

# Class 1: Introduction to Time Series Models

Andrew Parnell  
`andrew.parnell@mu.ie`



<https://andrewcparnell.github.io/IntroTSA/>

PRESS RECORD

# Introductions

- ▶ Tell us who you are, what you are working on, and what you hope to get out of the week
- ▶ Timetable for the week
- ▶ A quick note about pre-requisites

## How this course works

- ▶ This course lives on GitHub, at [github.com/andrewcparnell/ecots](https://github.com/andrewcparnell/ecots) which means anyone can see the slides, code, etc, and make comments on it
- ▶ The timetable html document provides links to all the pdf slides, handouts, data and practicals
- ▶ Let me know if you spot mistakes, as these can be easily updated on the GitHub page
- ▶ There is an issues page if you want to ask questions

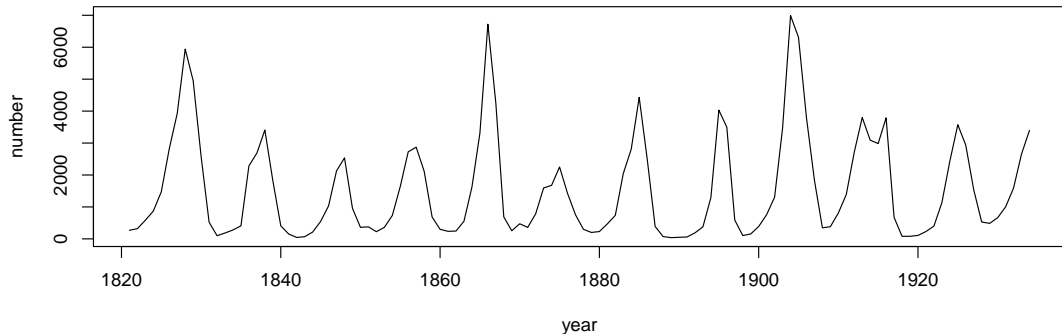
## R code, slides, and practicals

- ▶ All the slides and practicals are available in pdf format for you to annotate
- ▶ In the background, the slides and the practicals are written in Rmarkdown format, which means you can load them up in Rstudio and see how everything was created
- ▶ When you have spare time, feel free to load up the .Rmd files and run the code in the background

## R code in slides

- ▶ Many of the slides contain R code and output (some of which may be hidden in the .Rmd file)
- ▶ An example:

```
lynx = read.csv(file = '../data/lynx.csv')  
with(lynx, plot(year, number, type = 'l'))
```



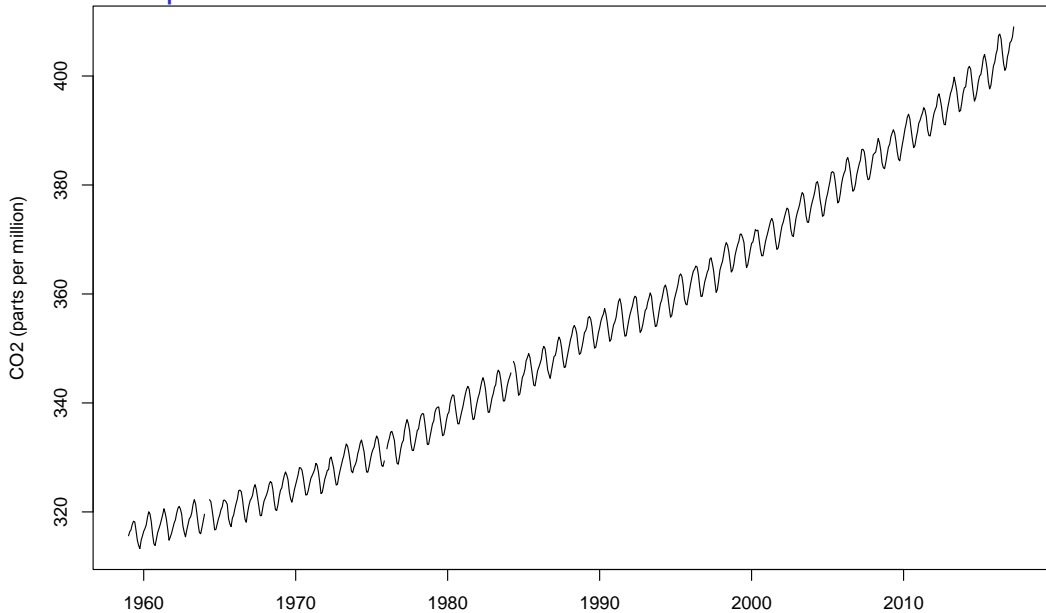
## Course format and other details

- ▶ Lectures will take place in the morning, practical classes in the afternoon
- ▶ Please ask lots of questions
- ▶ Some good books:
  - ▶ *Forecasting: Principles and Practice* by Hyndman and Athanasopoulos
  - ▶ *Hierarchical Modeling and Inference in Ecology* by Royle and Dorazio
  - ▶ *Bayesian Methods for Ecology*. by McCarthy
  - ▶ *Bayesian Data Analysis* by Gelman et al
- ▶ Looking for data? Try the `tsdl` R package
- ▶ (see also sources in Practical 3)

## What is a time series?

- ▶ A time series is any set of data where the response variable is measured over time
- ▶ There may be other variables included too (covariates)
- ▶ Time may be discrete (1, 2, 3, 4, ...) or continuous (1.7, 2.53, 7.12, ...)
- ▶ There may be missing values or outliers
- ▶ Occasionally there may be more than one response variable (multivariate time series)

## A time series plot: CO2 data





## General features of a time series

- ▶ Trend: long term behaviour. May be a straight line or something more complicated
- ▶ Seasonal: repeated behaviour. May be yearly, monthly, daily, etc. Likely to be dependent on the time resolution
- ▶ Error: Leftover uncertainty beyond the trend and seasonal behaviour. May have interesting statistical patterns.

## Writing time series mathematically

- ▶ If we write  $y_t$  as the value of the response variable at time  $t$  then the series can be *decomposed* as:

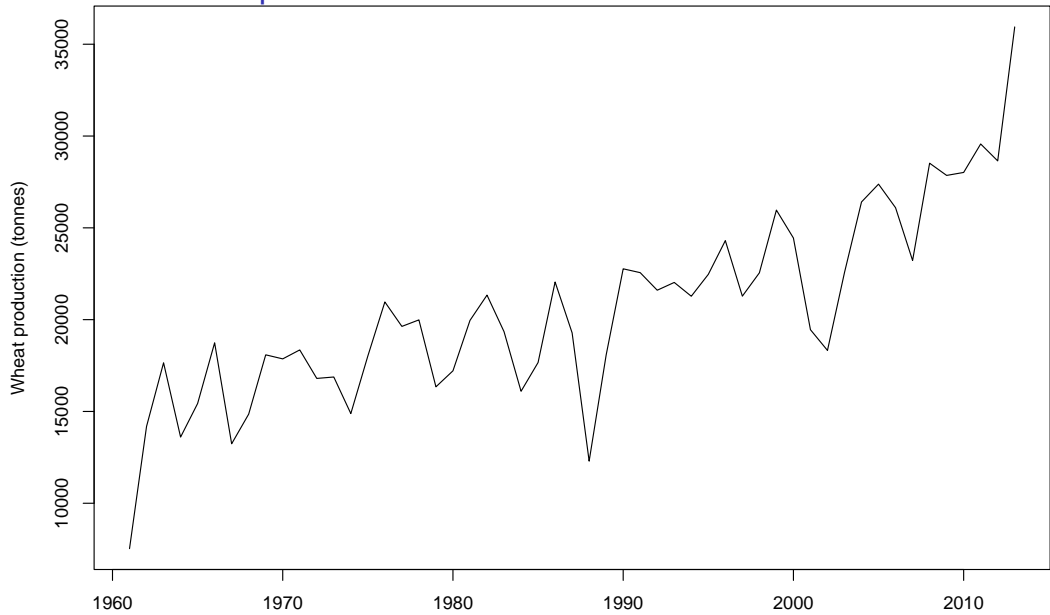
$$y_t = \text{trend}_t + \text{seasonality}_t + \text{error}_t$$

- ▶ Most time series models concentrate on the error structure
- ▶ Time Series analysis is usually harder if you need to identify the seasonality too

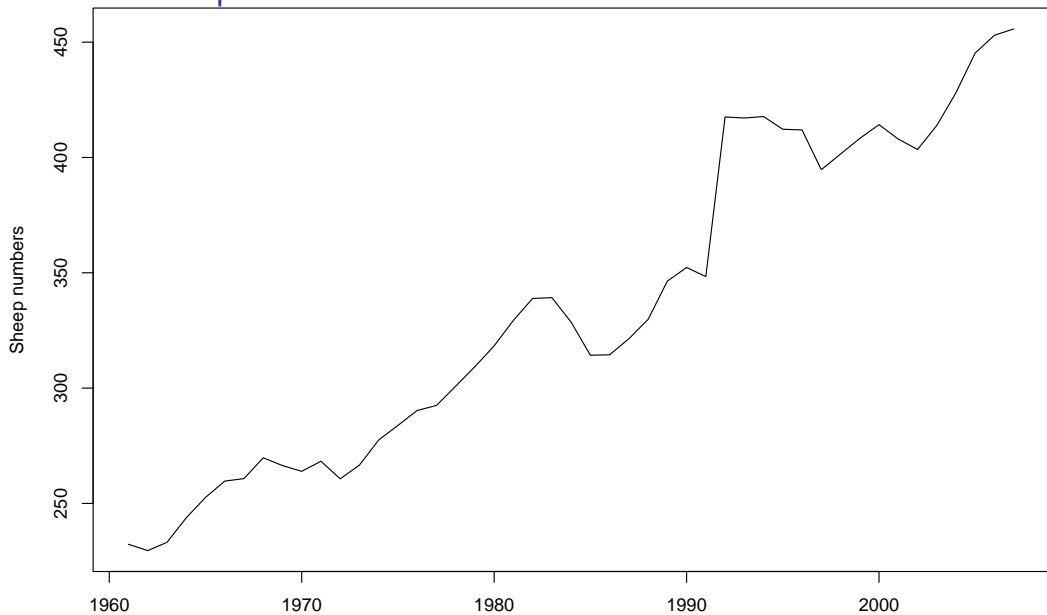
## Participation exercise

- ▶ I want you to look at the time series on the following slides and...
- ▶ ... identify the trend. Is it linear or non-linear?
- ▶ ... identify the seasonality (if any). Can you estimate the frequency?
- ▶ ... look at the residual errors after accounting (in your head) for trend and seasonality. Can you spot any patterns or strange observations?

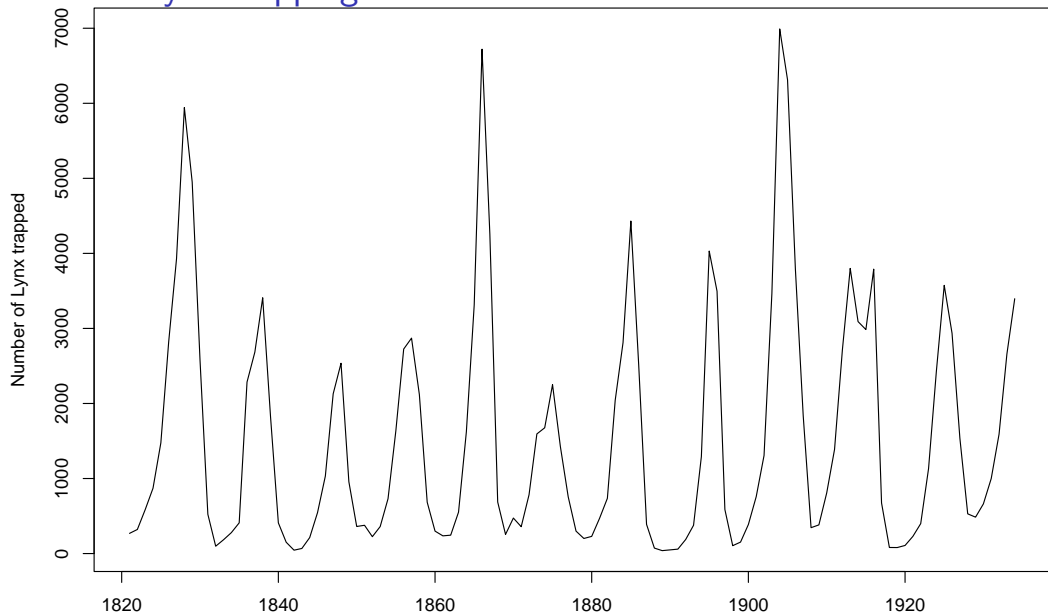
## Data set 1: Wheat production in Canada



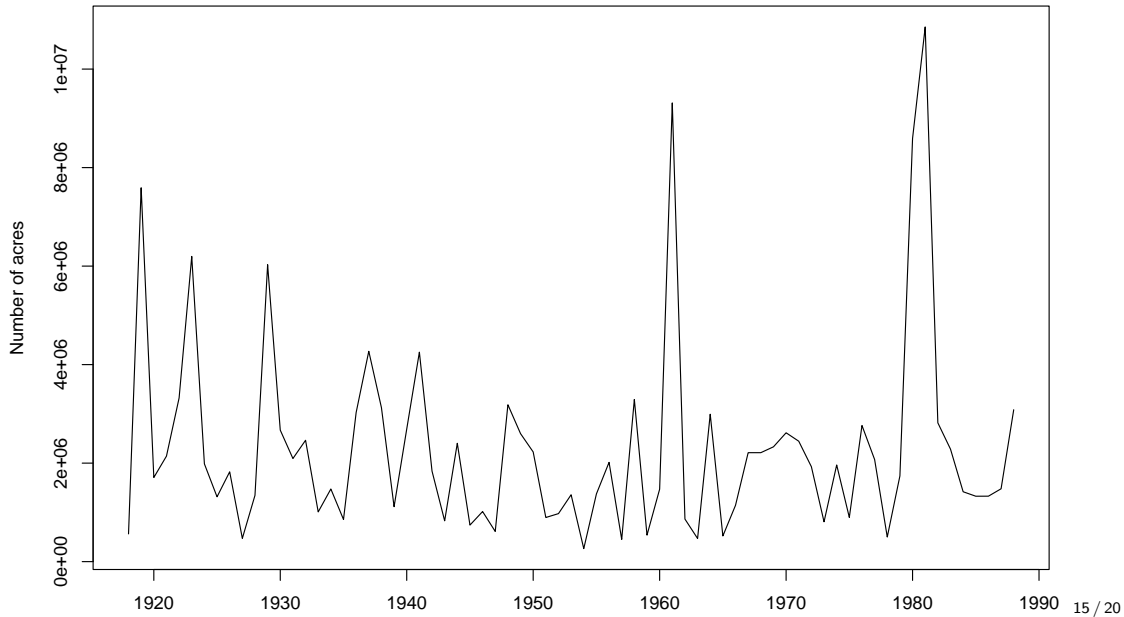
## Data set 2: Sheep numbers in Asia



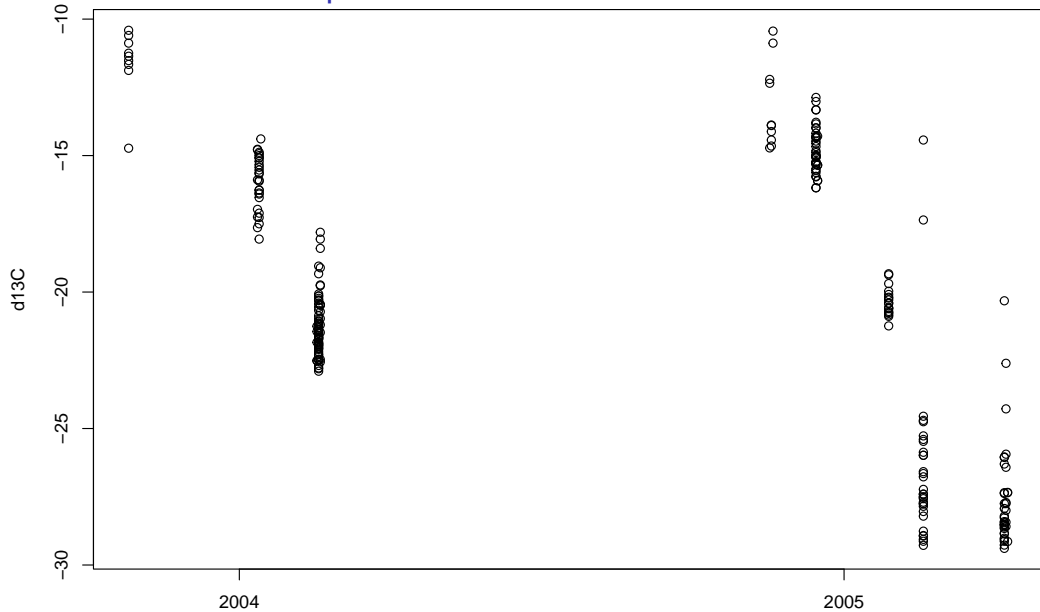
## Data set 3: Lynx trappings in Canada



## Data set 4: Forest fires in Canada

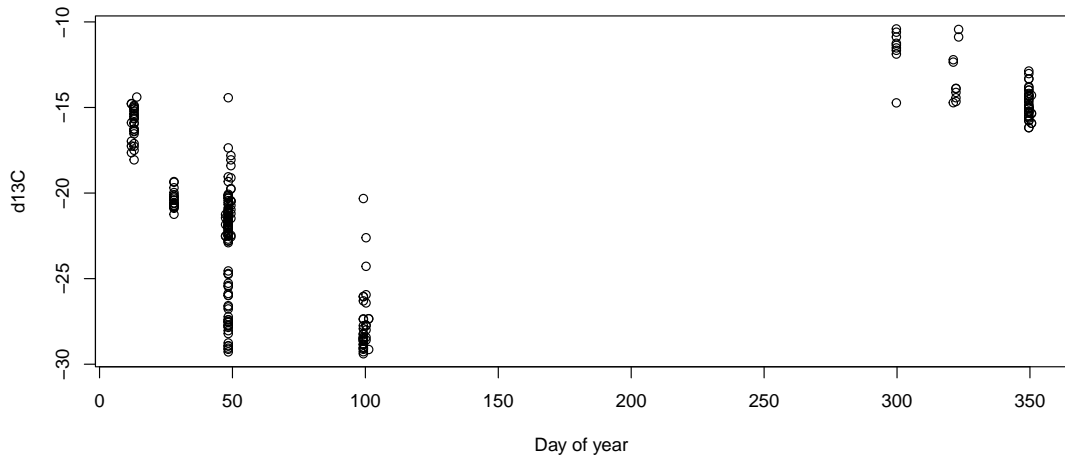


## Data set 5: Geese isotopes

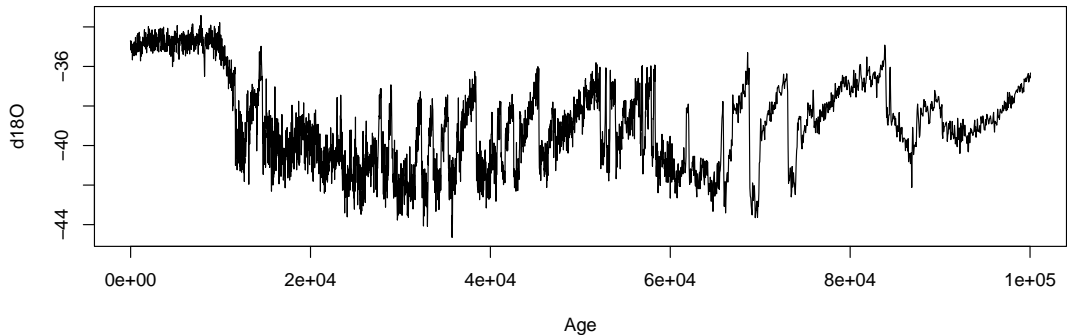




## Data set 5: Geese isotopes again



## Data set 6: Oxygen isotopes in ice



```
table(diff(ice$Age))
```

```
##
```

```
##    20    40    50    60   100   120
```

```
## 3035     2  779     2     1     1
```

# Goals of time series analysis

- ▶ Predict future values of the response variable
- ▶ Interpolate or smooth the response variable for missing or non-measured times
- ▶ Explain which factors are causing the time series to change
- ▶ Understand the underlying behaviour of the time series

# Summary

This course takes a practical approach, and should help you:

- ▶ Understand modern time series modelling techniques
- ▶ Get and use tools for thinking about and dealing with uncertainty
- ▶ Fit time series models, and make predictions
- ▶ Understand your time series data, and the process that generates it