

Comparative Computing



R, Python, Stata, and the shell

Winter Institute in Data Science and Big Data

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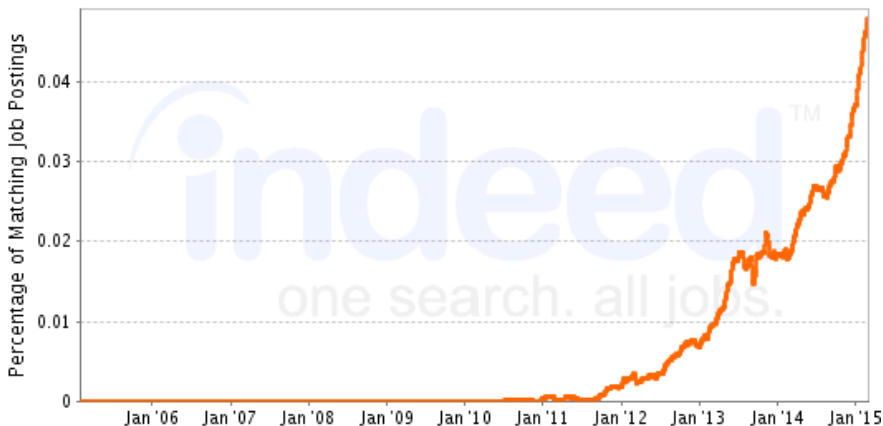


AMERICAN UNIVERSITY
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“Data scientist: The sexiest job of the 21st century”

Job Trends from Indeed.com

— "data science"



Source: <https://hbr.org/2012/10/data-scientist-the-sexiest-job-of-the-21st-century>

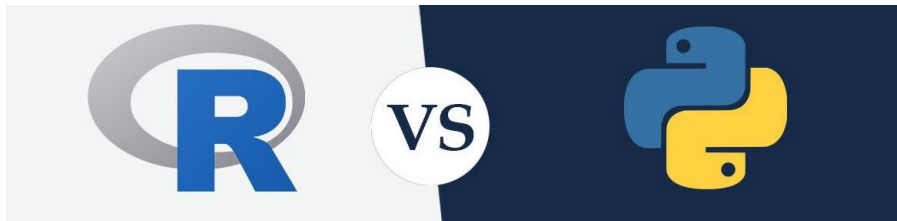
But what programming language to use?

- Widely spread in academia: Stata
- R and Python
- Often neglected add-on: The shell

Stata

- Proprietary (and expensive – \$179/year for single student license)
- No choice of IDE
- Not really used in data science
- Nonetheless: What does it look like?
- Why no love for Stata?
 - ▶ Hard to analyse multiple datasets
 - ▶ Only most expensive versions work with large datasets
 - ▶ Limited resources/functionalities (e.g. predictive modelling, web scraping)
 - ▶ Not consistent with computer science programming (point-and-click)
 - ▶ Not open source
- “Data scientists rely on Stata because of its strong programming capabilities, reproducibility, extensibility, and interoperability”

Which leaves us with ...



R: Lingua franca of statistics



- Open-source
- For statistical analyses
- Academics, researchers, data scientists
- Huge support community
- 1000s of packages (CRAN, GitHub)
- Outstanding visualization
- Advanced reporting (Markdown)



- Can be slow
- Steep learning curve
- Finding right packages can take time
- One IDE: RStudio

Source: <https://www.datacamp.com/community/tutorials/r-or-python-for-data-analysis>

Python: Multi-purpose language

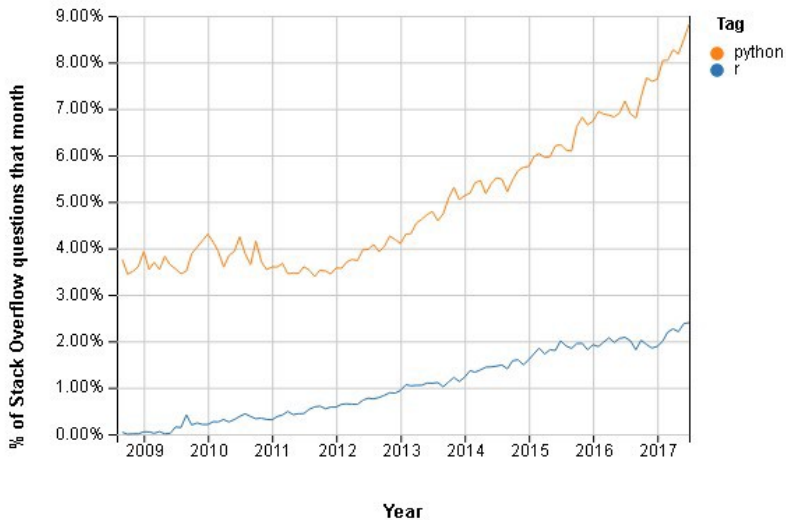


Open-source
For productivity and code readability
Programmers, developers, engineers
Huge support community
100,000s of packages (PyPI)
Moderate learning curve
Fast
Advanced deep/machine learning
Several IDEs: Spyder, Jupyter, Rodeo

Convolutd static visualization
Few(er) data science packages
Finding right packages can take time
Syntax changes between Python 2 and 3

Source: <https://www.datacamp.com/community/tutorials/r-or-python-for-data-analysis>

Popularity



Source: <https://dzone.com/articles/r-or-python-data-scientists-delight>

Jobs



Source: <https://www.guru99.com/r-vs-python.html>

Application

- Fit logistic regression, predict flower species based on measured features

Application

- Bootstrapping: Randomly resample 100,000 times from a population

Application

- File loading (4.8 GB)

```
library(tidyverse)
start_time <- Sys.time()
df <- read_csv("file_loading/library-collection-inventory.csv")
end_time <- Sys.time()
end_time - start_time
# Time difference of 6.814214 mins
```

```
import time
import pandas as pd
start = time.time()
y1 = pd.read_csv('file_loading/library-collection-inventory.csv')
end = time.time()
print("Time difference of " + str(end - start) + " seconds")
# Time difference of 130.32760381698608 seconds
```

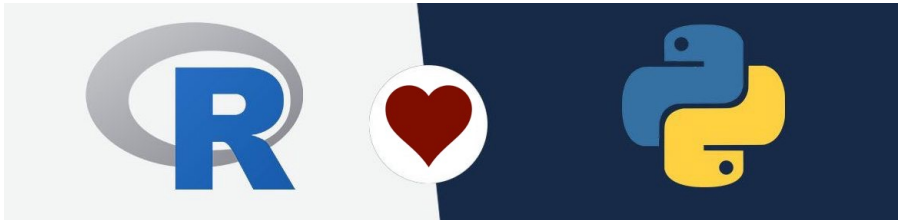
And the winner is . . .



... there isn't one

- One language isn't better than the other
- Both have pros and cons
- It all depends on what you need it for
 - ▶ What problems do you want to solve?
 - ▶ Which language do you have support for?
 - ▶ What are the net costs of learning the language?
- Researcher/Data Scientist? → R
- Developer/Programmer? → Python
- Best solution?

Use both!



Interwoven code: Machine learning

- Application: Build a random forest model that predicts wine quality
 - ▶ R for exploration because of `tidyverse` efficiency
 - ▶ Python for machine learning because of `sklearn` pipeline capability
 - ▶ R for visualization because of `Markdown` and `ggplot2`



Adapted from: <https://www.business-science.io/business/2018/10/08/python-and-r.html>

Add-on: the shell

- Open-source
- Around since the dawn of computers (more or less)
- Hugely beneficial
 - ▶ Remote machines
 - ▶ Cloud computing
 - ▶ Scripts that run for a long time