# MPP-E1180 Lecture 1: Introduction to the Course

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- Room: 1.52
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# Objectives for the week

- Introduce the course goals, plan, and expectations/assessment
- Introduce Collaborative & Reproducible Data Analysis
- Set up computational research environment

# Objectives for the course

**Collaboratively** and **reproducibly**

- Gather/clean social data
- Analyse it
- Present results (in a variety of mediums)

Learn how to actually **do** data analysis using **best practices**

# Motivation: Academic

- Skills needed to do **original quantitative research** for your **thesis**.
  - The final project will be a **trial version** of your thesis.
- State-of-the-art tools needed for **future high-level academic research**.
  - Take advantage of new data sources.
  - Avoid effort duplication.
  - Make your research reproducible.
  - Present your results in multiple forums.

# Motivation: Government

- Government agencies are increasingly adopting the technologies and methods of open data science.

# Motivation: Government

- Public data is being made increasingly **accessible**.
  - e.g. World Bank Development Indicators, GovData Germany, data.gov.uk, New York City, data.gov
- Governments rely on data analysis for evidence based decision-making.
  - Tools of open data analysis enable better use of data **within** and **between** government actors.
  - Governments can take advantage of analyses done by **third parties**.

# Motivation: Government

- They are also **sharing** and **collaboratively** developing code; **reducing development costs** and **improving applications**.
- Version control to **increase engagement with the legislative process**.
  - San Francisco laws are now forkable.

# Motivation: Business

- Data analysis and R programming skills in particular are **highly valued** in businesses such as finance and management.

**AVERAGE SALARY FOR High Paying Skills and Experience**

| SKILL | 2013 | YR/YR CHANGE |
|---|---|---|
| R | $ 115,531 | n/a |
| NoSQL | $ 114,796 | 1.6% |
| MapReduce | $ 114,396 | n/a |
| PMBok | $ 112,382 | 1.3% |
| Cassandra | $ 112,382 | n/a |
| Omnigraffle | $ 111,039 | 0.3% |
| Pig | $ 109,561 | n/a |
| SOA (Service Oriented Architecture) | $ 108,997 | -0.5% |
| Hadoop | $ 108,669 | -5.6% |
| Mongo DB | $ 107,825 | -0.4% |

# Why Collaborative?

- Research is collaborative (even if you don't know it).
- Need tools and shared best practices to enable effective collaboration between **explicit research partners**.
- Need tools and shared best practices to enable collaboration between researchers who are **not explicitly** working together often in **unexpected ways**.
  - **Avoids effort duplication**
  - Enables **cumulative knowledge development**
- Tools for collaboration tend to enhance **reproducibility**.

# What is reproducibility?

**Really reproducible** research (Peng 2011, 1226):

> *the data and code used to make a finding are available and they are sufficient for an independent researcher to recreate the finding.*

- In practice reproduciblity is enhanced by **literate programming** where the data, analysis, and presentation of the results are 'weaved' or 'knitted' together.
  - Make available the research, **not just the advertising** for the findings (e.g. papers, book).

# Reproducibility vs. Replication?

**Reproducibility**: an independent study makes the same findings using the **same data** and **code** as the original researchers.

**Replicability**: an independent study makes the same conclusions as the original using **other data, code, and even methods**, i.e. independent verification.

# Reproducibility vs. Replication?

**''A study can be reproducible and still be wrong"** Peng 2014.

E.g. a finding that is statistically significant in one study may remain statistically significant when reproduced using the original data/code, but **replication studies are unable to find a similar result**.

The original finding could just have been noise.

# Why reproducibility?

- **Replication** is the ''**ultimate standard**" for judging scientific claims (Peng 2011).
- **Reproducibility**
  - **Enhances replication** (other researchers can understand how an analysis was actually done)
  - Is a **minimum standard** for judging scientific claims when replication is not possible.

# Why reproducibility?

- Reproducibility helps **avoid effort duplication**:
- Others don't need to waste time:
  - Gathering data that has already been gathered.
  - Discovering procedures that have already been discovered.
- This enables **effort duplication**.

# Why reproducibility?

- Reproducibility also makes it possible to **find and correct errors**.
  - Recent example: translation errors in the World Values Survey.
- Data errors can cause spurious findings that ultimately **waste researchers time**, because they try to explain 'wrong' findings.

# Why reproducibility?

- **Higher research impact**
  - Reproducible research is likely to be more **useful for other researchers**. They can use your data and learn from your code and methods.
- **Better work habits**
  - If you are thinking about reproducibility from the beginning your files will be **better organised** and your work will be **better documented**.
  - This allows you to **build on your own work** more effectively.

# Reproducible Workflow

# Practical Tips for Reproducible Research

- Document Everything!
- Everything is a (text) file.
- All files should be human readable.
- Explicitly tie your files together.
- Have a plan to organise, store, and make your files available.

# Course Prerequisites

- **Introductory-level statistics**
  - Basic descriptive statistics (e.g. data types, ways of describing distributions)
  - Basic inferential statistics: (significance testing, linear regression)
  - Exposure to statistics software (e.g. SPSS, STATA)
- Knowledge of particular software or computer programming is **not expected**
- **Patience**
  - Work hard so you can be lazy.

# Course Outline (1)

**Part I: Motivation and Getting Started**

- Introduction to the Course
- Files, Files Structures, Version Control, and Collaboration
- Introduction to the R Programming Language

**Part II Markup Languages and Literate Programming**

- Introduction to Markup Languages and Literate Programming (1)
- Introduction to Markup Languages and Literate Programming (2)

# Course Outline (2)

**Part III: Data Gathering, Trasformations, and Analysis**

- Automatic Data Gathering via Curl, API Packages + Cleaning
- Automatic Data Gathering via Web Scraping
- Statistical Modelling with R

**Part IV: Communicating Results from Statistical Analyses**

- Automatic Table Generation and Static Visualisation
- Dynamic Visualisation

**Part V: Collaborative Research Project**

# Typical Class Plan

- $\sim$ 1 hour lecture
- $\sim$ 1 hour seminar
  - **Apply** what we learned in the lecture/readings to achieve specific goals, i.e. **no set pattern** to copy by rote.
  - **Pair programming**: work together with others to acheive these goals.
  - **Documentation**: document your work with Git/GitHub.
    - Your seminar work should be **reproducible**.
    - It should be **useful** to your **future self** and **others**.

# Assessment

- 3 Pair Assignments (Weeks 3, 6, 9)
  - 10% each
- Collaborative Research Project (Presentation: Week 12, Website/Paper: Exam Week)
  - 50%
- Attendance & Active Participation
  - 20%
- No traditional midterm or final exam

# Assessment Details (1)

- All assignments must be developed and submitted electronically on GitHub.
- Late assignments: -10% every day that the assignment is late.
- All assignments must be completed in **pairs**.
  - Each pair member receives the same score
  - Exception: very large discrepancy in contributor statistics

# Assessment Details (2)

- All assignments must be **reproducible**.
- **Due**: Midnight on Friday of the week it is due.
- More details will be given on the specific pair assignments/research project in future classes.

# Assessment (attendance, participation)

- Usual Hertie Rules for attendance (examination rules §4)
- Participation:
  - **Traditional Participation**, e.g. engaging in class discussions, doing readings
  - **Non-Traditional Participation**: pair programming in seminars, document your seminar work on GitHub, pull request to the course repository (syllabus/lecture slides) and other groups' projects

# Syllabus & Lecture Slides

`https://github.com/HertieDataScience2014/SyllabusAndLectures`

**Syllabus** (README.md)

- The syllabus will be **updated**. **Check regularly**.
  - Course **difficulty** is **monotonically decreasing** from the original (11 September) baseline.

**Lecture Slides** (LectureSlides/)

- Accessible as both HTML (recommended) or PDF.
- Slides will be **optimized for the web**.

# Reading

**Core Texts**

- Gandrud, Christopher. 2013. *Reproducible Research with R and RStudio*. Chapman & Hall/CRC Press, Oxford. (RRRR)
- Crawley, Michael J. 2005. *Statistics: An Introduction Using R*. John Wiley and Sons Ltd., Chichester.

Both are available in the library.

**Other readings** generally available online (see syllabus) or I will make a copy available.

# Seminar to-do

- **Objective**: setup software (all software is free)
- **Highly recommend**: use your own laptop

# Modern Web browser

- Make sure you have a modern web browser, e.g.:
  - Chrome

# GitHub

Setup Git/GitHub for version control, collaboration, and remotely storing your files.

- Set up (free) GitHub account: `https://github.com/join`
- **Give me your GitHub username** so that I can add you to the **HertieDataScience2014** group (`https://github.com/HertieDataScience2014`).
- Install GitHub application:
  - Mac: `https://mac.github.com/`
  - Windows: `https://windows.github.com/`

# Statistics software

- **Install** software:
  - R: `http://cran.rstudio.com/`
  - RStudio (dev build): `http://www.rstudio.org/download/daily/desktop/`
- Make sure that you can install R packages:

```
# Install the ggplot2 package
install.packages('ggplot2')

# Check to see if it loads properly
library(ggolot2)

ggplot(mtcars, aes(hp, mpg)) + geom_point()
```

# LaTeX

- Install a LaTeX distribution. Creates well formatted PDF versions of your presentation documents.
  - Mac: `https://tug.org/mactex/`
  - Windows: `http://miktex.org/download`
- This is a large download, so maybe do it in your spare time.

# Pandoc

Install Pandoc. We won't use this directly, but it is needed for creating presentation documents in multiple formats.

- `http://johnmacfarlane.net/pandoc/installing.html`

# Post-Installation

Play around with the software (especially RStudio)