# Welcome to Forecasting Using R

FORECASTING IN R



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## What you will learn

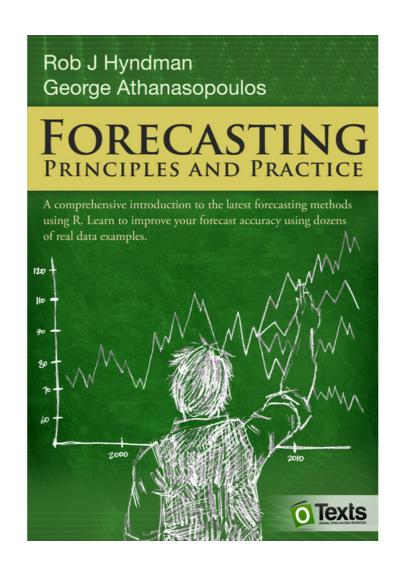
- Exploring and visualizing time series
- Simple benchmark methods for forecasting
- Exponential smoothing and ARIMA models
- Advanced forecasting methods
- Measuring forecast accuracy
- Choosing the best method

#### Course textbook

Hyndman, R. J. & Athanasopoulos, G. (2017)

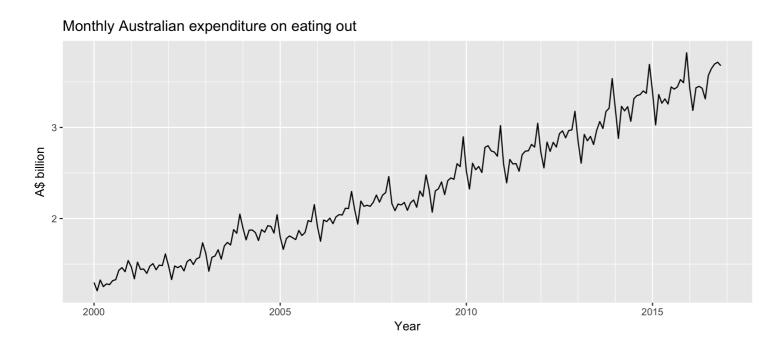
Forecasting: principles and practice, 2nd edition

- Free and online at OTexts.org/fpp2/
- Data sets in associated R package fpp2
- R code for all examples



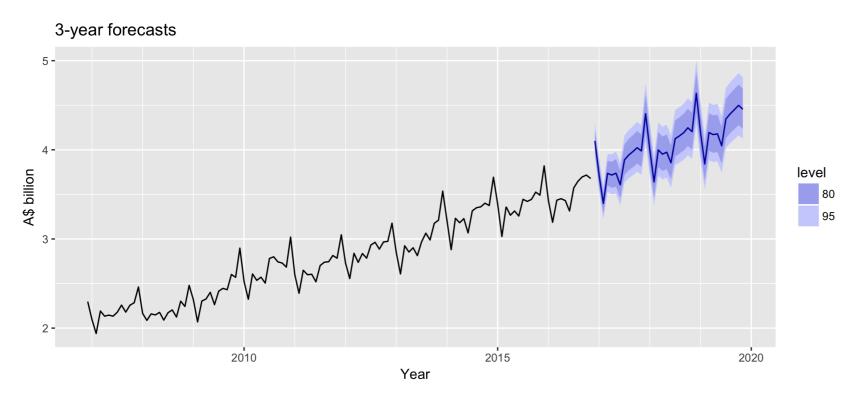
#### Time series data

- Series of data observed over time
- Eg.: Daily IBM stock prices, monthly rainfall in London,...



Forecasting is estimating how the sequence of observations will continue into the future.

# Forecasts of monthly Australian expenditure on eating out



- What forecasting methods are available that take account of trend, seasonality and other features of the data?
- How to measure the accuracy of your forecasts?
- How to choose a good forecasting model?

# Let's practice!

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# Trends, seasonality, and cyclicity

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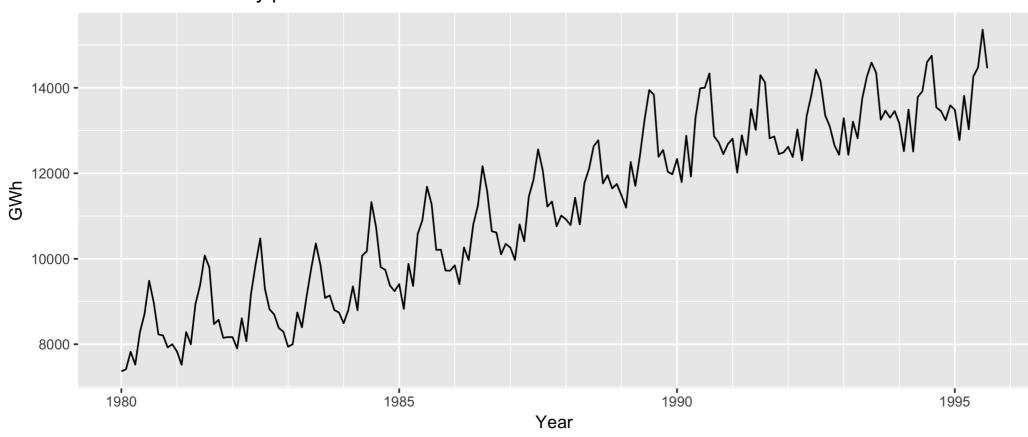
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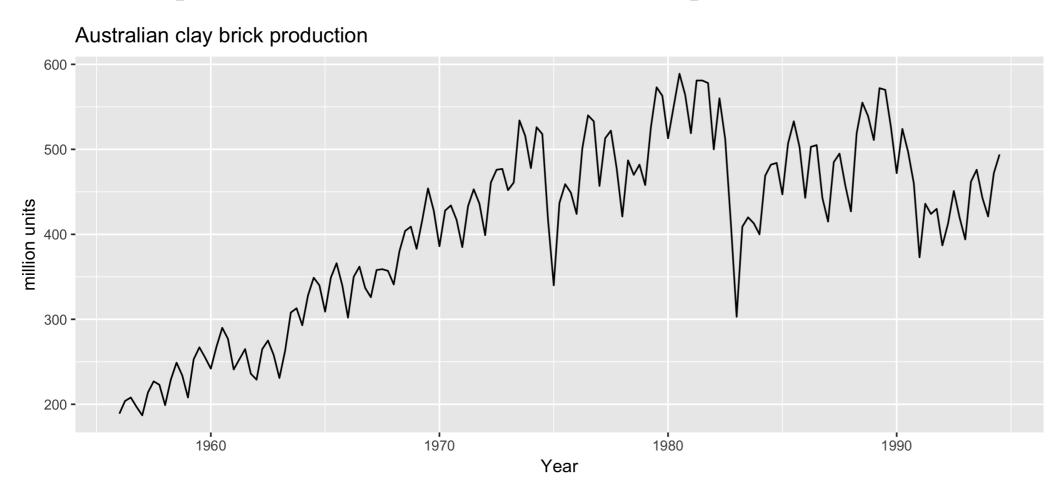
# Time series patterns

Pattern	Description
Trend	A pattern exists involving a long-term increase OR decrease in the data
Seasonal	A periodic pattern exists due to the calendar (e.g., the quarter, month, or day of the week)
Cyclic	A pattern exists where the data exhibits rises and falls that are not of fixed period (duration usually of at least 2 years)

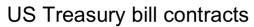


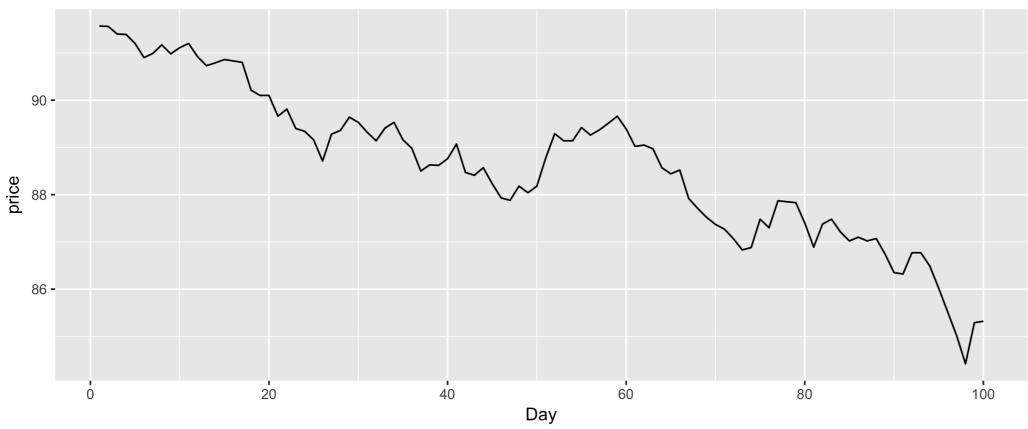






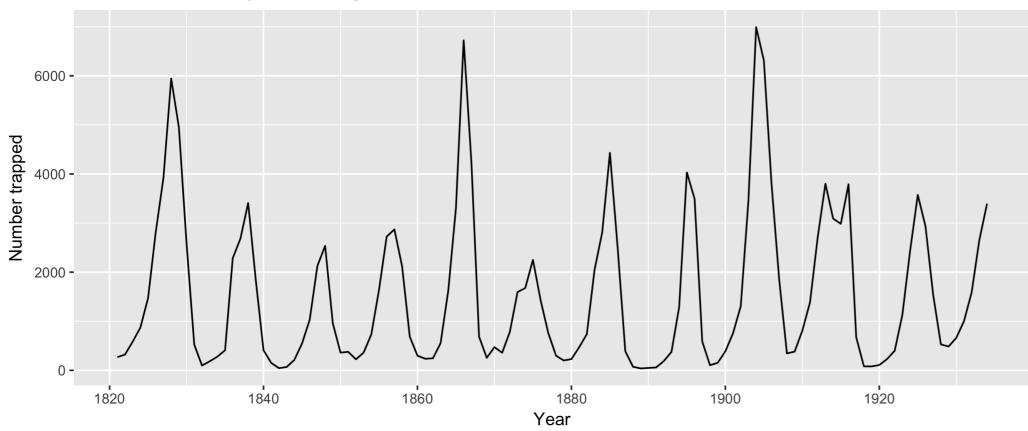














## Seasonal or cyclic?

Differences between seasonal and cyclic patterns:

- Seasonal pattern constant length vs. cyclic pattern variable length
- Average length of cycle longer than length of seasonal pattern
- Magnitude of cycle more variable than magnitude of seasonal pattern

The timing of peaks and troughs is predictable with seasonal data, but unpredictable in the long term with cyclic data.

# Let's practice!

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# White noise

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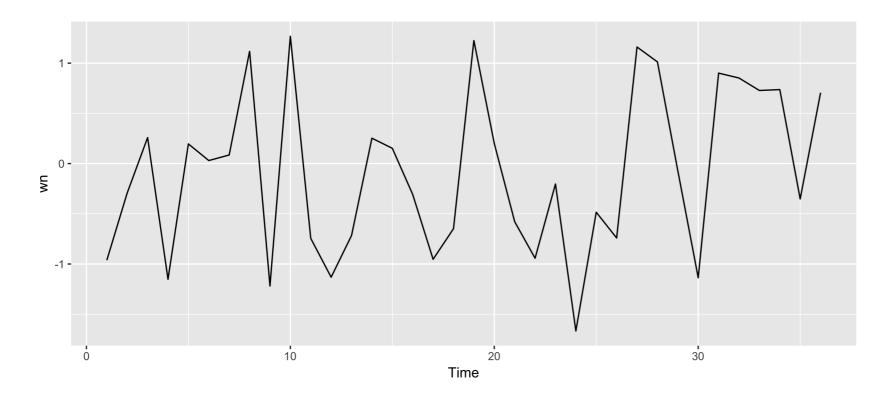
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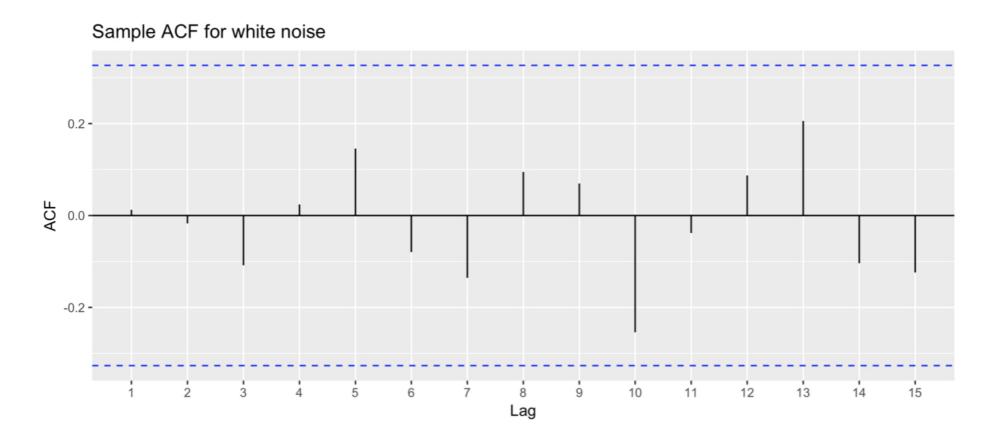
#### White noise

```
set.seed(3)  # Reproducibility
wn <- ts(rnorm(36))  # White noise
autoplot(wn)  # Plot!</pre>
```



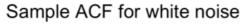
"White noise" is just a time series of iid data

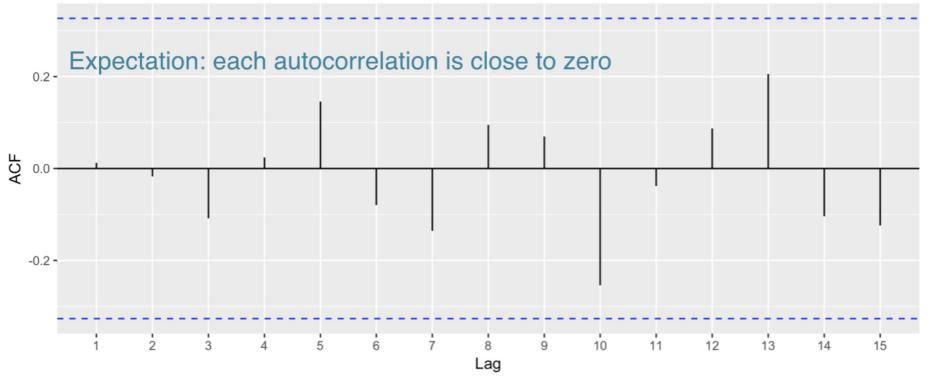
```
ggAcf(wn) +
    ggtitle("Sample ACF for white noise")
```





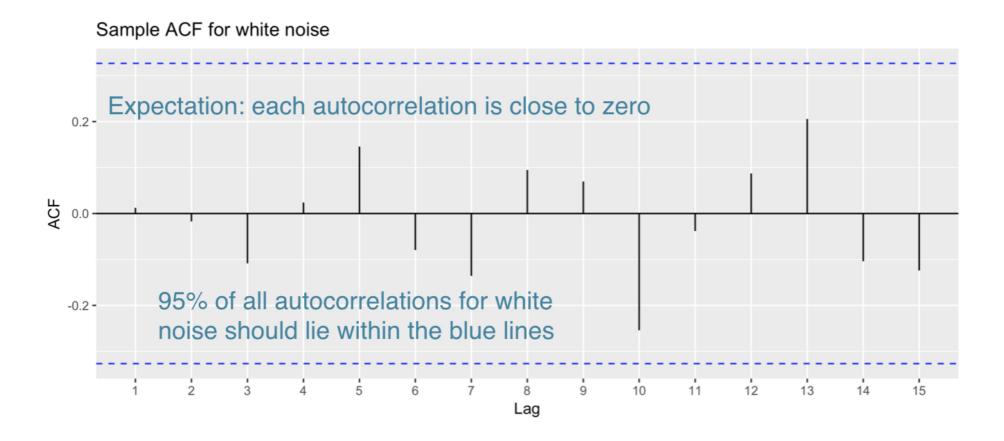
```
ggAcf(wn) +
   ggtitle("Sample ACF for white noise")
```





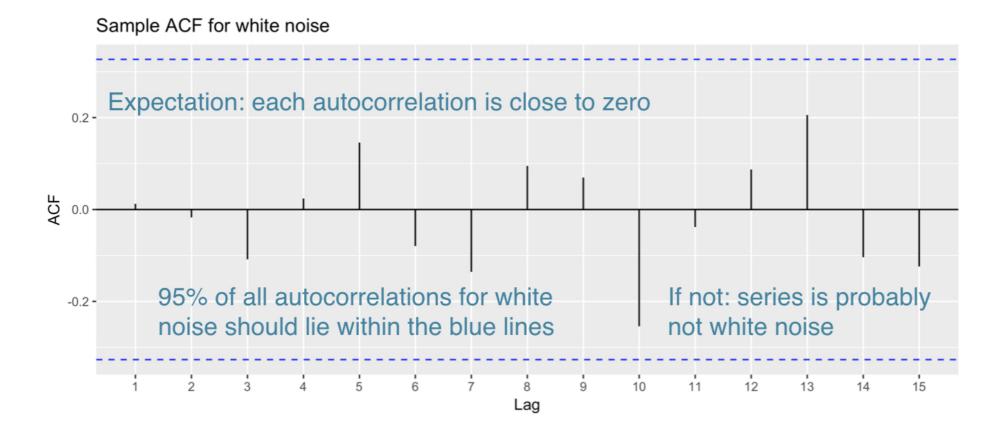


```
ggAcf(wn) +
    ggtitle("Sample ACF for white noise")
```





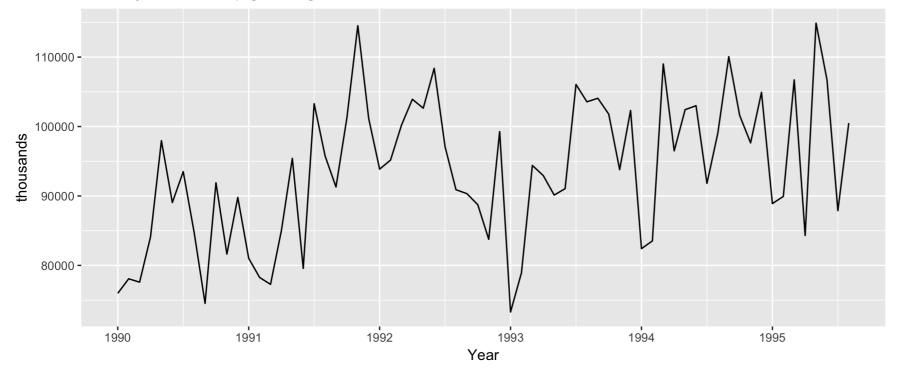
```
ggAcf(wn) +
    ggtitle("Sample ACF for white noise")
```





```
autoplot(pigs/1000) +
   xlab("Year") +
   ylab("thousands") +
   ggtitle("Monthly number of pigs slaughtered in Victoria")
```

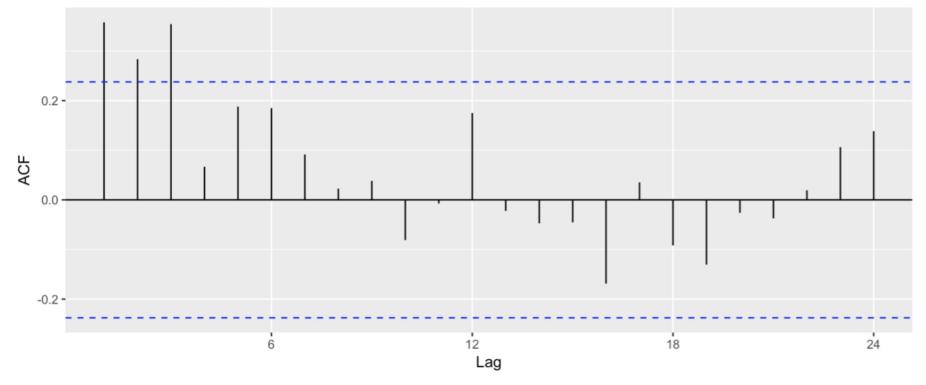
#### Monthly number of pigs slaughtered in Victoria





```
ggAcf(pigs) +
  ggtitle("ACF of monthly pigs slaughtered
    in Victoria")
```

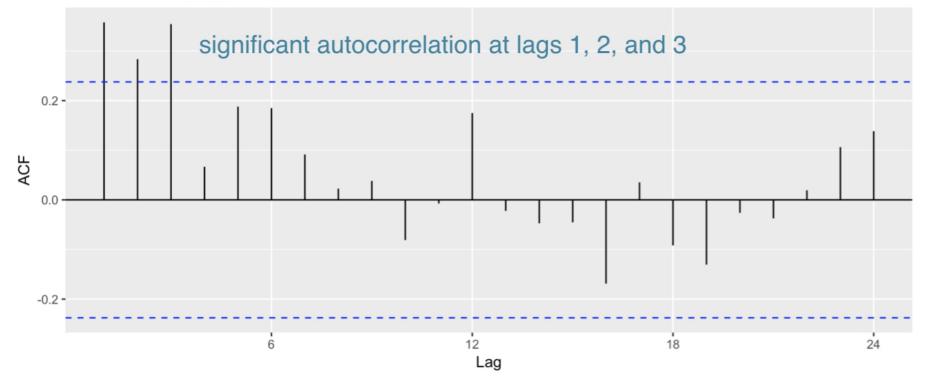






```
ggAcf(pigs) +
  ggtitle("ACF of monthly pigs slaughtered
    in Victoria")
```

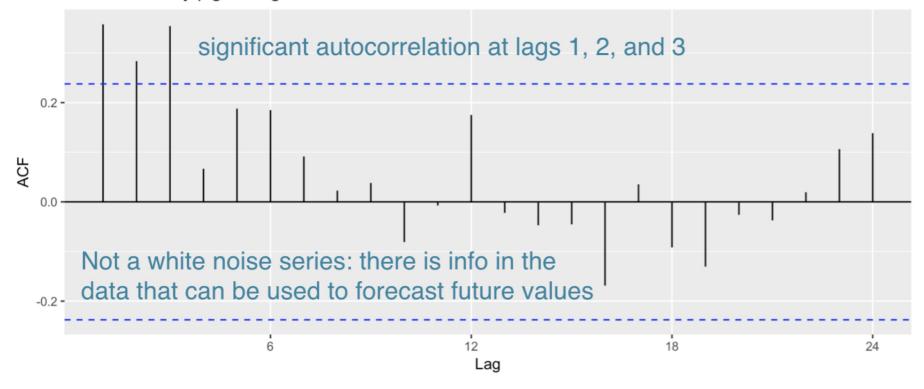






```
ggAcf(pigs) +
  ggtitle("ACF of monthly pigs slaughtered
    in Victoria")
```







# Ljung-Box test

The Ljung-Box test considers the first h autocorrelation values together.

A significant test (small p-value) indicates the data are probably not white noise.

```
Box.test(pigs, lag = 24, fitdf = 0, type = "Lj")
```

```
Box-Ljung test
data: pigs
X-squared = 634.15, df = 24, p-value < 2.2e-16
```

### White noise summary

- White noise is a time series that is purely random
- We can test for white noise by looking at an ACF plot or by doing a Ljung-Box test

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