

### Reproducible data science techniques in actuarial work What can actuaries learn from open science and other professions?

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#### **Reproducible data science techniques for actuaries**

We often focus on the glamourous side of data science: big data sets, artificial intelligence and the replacement of humans by machine.

This presentation considers data science from a different perspective - how do data scientists work?

We make the case that actuaries can learn from the working methods used by data scientists and apply them to existing actuarial work, as well as when adopting data science techniques.

#### Content

- Case studies from UK Government and the BBC News website
- How reproducibility can help address the challenges of working in data intensive fields
- Why this is important for actuaries
- A short introduction to some of the tools available

This presentation is accompanied by <u>exercises</u> covering some of the techniques and tools introduced.



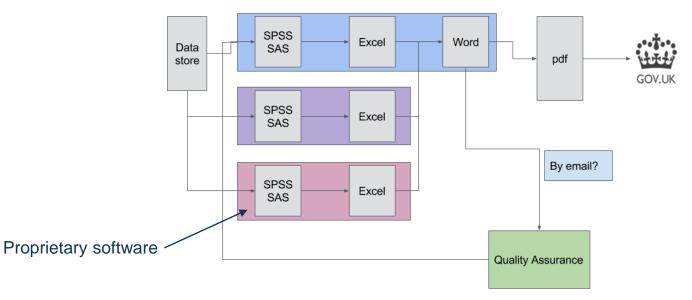
# **Case study**

#### Producing official statistics in UK Government

### **Producing official statistics in government**

Data scientists at the UK Government Digital Service have worked with a number of government departments to transform the production of official statistics.

In general, analytical processes in government involve a mixture of manual and semi-manual processes looking something like this:



#### **Government statistics – the existing process**

This process varies across government departments and projects but it typically involves:

- Extracting data from databases, spreadsheets and other sources
- Processing the data and carrying out analysis in proprietary software
- Manually copying results to Excel for further analysis and the production of tables/graphs
- Copying and pasting output to a Word document for reporting and formatting
- Converting the report to a PDF suitable for publication
- Along the way, documents might be emailed between colleagues resulting in multiple versions of files

#### What do your existing analytical processes look like? Does the above look familiar?

#### **Government statistics – problems with this approach**

The Government Digital Service recognised a number of issues with this approach:

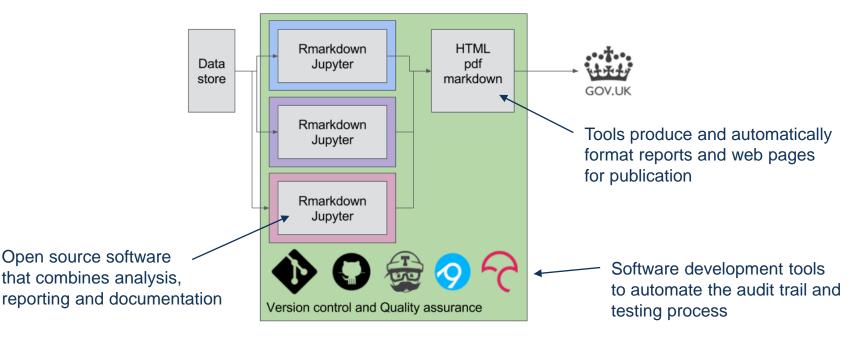
- Heavy reliance on manual processes is time consuming
- Errors in spreadsheets are common due to human error and are often difficult to catch
- Further errors can be introduced when copying data/output between software tools
- It is difficult to keep an audit trail of the process when working in a team (i.e. who did what)
- Checking and testing often happens at the end of the process, rather than embedded throughout
- Any changes made following checking/peer review require the manual process to be repeated at the risk of introducing further errors

They also identified the difficulty in reproducing the results of previous reports.

Are you confident you could easily reproduce your last board paper/client report/statistical analysis?

#### **Government statistics – their solution**

To try and solve these problems, they took ideas from software development and academia, using open source tools to devise a pipeline that reduced production time whilst arguably improving the quality of the output. It looks like this:



#### **Reproducible analytical pipelines**

The Government Digital Service call this approach Reproducible Analytical Pipelines (RAP):

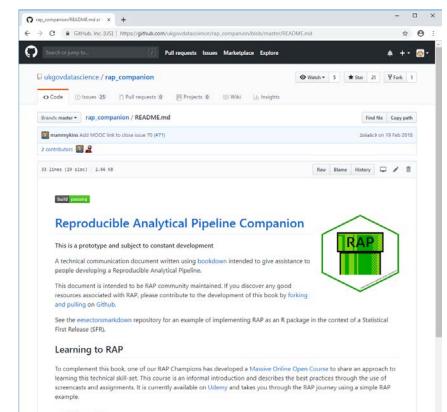
- Data is loaded into software once (no cutting and pasting!)
- One tool is used for all data processing, analysis, production of tables/graphs and report writing
- Tools and techniques from software development are used to automate testing and maintain an audit trail

With RAP, analytical teams can automate time-consuming processes of data assembly, verification and integration, generate charts and tables, and set up and populate statistical reports. The potential time savings for analysts are enormous, freeing them up to focus on the interpretation of the results. The other huge benefit comes from building a process that is fully transparent, auditable and verifiable – reducing risk and improving quality.

Matt Upson and Mat Gregory, Government Digital Service

#### **Government statistics – further reading**

- Blog posts <u>here</u> and <u>here</u> introduce the RAP project
- A free online book, the RAP Companion, is available <u>here</u> that explains the process in detail
- In addition, the data scientists involved in this project created a <u>free online course</u> at Udemy
- An example statistical release using RAP can be found on GitHub <u>here</u>



#### Collaboration

This is a community effort produced by those interested in automating the production of statistical reports in government. If you wish to contribute refer to the CONTRIBUTING.md.

Issues

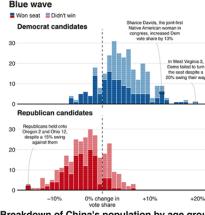


# **Case study**

#### Producing graphics for the BBC News website

### **BBC News graphics**

The BBC visual and data journalism team have used a similar approach to fundamentally change how they produce graphics for the BBC News website.



committed by young people on young people Number in each age group 2014 - 2017 Victims Suspects

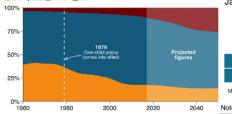
NB. Analysis excludes 3,474 victims and 2,003 suspects whose ages are unknown

Source: BBC Freedom of Information requests to UK police forces

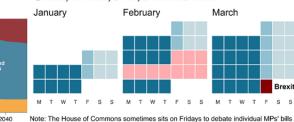
Homophobic hate crimes are mainly

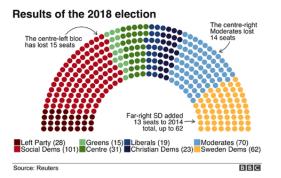
#### Breakdown of China's population by age group until Brexit Proportion of total population (1960-2050) Monday to Thur







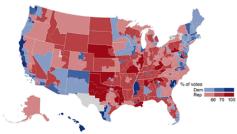




#### Democrats take the House

BBC







B B C Source: AP. Grey districts are undeclared

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## **BBC News graphics**

The statistical programming language R has been used by the BBC for data analysis for some time, however website graphics were previously produced by an in-house design team.

At the start of 2018, the BBC visual and data journalism team looked at whether R could be used for the whole process – in effect automating the design element of the process. In March 2018, they published the <u>first chart</u> that was produced from start to finish in R using the ggplot2 package.

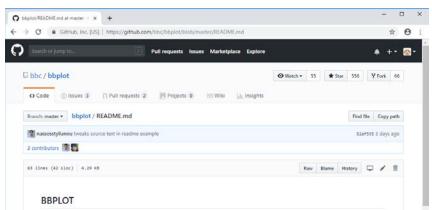
This streamlined the production of graphics and enabled more people at the BBC to easily produce graphics in their house style.

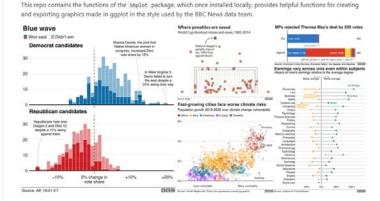
[This approach] saves a huge amount of time and effort, in particular when working with data that needs updating regularly, with **reproducibility a key requirement of our workflow**. In short, it was a game changer...

BBC Visual and Data Journalism team

## **BBC News graphics – further reading**

- Blog post <u>here</u> explains what the BBC did and what they learnt along the way
- The cookbook <u>here</u> is used by the BBC to share tips and tricks, and solutions to common problems
- The package that creates graphics in the BBC style can be found on GitHub <u>here</u>





#### Installing bbplot

bbplot is not on CRAN, so you will have to install it directly from Github using devtools .

If you do not have the devtools package installed, you will have to run the first line in the code below as well.



# **Reproducibility in actuarial work**

Why is this important for actuaries?

# What is reproducibility?

Reproducibility is the process of making code and data available so that others can easily replicate your analysis.

Data scientists achieve this by using automation and standard tools/processes that reduce the number of manual steps required in a piece of analysis. This allows them to work with larger datasets and more complex workflows.

A reproducible process is one that brings together the:



Data (or a representative subset e.g. in a big data situation)



Analytic code and algorithms



Documentation (for process, code and data)



Full computational environment



Packaged in a standard way

So that the work can be reproduced or reused in a similar future project (even years later)

### What are the key points from the case studies?

#### Reproducible workflows improve productivity

• Reproducible workflows are more efficient, freeing up analysts to focus on interpreting results, innovating and solving problems – not wrestling with software or formatting documents.

#### Reproducible workflows help improve the quality of work

• Reproducible work is more easily checked and audited. Having fewer manual processes limits the potential for human error. It allows work to be revisited, updated and improved down the line.

#### Use of open source software

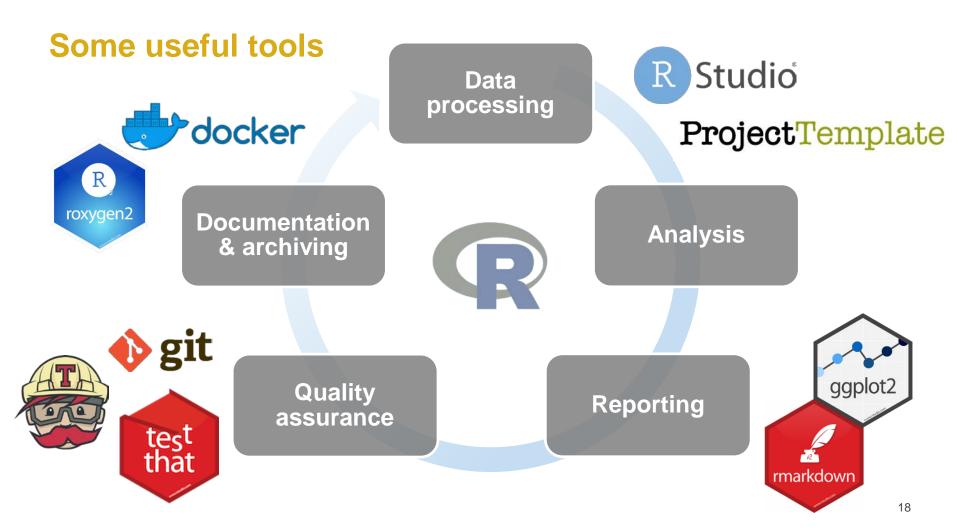
 Organisations are increasingly moving towards the use of open source software – making it possible to introduce reproducible workflows to the workplace.

#### Sharing best practice throughout the organisation

 Government data scientists have shared their findings with over 600 colleagues across government, written a freely available book and a online course. BBC data journalists created a cookbook and a 6 week internal course to upskill colleagues. Their tools are open source and freely available to all (not just the data scientists).

## Why is this important for actuaries?

- Actuaries carry out complex analytical work in data intensive fields. The work will continue to get more complex! Other professions are facing the same challenges – actuaries can learn from work done elsewhere.
- Having the right workflow and tools in place make it easier for actuaries to adopt other techniques from data science and to innovate in future.
- Adopting a reproducible workflow is arguably the first step in automating business as usual work.
- A reproducible workflow provides a solid base from which to demonstrate auditability and TAS compliance.
- Proprietary software remains important but actuaries experimenting with data science are beginning to use open source tools. The 2019 curriculum means the next generation of actuaries will be familiar with R.



## Useful tools for building a reproducible workflow

Studio RStudio is a free and widely used development environment for R that integrates with the tools below.

**ProjectTemplate** ProjectTemplate automates the menial parts of statistical analysis and provides a standard way of working in R.



**y**git

test that

R

roxygen2

<u>R Markdown</u> is a notebook interface that allows code to sit alongside narrative text and can be used for reporting as part of a reproducible framework with <u>ggplot2</u> for creating charts and visualisations.

<u>Git</u> is a version control system for managing code and audit trails – it can be used privately in an organisation or with a web-based service such as <u>GitHub</u>.

testthat is a formal automated testing ("unit testing") package for R.

<u>TravisCl</u> integrates with GitHub to automatically run your tests when code is updated.

roxygen2 automates the production of documentation for your code in R.

Docker packages dependencies inside a container which can run consistently on any infrastructure (see <u>checkpoint/packrat</u> for a lighter touch solution).

### Challenges of building a reproducible workflow

- A reproducible approach isn't the solution to all of the complications of analytical work but it's a good first step.
- Developing a reproducible pipeline can be difficult and time-consuming it's worth exploring whether other solutions are available before developing one from scratch.
- Solutions are generally built around open source software which can be a culture change for some organisations although this is changing.
- Training requirements for new tools and processes.

### Interested in applying these techniques?

#### **Exercises**

We have prepared exercises at <u>https://philipdarke.com/reproducible-actuarial-work/</u> to illustrate how a reproducible workflow can be set up and how some of the tools introduced above are used. The exercises use R to process a data set, carry out some basic analysis and produce a report.

#### Introducing a reproducible process in your organisation

- Take an existing process that is repeated often
- Develop a minimal viable solution using the tools and techniques introduced (see the exercises)
- Pilot it and let others contribute
- Share what you learn within your organisation
- Consider what you can share more widely e.g. with the actuarial community



We are interested in this! Please get in touch with questions, comments and suggestions.

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**Philip Darke** is an actuary with over 10 years' consulting experience at Mercer. Philip believes data science will play a significant role in the future of the actuarial profession, and is currently studying at the EPSRC Centre for Doctoral Training in Cloud Computing for Big Data at Newcastle University. philipdarke.com



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