Class 1: An introduction to Bayesian Hierarchical Modelling

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https://andrewcparnell.github.io/bhm_course

PRESS RECORD

Let's get started

- ► Introduction from Oliver Hooker, PR Statistics
- ► Tell me:
 - who you are,
 - where you are from,
 - your previous experience in working with R and regression models,
 - what you are working on,
 - what you want to get out of the week
 - what you are most looking forward to when we can all go outside again
- ► Timetable for the week
- Pre-requisites

How this course works

- ► This course lives on GitHub, which means anyone can see the slides, code, etc, and make comments on it
- ► The timetable document (index.html) provides links to all the pdf slides and practicals
- ► The slides and the practicals are all written in Rmarkdown format, which means you can load them up in Rstudio and see how everything was created
- Let me know if you spot mistakes, as these can be easily updated on the GitHub page
- ► There is a bhm_course.Rproj R project file from which you should be able to run all the code

Copyright statement

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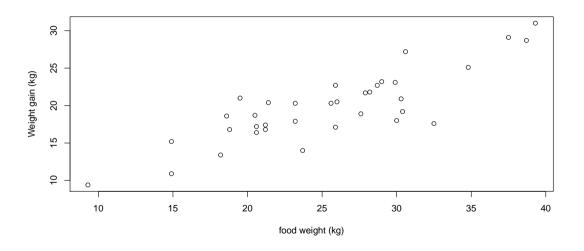
Course format and other details

- ▶ Lectures will take place in the morning via Zoom, practical classes in the afternoon
- ▶ In the practical classes I will go round the room asking people how they are getting on
- ▶ If you want to send me a private message use Slack
- Please ask lots of questions, but MUTE YOUR MICROPHONE when not asking them
- Some good books:
 - Data Analysis using Regression and Hierarchical Models by Gelman and Hill
 - Bayesian Data Analysis by Gelman et al

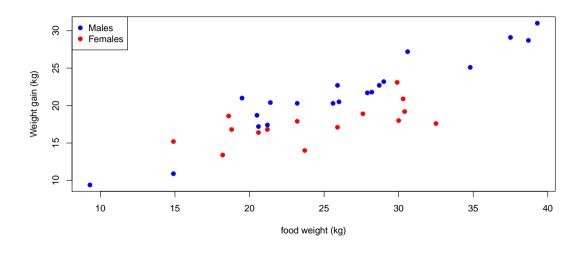
What is a Bayesian hierarchical model?

- ► A model is just a representation/approximation of the real world, here expressed in equations
- ► Hierarchical means that the model is built up in *ordered layers* which makes it easier to fit very complex models
- ▶ Bayesian means the model involves both a *likelihood* and a *prior* probability distribution (more on this tomorrow)

Thinking hierarchically: example 1



More information:



Example 2: 8 Schools

We have 8 schools in a region, with a relative performance score (column score) compared to the national average and a standard deviation (sigma) based on 3 repeated visits

```
## 1 1 28 15
## 2 2 8 15
## 3 3 -3 16
## 4 4 4 7 11
## 5 5 -1 9
## 6 6 1 11
## 7 7 18 10
## 8 8 12 18
```

- If you had to pick an overall score for this region how would you calculate it?
- ▶ If you had to guess the score of a new measurement for school 1 what value would you use?

Example 3: Earnings data

1192 observations on earnings (in USD) and various measurements about ethnicity, age, height, etc

```
earn age eth height height cm
                                           v x x centered
## 1 50000
                            187.96 10.81978 74
                                                 6.932011
## 2 60000
                            167.64 11.00210 66
                                               -1.067989
## 3 30000
                            162.56 10.30895 64
                                                -3.067989
## 4 51000
                            160.02 10.83958 63
                                                -4.067989
                            162.56 9.10498 64
     9000
                                                 -3.067989
## 6 29000
                            157.48 10.27505 62
                                                -5.067989
```

- Does height affect earnings?
- ▶ Are there different rates of change for different groups (e.g. age/ethnic groups)?

Example 4: Swiss Willow Tit data

3 replicate measurements on whether Swiss Willow Tits were found with covariates on forest cover and elevation

```
## rep.1 rep.2 rep.3 c.2 c.3 elev forest dur.1 dur.2 dur.3 length alt
## 1 0 0 0 0 0 420 3 240 58 73 6.2 Low
## 2 0 0 0 0 0 450 21 160 39 62 5.1 Low
## 3 0 0 0 0 0 1050 32 120 47 74 4.3 Med
## 4 0 0 0 0 0 0 1110 35 180 44 71 5.4 Med
## 5 0 0 0 0 0 510 2 210 56 73 3.6 Low
## 6 0 0 0 0 630 60 150 56 73 6.1 Low
```

- ▶ How do the covariates affect the chance of finding the birds?
- Are these effects linear?
- What do we do with the missing data?

More data sets in the data directory

- ► The data directory contains a few more data sets which we will play with throughout the week
- The data_descriptions.txt file shows what they contain
- ▶ If you have some spare time it's worth loading them in, exploring relationships, and fitting some simple models

Summary

- In hierarchical models we avoid fitting models separately as much as possible
- By fitting models together we borrow strength from the different groups in the data and reduce uncertainty
- ▶ Bayesian models allow us to incorporate all the available data into providing information on the question we want to answer