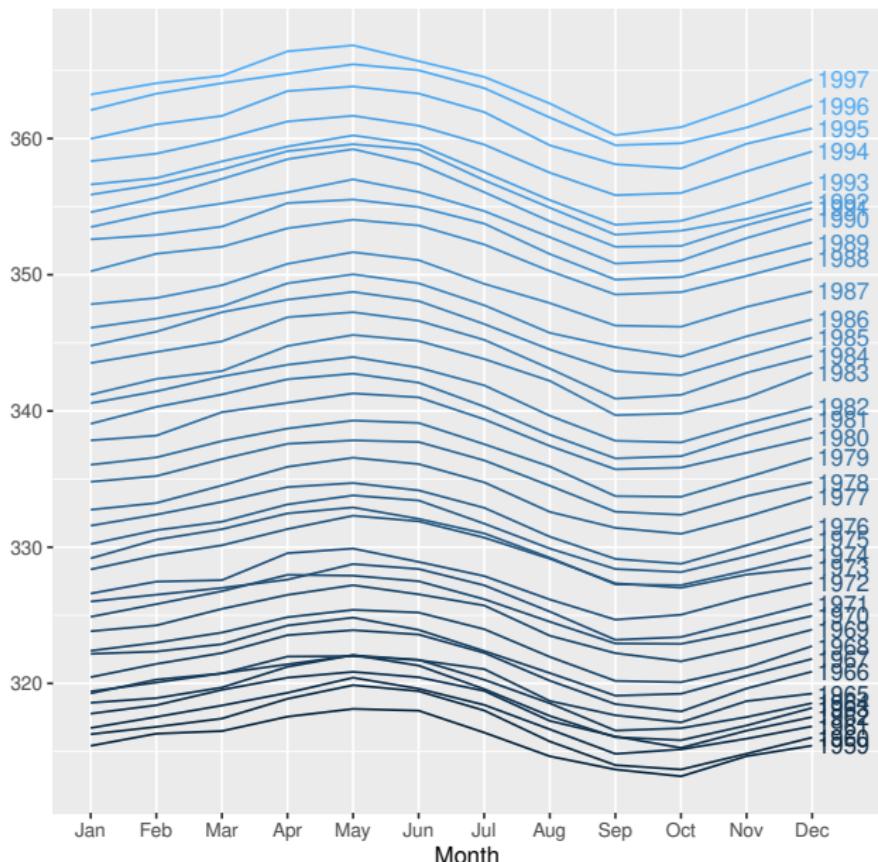
Accidental Deaths in the US 1973–1978



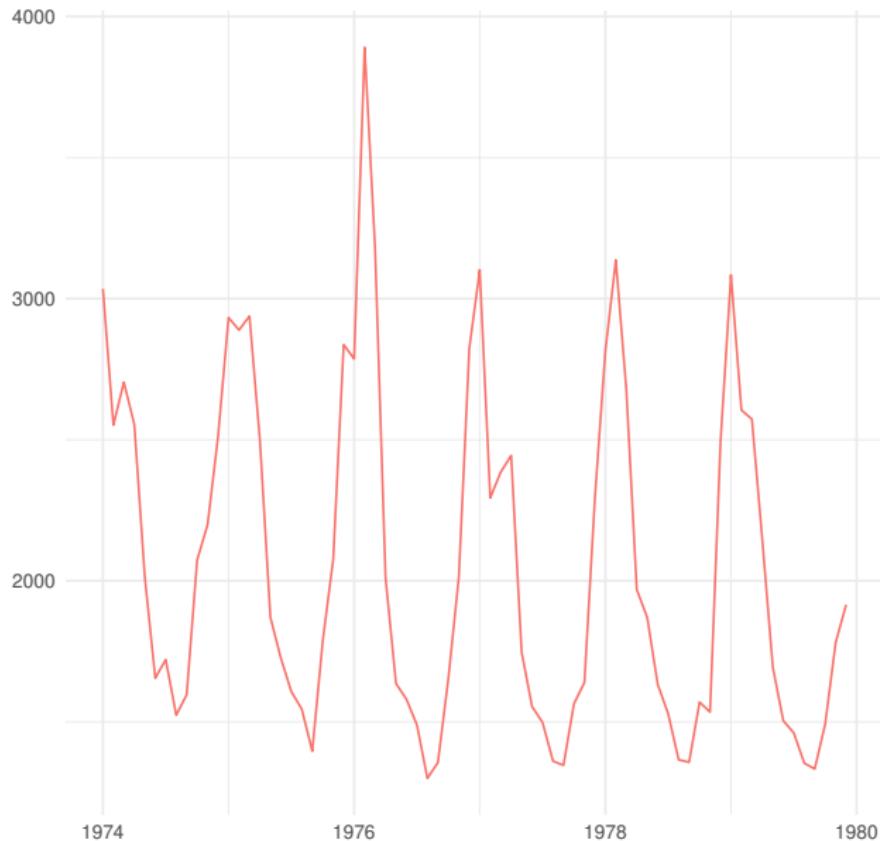
Monthly atmospheric concentrations of CO₂ (in ppm),
Mauna Loa Observatory, Hawaii



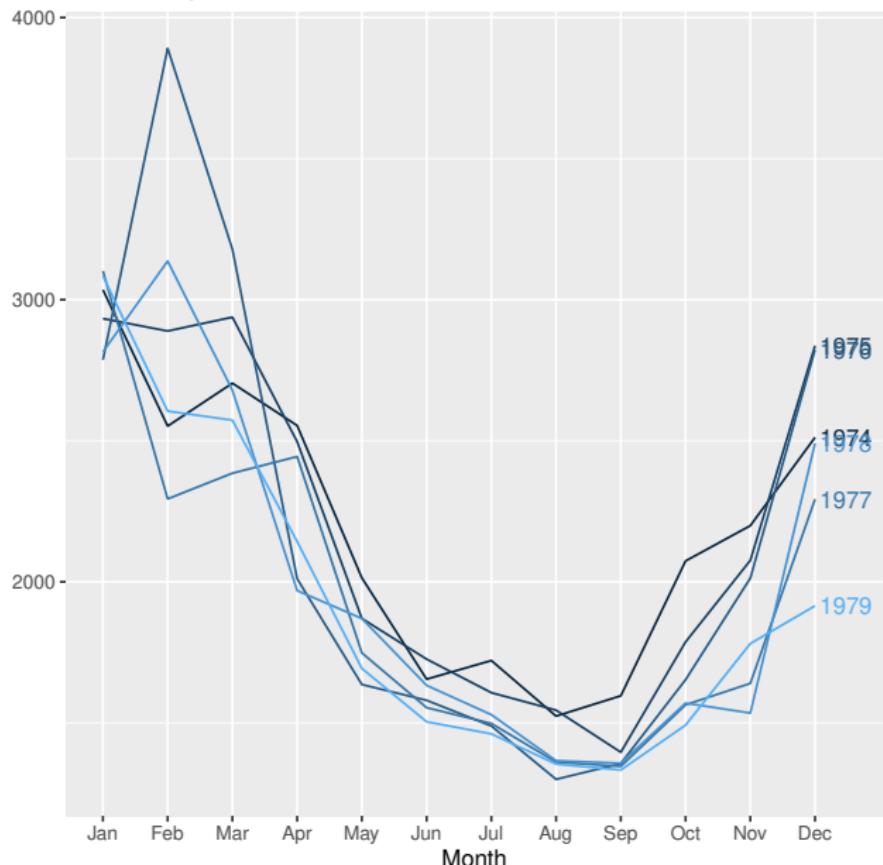
Seasonal plot: co2



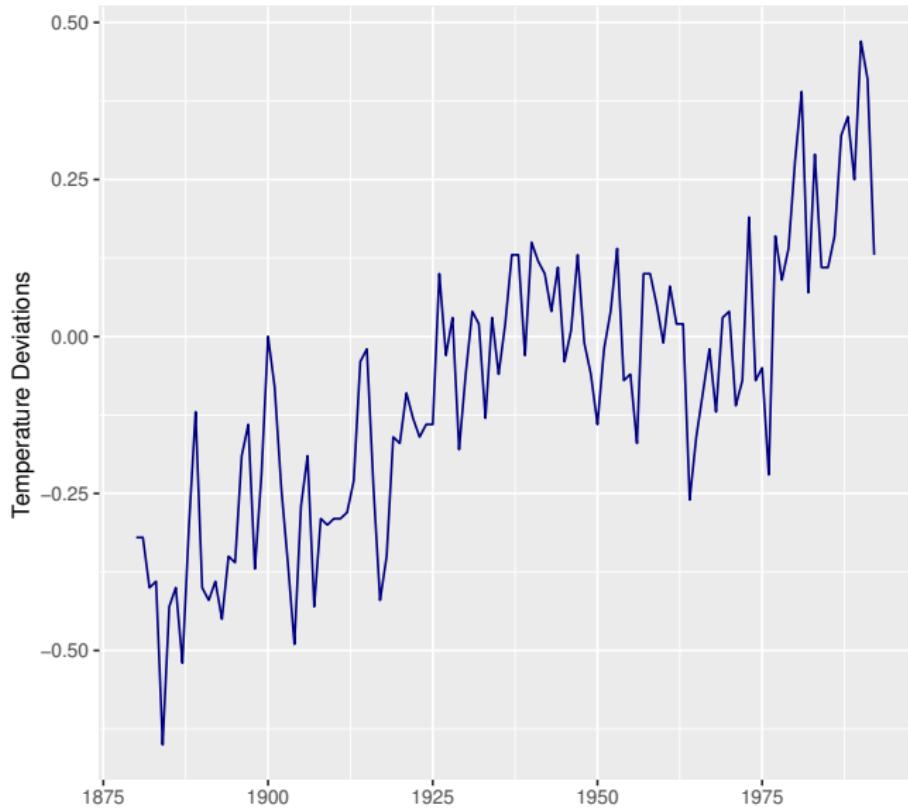
Monthly Deaths from Lung Diseases in the UK



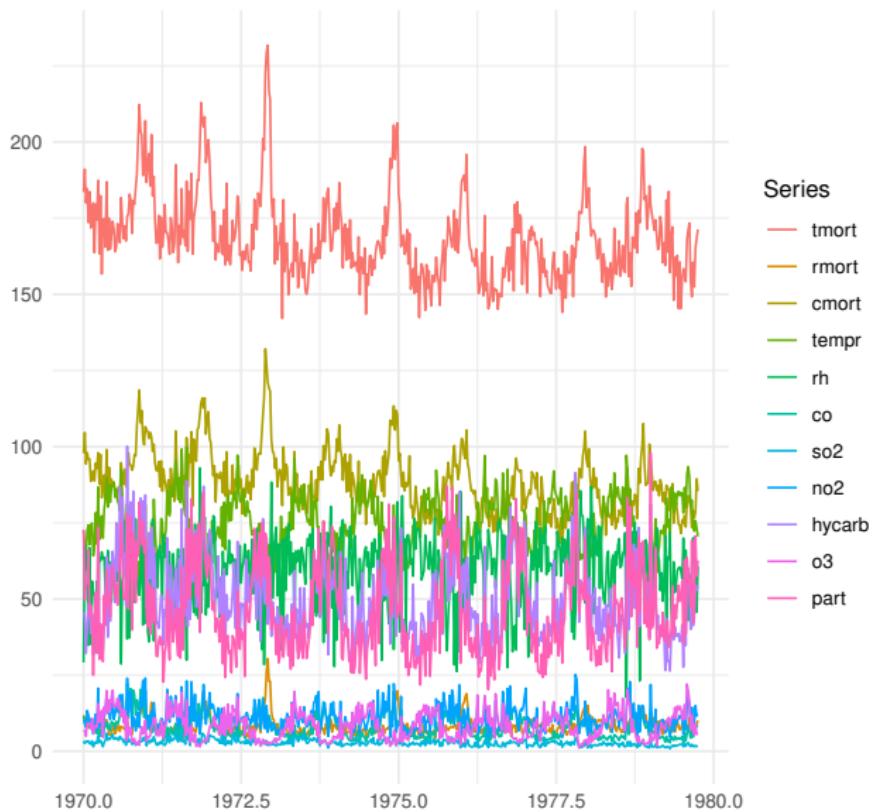
Seasonal plot: Deaths



Global mean land–ocean temperature deviations, 1880–2015
<http://data.giss.nasa.gov/gistemp/graphs>



LA Pollution–Mortality Study (1970–1979, weekly data)
<http://www.stat.pitt.edu/stoffer/tsda/>



```
> help(lap)
```

lap

package:astsa

R Documentation

LA Pollution-Mortality Study

Description:

LA Pollution-Mortality Study (1970-1979, weekly data).

Format:

The format is: mts [1:508, 1:11]

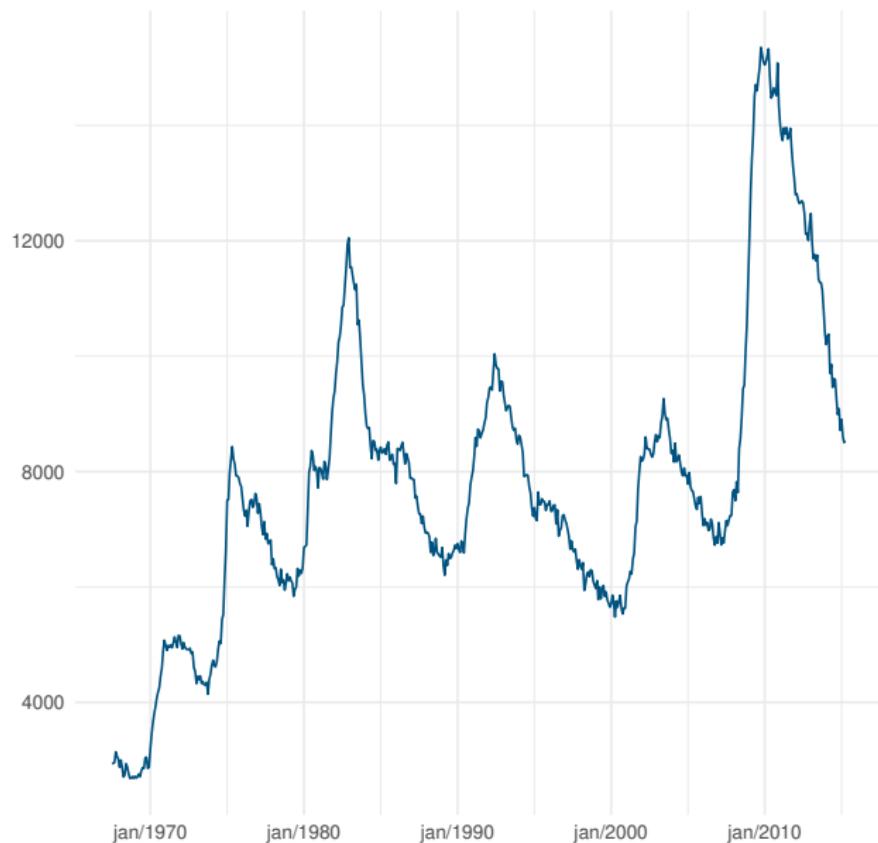
Details:

columns are time series	with names
(1) Total Mortality	'tmort'
(2) Respiratory Mortality	'rmort'
(3) Cardiovascular Mortality	'cmort'
(4) Temperature	'temp'
(5) Relative Humidity	'rh'
(6) Carbon Monoxide	'co'
(7) Sulfur Dioxide	'so2'
(8) Nitrogen Dioxide	'no2'
(9) Hydrocarbons	'hycarb'
(10) Ozone	'o3'
(11) Particulates	'part'

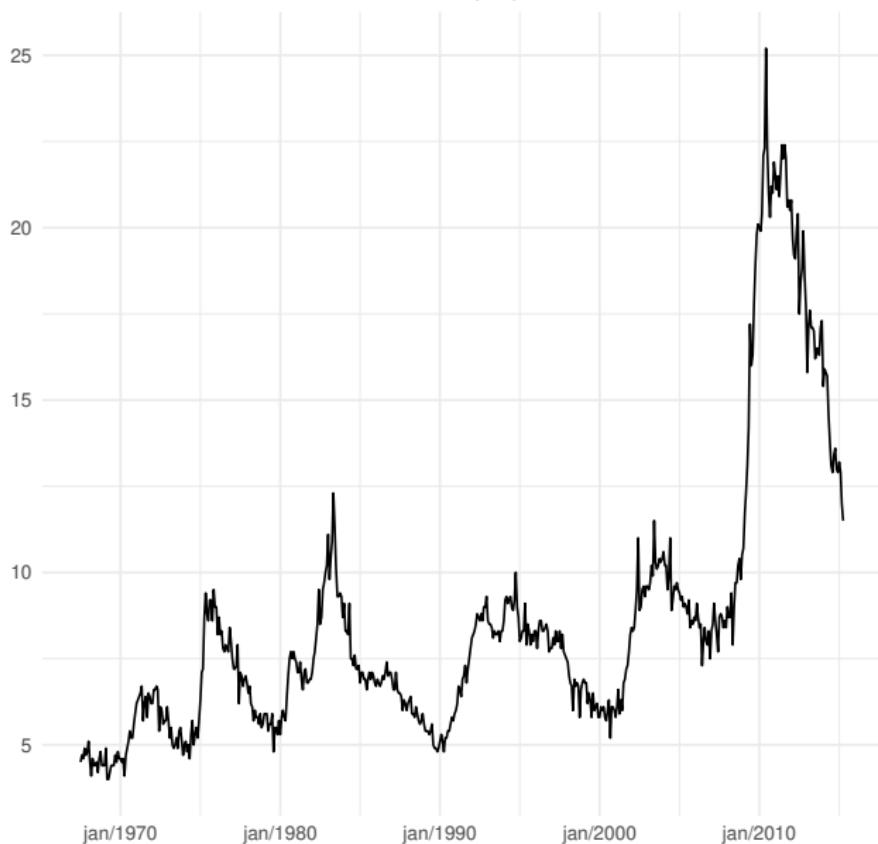
References:

<<http://www.stat.pitt.edu/stoffer/tsa4/>> and <URL:
<<http://www.stat.pitt.edu/stoffer/tsda/>>

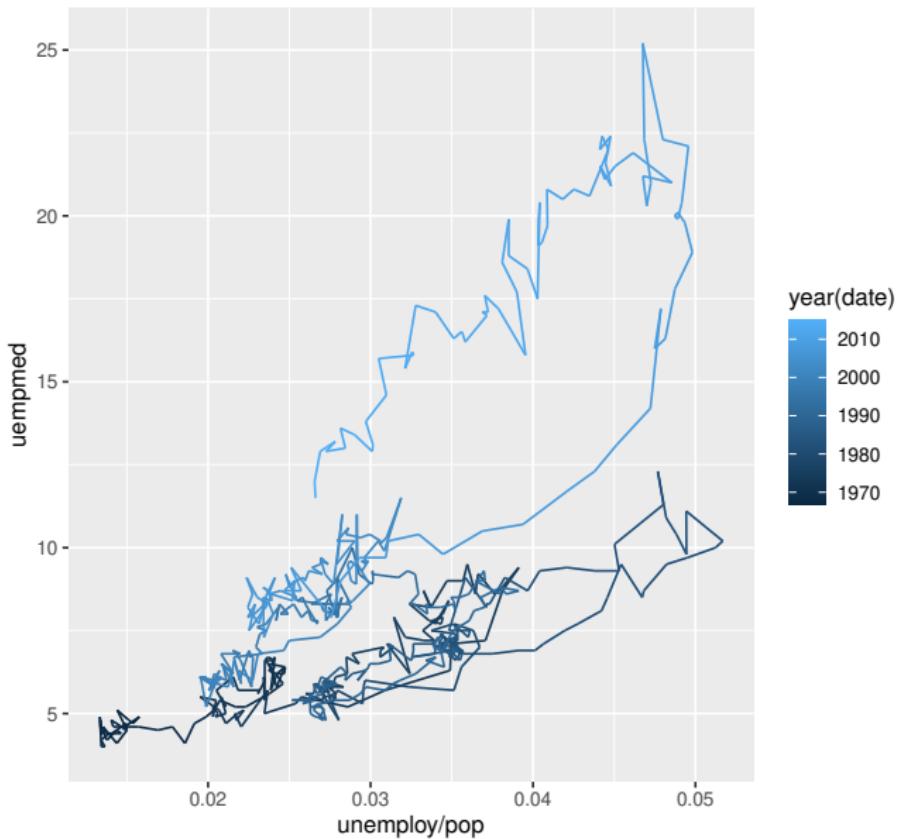
US Unemployed in Thousands (<http://research.stlouisfed.org/fred2>)



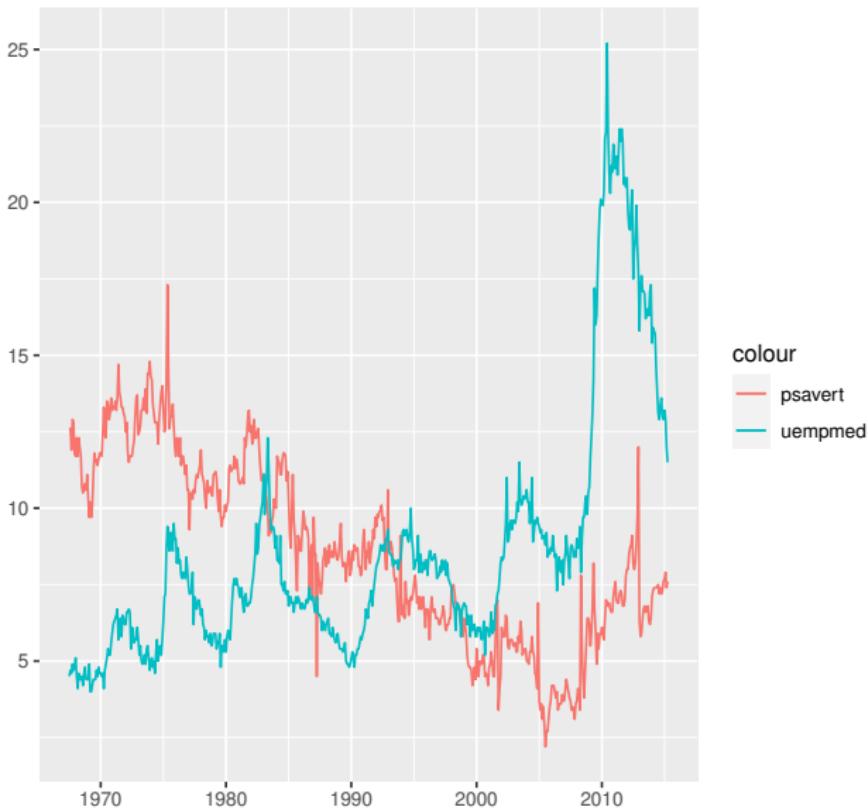
US median number of weeks unemployed



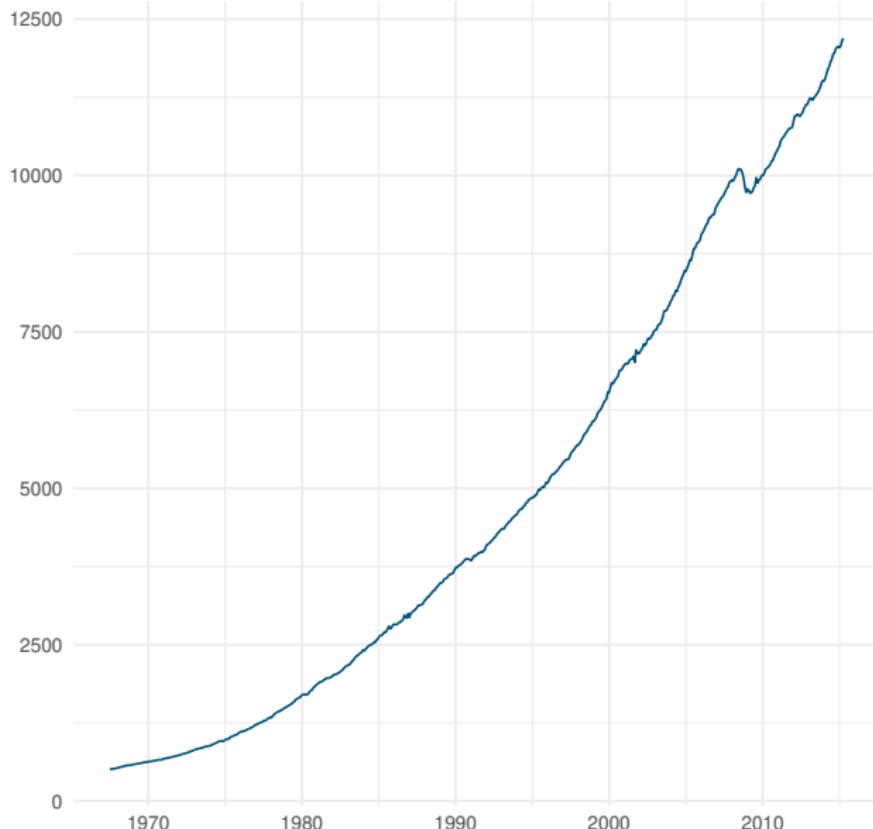
Percent of people unemployed versus
median length of unemployment



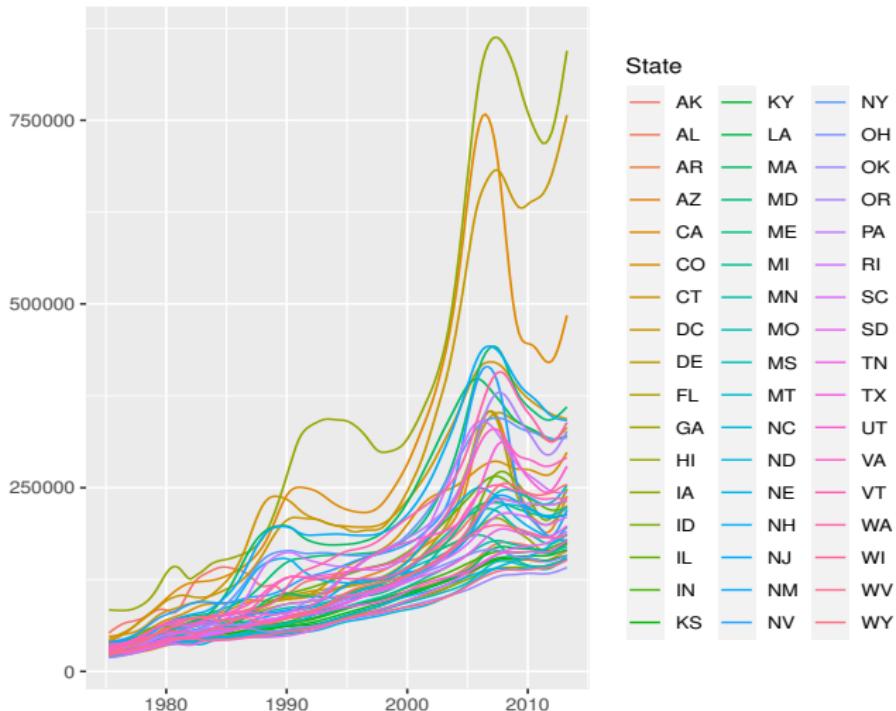
US median number of weeks unemployed (uempmmed)
and personal savings rate (psavert)



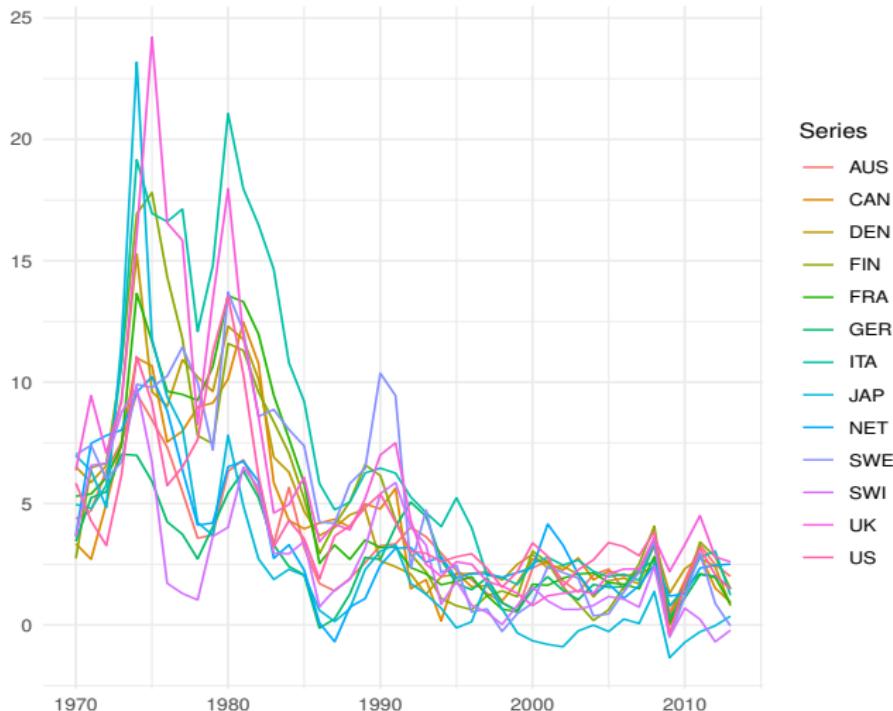
Monthly US personal consumption expenditures, in billions of dollars

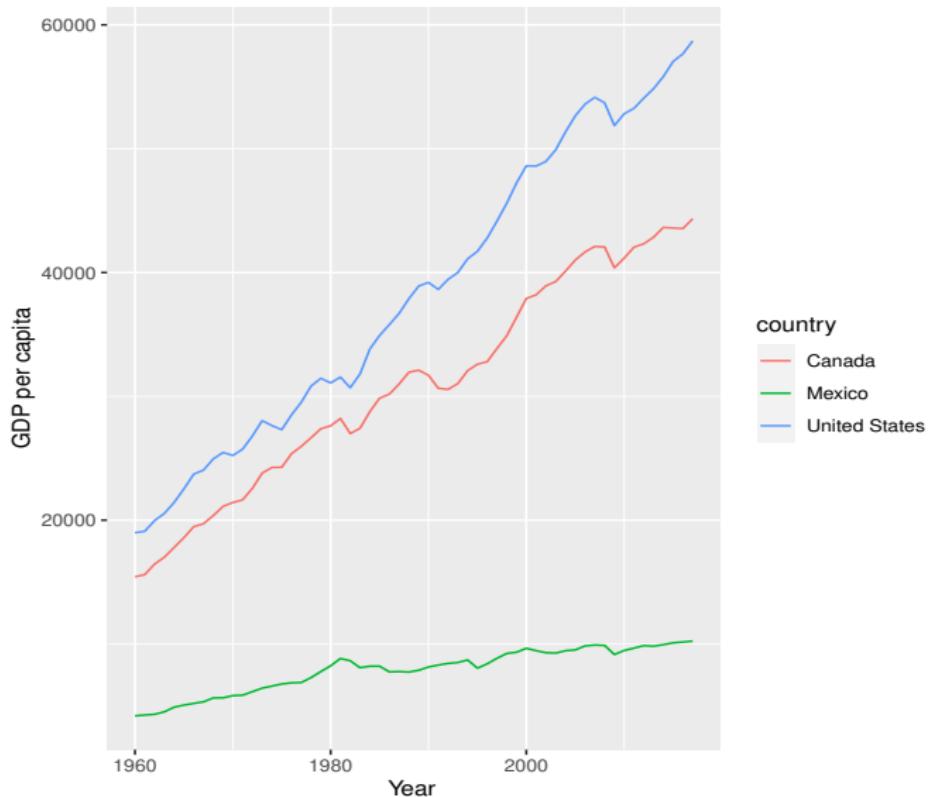


Housing prices in each US state, 1975–2013
(github.com/mattm/r-cheat-sheet/blob/master/data)



Annual inflation rates (aggregate) for 13 different countries 1970–2013





Financial Returns

In Finance, risk is frequently measured in terms of price variations of an asset. Let P_t the price of an asset at time t .

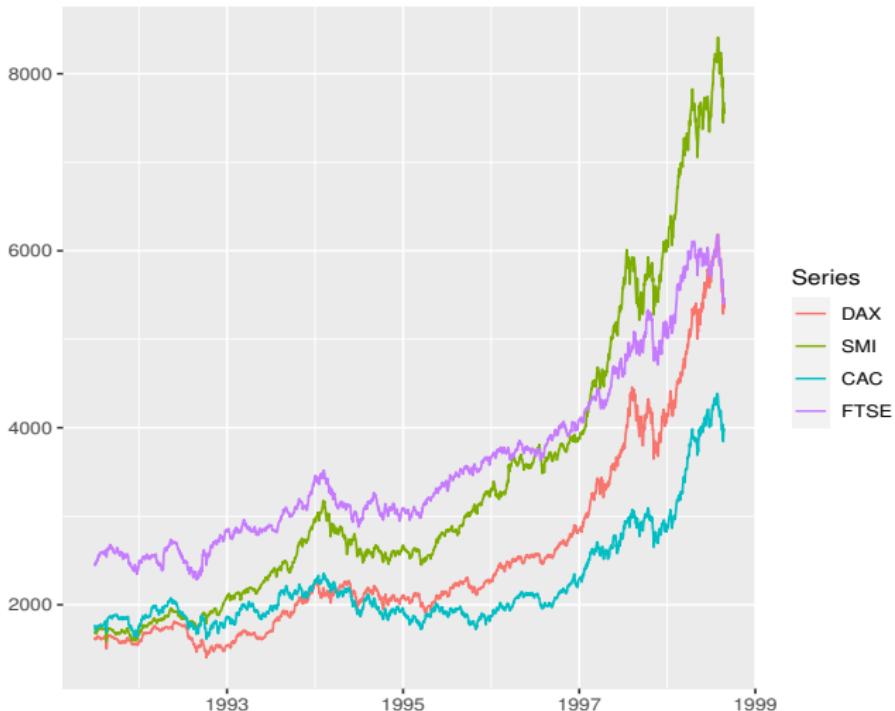
- ▶ The price variation between $t - 1$ and t (no dividends paid) is, $P_t - P_{t-1}$.
- ▶ Relative variation of prices,

$$R_t = \frac{P_t - P_{t-1}}{P_{t-1}} = \frac{P_t}{P_{t-1}} - 1.$$

- ▶ Continuously compounded return, or log-return,

$$r_t = \log\left(\frac{P_t}{P_{t-1}}\right) = \log(1 + R_t).$$

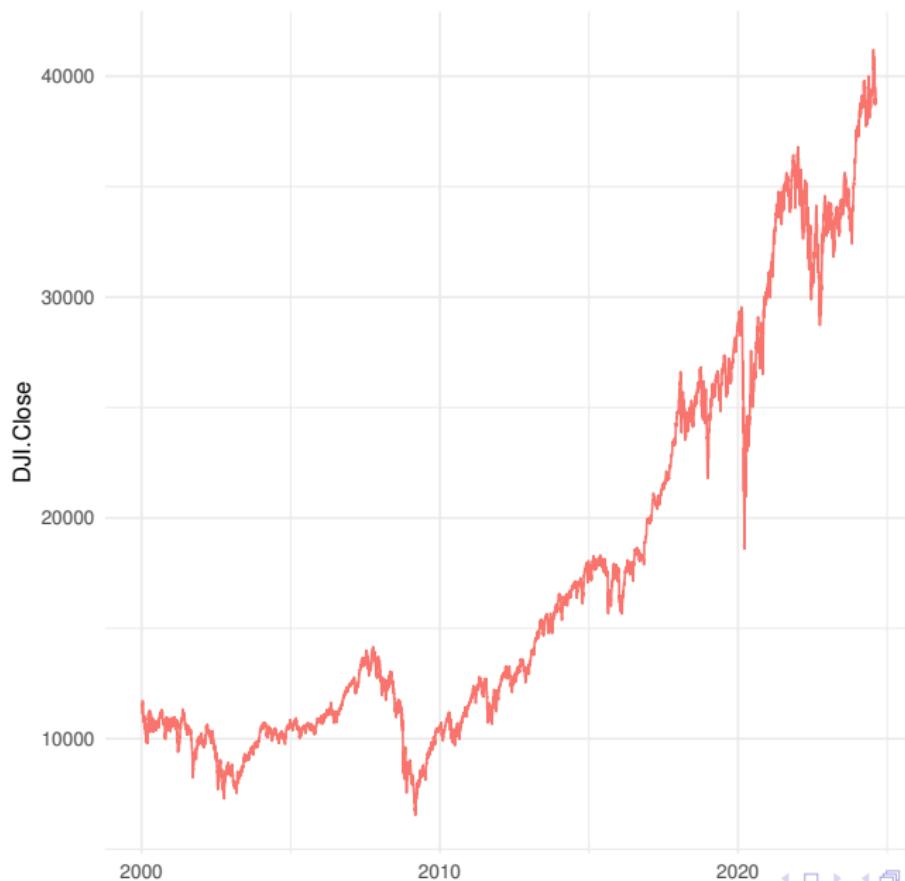
Daily Closing Prices of Major European Stock Indices, 1991–1998
EuStockMarkets package:datasets



S&P 500 and Shanghai market indexes



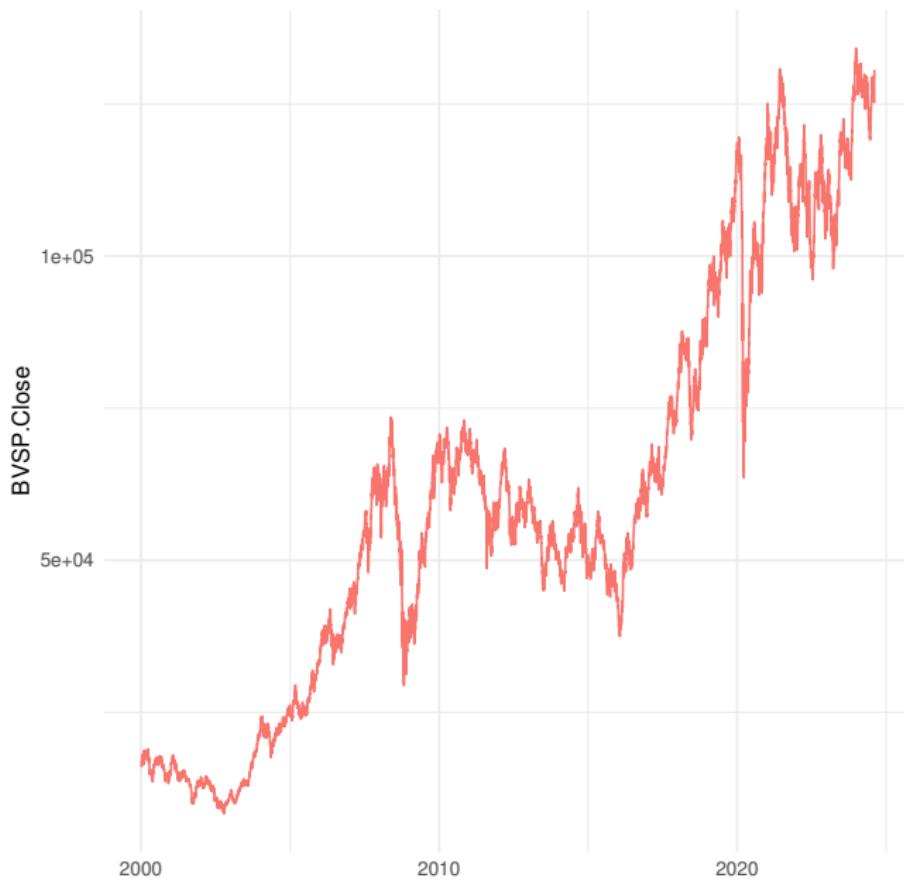
Dow Jones Industrial Average (DJI) closing adjusted daily data (Yahoo Finance) from 05/01/2000 to 09/08/2024.



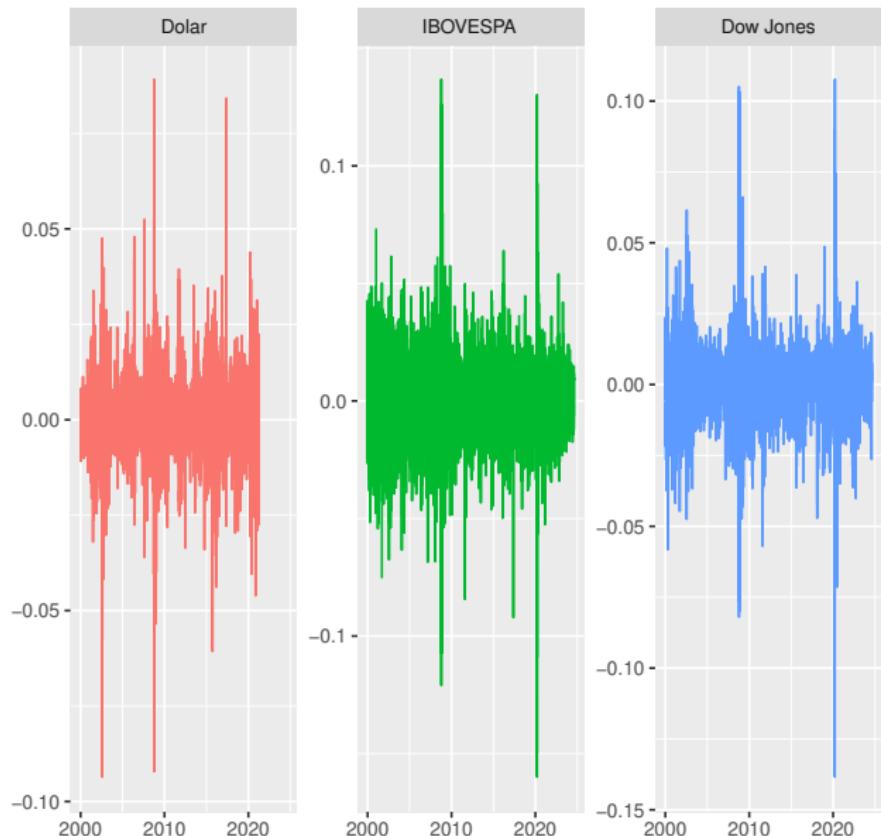
USD/BRL exchange rates from 05/01/2000 to 05/04/2021.



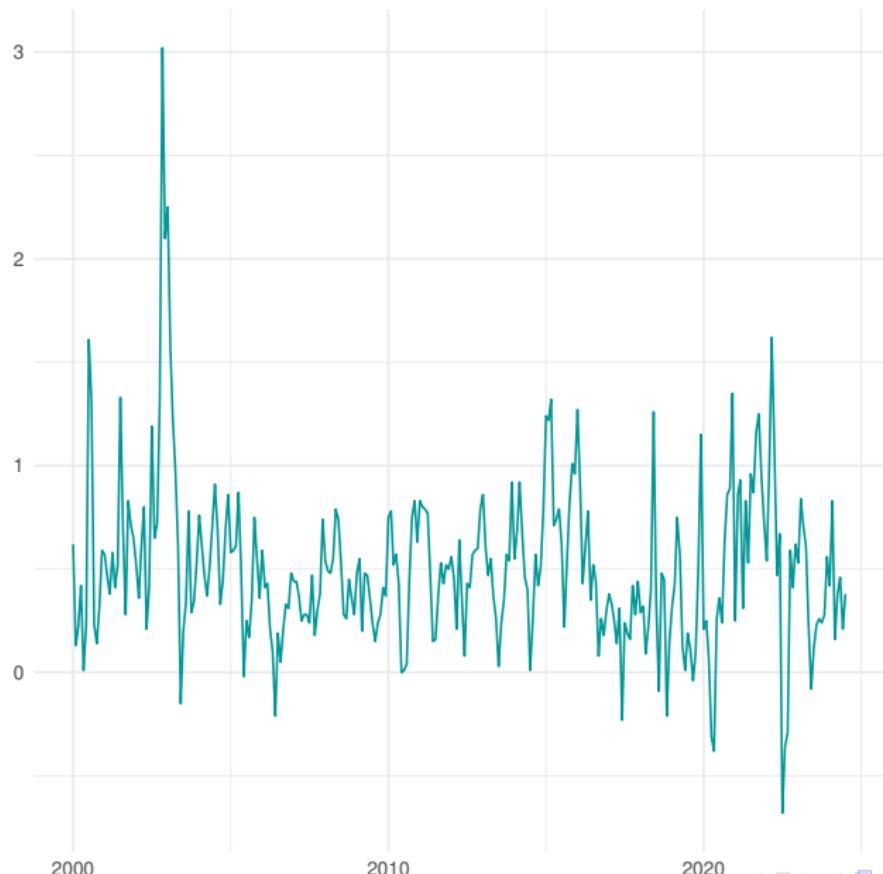
BOVESPA Index daily (closing) from 05/01/2000 to 09/08/2024.



Daily returns USD/BRL, IBOVESPA and Dow Jones



Monthly Extended National Consumer Price Index (IPCA) from jan 2000 to jul 2024.



Some Stylized Facts

- ▶ Returns tend to show heavier tails with a higher peak around the mean.
- ▶ Returns tend to show different variabilities along time.
- ▶ Returns variability tend to appear in clusters. Larger returns tend to be followed by larger returns. Likewise for small returns.
- ▶ Returns variability tend to increase more after a price decrease (negative return) than after a price increase of the same magnitude (leverage effect).

Read Chapter 1 of Prado & West