Welcome to **Forecasting Using R** FORECASTING IN R



Rob J. Hyndman Professor of Statistics at Monash University



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



R datacamp

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?

R datacamp

Let's practice!



Trends, seasonality, and cyclicity



Rob J. Hyndman Professor of Statistics at Monash University



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)



Australian electricity production



R datacamp

Australian clay brick production



R datacamp

US Treasury bill contracts



R datacamp

Annual Canadian lynx trappings



R datacamp

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!



White noise

FORECASTING IN R



Rob J. Hyndman Professor of Statistics at Monash University



White noise

<pre>set.seed(3)</pre>	#
wn <- ts(rnorm(36))	#
autoplot(wn)	#

tacamp

Reproducibility

)) # White noise

Plot!



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

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

Sample ACF for white noise

tacamp



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

Sample ACF for white noise



R datacamp

ggAcf(wn) +

ggtitle("Sample ACF for white noise")





R datacamp

ggAcf(wn) +

ggtitle("Sample ACF for white noise")





```
pigs <- window(pigs, start=1990)</pre>
```

```
autoplot(pigs/1000) +
```

```
xlab("Year") +
```

```
ylab("thousands") +
```

ggtitle("Monthly number of pigs slaughtered in Victoria")



Monthly number of pigs slaughtered in Victoria



R datacamp



datacamp



R datacamp

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



Let's practice!

