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B1: Approaches to Building Models in R

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Agenda

- Introduction to R
- Key tools
- Simplified modelling process demonstration
- Benefits of using R for building models
- R limitations
- Resources

Introduction to R

- Open source statistical programming language based upon “S”
- R is one of the most popular data science tools (along with Python)
- The base functionality can be expanded using “packages”
- The usage of R has dramatically increased over recent years:
 - Popular with educational and research communities (e.g. LondonR)
 - Known to be used at many of the leading tech firms (Airbnb, Facebook, Google, Twitter, Uber, etc.)
 - R Consortium support from Google, IBM, Microsoft, Oracle, etc.
 - Microsoft have invested significantly in R after their purchase of Revolutions Analytics (R Open, R Server, SQL Server, AzureML)
 - Insurance applications (e.g. `library(actuar)`, `library(ChainLadder)`, R in Insurance)



Key tools (1/2) – RStudio, R Markdown, R Notebooks

- RStudio is a popular Integrated Development Environment (IDE) for R
- R Markdown documents can be “knit” into HTML, PDF, Word documents or even PowerPoint slides
 - *“R Markdown documents are fully reproducible. Use a productive notebook interface to weave together narrative text and code to produce elegantly formatted output. Use multiple languages including R, Python, and SQL.”*
(<http://rmarkdown.rstudio.com/>)
- R Notebooks are an extension of R Markdown documents that allow outputs to be inline with code (similar concept to Jupyter Notebooks)



Key tools (2/2) – library(checkpoint)

- The successful execution of an R project relies on numerous packages
- However the packages installed by one user may differ from another
- To ensure reproducibility, the checkpoint package downloads the packages as of a particular date for use within an R project

```
library( checkpoint )
```

```
checkpoint( "2017-06-06" )
```



Simplified modelling process



1. Acquire data

- Common file formats are easily read into R
 - `library(data.table)`, `fread(...)` for CSV (as an alternative to `read.csv(...)`)
 - `library(readxl)` for Excel
 - `library(haven)` for SAS datasets
- Access and submit SQL queries using ODBC and `library(dplyr)`
- Other packages can be used to access online APIs or scrape information from the internet
- For testing purposes `library(insuranceData)` may be useful
- Use `readRDS(...)` and `saveRDS(...)` to read and save R objects



2. Process data

- Data is usually stored in a `data.frame` object
 - Consider using `tibble` which makes some improvements in the default `data.frame` behaviour
- Two main packages are used for processing data in R
 - `library(dplyr)` uses action verbs to act upon data frames
 - `library(data.table)` is faster and more powerful however the syntax is more challenging to learn
- `library(feather)` can be used to datasets in a faster binary format
 - `feather` was co-created by the authors of `dplyr` and Python package `pandas`
 - This package was meant to be cross compatible between R and Python



3. Explore and visualise

- `library(ggplot2)` is a very popular graphics package for R
 - “The Grammar of Graphics” is written by Leland Wilkinson and defines a structure to the way data is presented and visualised
 - Graphs are built up by defining `aesthetics` (assign data to `x`, `y`, `fill`, etc.), `geoms` (types of plots), and other components such as labels, axes, titles, themes etc. `facet_grid` can create panels by dimension.
 - Commonly used in professional publications (newspapers, data journalism, etc.)
 - Functionality can also be expanded using extensions
 - `library(ggvis)` is the next iteration of “gg” graphics with interactive components on the `Shiny` platform
- Alternatives include the Base R graphics and `library(lattice)` which focuses on displaying multivariate relationships
- Note: two axes charts are not easy to implement in R



4. Build models

- `glm(...)` is already included within the included `library(stats)` package
 - Use `step(...)` to execute stepwise regression
 - Use `drop1(glmfit, test="Chisq")` to test factor significance
 - `library(broom)` makes it easier to process `glm(...)` output using the verbs `tidy(...)`, `glance(...)`, `augment(...)`
 - Note: `glm` objects should be pared down to save memory (<http://www.win-vector.com/blog/2014/05/trimming-the-fat-from-glm-models-in-r/>)
- A selection of popular machine learning packages is listed below:

Package	Description
<code>earth</code>	Multivariate Adaptive Regression Splines
<code>gam</code>	Generalized additive models with smoothness estimation
<code>gbm</code>	Gradient Boosted Regression models
<code>glmnet</code>	Lasso and Elastic-Net Regularized GLMs
<code>lme4</code>	Linear Mixed-Effects Models using 'Eigen' and S4
<code>xgboost</code>	Extreme Gradient Boosting

Package	Description
<code>caret</code>	(C)lassification And (RE)gression (T)raining
<code>h2o</code>	R scripting functionality for H2O, open source math engine
<code>sparklyr</code>	R Interface to Apache Spark



5. Validate and score (1/2)

- Models can be easily scored on different datasets using the associated `predict(...)` functions
- Deviance can be extracted from `glance(...)`
- `library(Hmisc), rcorr.cens(...)` used to determine the Gini coefficient
- `library(ggplot2)` can be used to create coefficient and standard error graphs using `geom_pointrange(...)`
- Other techniques to consider
 - k-fold cross-validation can be implemented using `cv.glm(...)` although difficult to interpret and perhaps better done using a `library(purrr)` approach
 - `library(broom)` has functionality to bootstrap models



5. Validate and score (2/2)

- `data.frames` are not constrained to holding “data”; they can also hold lists of other objects
- For example, the structure below could be useful in examining different model types

Index	Description	Model_Object	Gini_Train ...	Gini_Val ...	Graphs ...
1	Gradient Boosted Tree	model_gbm	0.342342	0.27373	list...
2	Lasso	model_glmnet	0.340994	0.2699	list...
3	Mixed Model	model_mm	0.330509	0.2238	list...
4	XGBoost	model_xgboost	0.350329	0.28882	list...

- `library(purrr)` facilitates computing metrics and diagnostics across multiple models
- This structure also makes it easy to extract all relevant information for one model iteration

6. Compile results

- Previously discussed R Markdown and R Notebook
- R Markdown also supports Python code chunks
- Government Digital Services (GDS) is using R Markdown for a project
 - Working on a project with the Department for Culture, Media, and Sport (DCMS) on the production of “Economic Estimates for DCMS Sectors Statistical First Release (SFR)”
 - *“At any point in the future we should be able to look back at this work and be able to reproduce everything that we have done today - something that is difficult with manual/semi-manual processes.”* (GDS)
 - <https://gdsdata.blog.gov.uk/2017/03/27/reproducible-analytical-pipeline/>



Benefits of using R for building models

- In “Assessing data analysis and programming” (Hadley Wickham, Garret Grolemund), three properties of good data analysis are noted:
 - [Reproducibility] Rather than using multiple pieces of software which require user intervention, the modelling process can be completed more efficiently from end to end exclusively in R
 - [Automation] With a robust process in R, changes to data and models are easily managed within the processing framework, e.g. updating models with new data and creating comparisons against existing models
 - [Communication] Description of the modelling process is improved with Markdown and Notebook documents
- Packages can be used to enhance and support every phase of the modelling process
- New techniques can be quickly tested within the modelling framework
- Very easy to get started
- Community support and established knowledge base online



R limitations

- All objects are stored in memory
- For larger scale projects, professionally supported versions or cloud-based solutions should be considered
- Python offers functionality that in many cases exceeds R and is perhaps preferred by data scientists and computer programmers <https://www.datacamp.com/community/tutorials/r-or-python-for-data-analysis>
- Newer packages and developments may not be as thoroughly vetted as commercial software options



Resources

- R help function, e.g. `help(glm)`
- Package vignettes which provide examples
- Hadley Wickham is a well known author/contributor of many of the packages discussed today and Chief Scientist at RStudio <http://www.tidyverse.org/>
- RStudio Cheat Sheets (`dplyr`, `ggplot2`, `rmarkdown`, etc.) <https://www.rstudio.com/resources/cheatsheets/>
- Cross Validated (Stack Overflow) <https://stats.stackexchange.com/>
- YouTube



Questions

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