```
library(dplyr)

rladies_global %>%
  filter(city == 'London')
```



@AnalyticsPanda

# Demystifying Data Science Job Hunting

UCL - Structures Lab
June '17





## Objective



#### **Your Data Science**

#### **Transition**

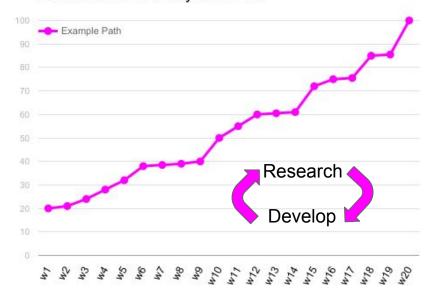


#### Identify Goal/s



#### **Upskilling: Succeed or Learn**

Data Science Journey over time



#### **Achieved!**



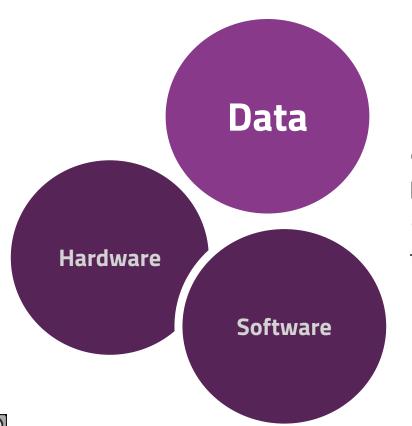


## Base Definitions





#### Tech Industry - not all Apps, Robots & PhDs



P.S you do **not** have to have a PhD in Computer Science to work in Tech!!

"69% of all (coders) tell us they are at least partly self-taught.

13% of respondents across the globe tell us they are *only* self-taught."

#### -Stack Overflow

Developer Survey Results 2016 56,033 coders in 173 countries





#### Maths & Computer **Statistics** Science Knowledge Communication & Empathy Domain **Expertise** (Trusted

Advisor)

#### What is Data Science?

My suggestion:

"Creating Value from Data by applying science/scientific methods"

- General consensus = <u>There is no consensus!</u>
- DJ Patil & Jeff Hammerbacher purport to have first coined the term "Data Scientist" as a job title in 2008
- Hybrid & Interdisciplinary
- P.S Don't assume an Organisation recruiting for a Data Scientist actually knows what a Data Scientist is.....!





#### **Data Science - Misconceptions**

**Data Science != Hadoop** 

**Data Science != Big Data** 

**Data Science != Machine Learning** 

**Data Science != Just Programming** 

**Data Science != Successful Data Science** 





### Realities



#### **Ambiguous Profession - no standardisation**



- Rule of Thumb: Assume every Data Science vacancy is Unique
- Fast Tech Evolution + Differing Maturity of org/industry + Different literacy of the specific people involved in recruitment process + Current Projects + Current Tech Resources
- Rule of Thumb: Prepare to advise the recruiters how to recruit, Don't expect the Hiring Manager has any better idea about the role than you do!
- If you've been actively in job hunting, you'll almost certainly know the current market better than those making the hiring decision!!
- Rule of Thumb: Keep focussing on what YOU can Deliver, and QUANTIFY as much as possible!! | MONEY TALKS
- Obviously try and tailor what you can offer to the specificities of the exact vacancy, but if they
  don't appreciate what you offer, this may well be a reflection of their inabilities, NOT yours!
- A good company will be able to make good judgement
- Be open to feedback being valid, but don't assume it is: gender bias is still rife amongst both technical/untechnical



#### Framing, Framing - Sell Yourself



- Rule of Thumb: Know your proposition (what you offer) like the back of your hand!!
- You should be able to pitch yourself in 15 seconds, or 2 hours | And with confidence
- Rule of Thumb: You don't need to tick all the job spec boxes, You just need to be the best candidate who applies | Success is RELATIVE, not absolute | Based on your competition
- Job specs are wish-lists, NOT a representation of reality
- It is well-known that successful candidates will only fulfill maybe 60% of requirements
- Rule of Thumb: You MUST portray yourself as best as possible | IMPRESS THEM!
- Your competition are likely already over-inflating their CVs, so if you downplay yourself you've created an even bigger discrepancy
- Rule of Thumb: Recruiters are time-pressured and making Snap Decisions, so make it easy for them to review you | IMPRESS THEM FAST!
- The hiring manager may have 30secs to click Yes or No on your application to invite you to interview: you CANNOT afford to be humble, subtle and indirect!



### Salary Negotiations - START AS HIGH AS POSSIBLE



- Rule of Thumb: START HIGH
- Employers will offer the absolute minimum they can get away with
- You will end up with a better result if you go high and get negotiated down vs go low initially and then try to suggest you meant you wanted more (also shows lack of judgement)
- After all, if you undervalue yourself, why shouldn't they?



#### Please takeaway these points!



### TRYING TO BE LIKED & TRYING TO CONFORM WILL GET YOU ABSOLUTELY NOWHERE

For you to WIN in Tech, you HAVE to play the game - it's your choice to play the game or not, but your likelihood of success will be minimal if you don't

We are not in a position where we can afford to be SELF-DEPRECATING | This is a <u>COMPETITION</u>, and as we all know, our minority identities already put us at a disadvantage - we must be prepared for a challenge!





## Maturity





#### **Maturity Model**



Level	Structured Processes	Discoverability & Compounding	Analytical Speed & Agility	Breadth & Depth of Impact	Organizational Cohesion
1 Ad Hoc Exploration	Practitioners operate autonomously in a black box	Assets stored locally, emailed around	Limited talent and tools	lvory tower, no tangible value	Analytics island, purely transactional
2 Repeating, but Limited	Recurring workflows discussed, no enforcement	Assets stored centrally, but lack metadata / permissions	Some tools and talent investment	Static reports in a few business areas	Some collaboration with line managers
3 Defined and Controlled	Formalized process, manually enforced	Assets stored and tagged centrally with metadata and permissions	Rapidly test ideas with novel methods / tools	Results translated into multiple operational workflows	Analytics are key stakeholders in strategic decisions
4 Optimized and Automated	Best practices codified into infrastructure, transparency for all	All asset versions stored/tagged, searchable, reproducible	Cutting-edge tools, comfortable at the analytical frontier	Data products drive org with robust safeguards	Analytics enmeshed in business and proactively anticipates needs



Source: https://blog.dominodatalab.com/introducing-the-data-science-maturity-model/

Source: http://www.prweb.com/releases/legal-department/maturity-model/prweb12575388.htm



## Targeting





### Industries/Sectors In order of DS sophistication

Legacy/Slow

Public Sector

Retail

Media

Marketing

Middling

Consultancies

Marketing

E-commerce

Finance

**Cutting/Bleeding Edge** 

Academia

Major tech players

Tech start-ups



### Latest self-reported Salaries for Data Scientists (London, Glassdoor)

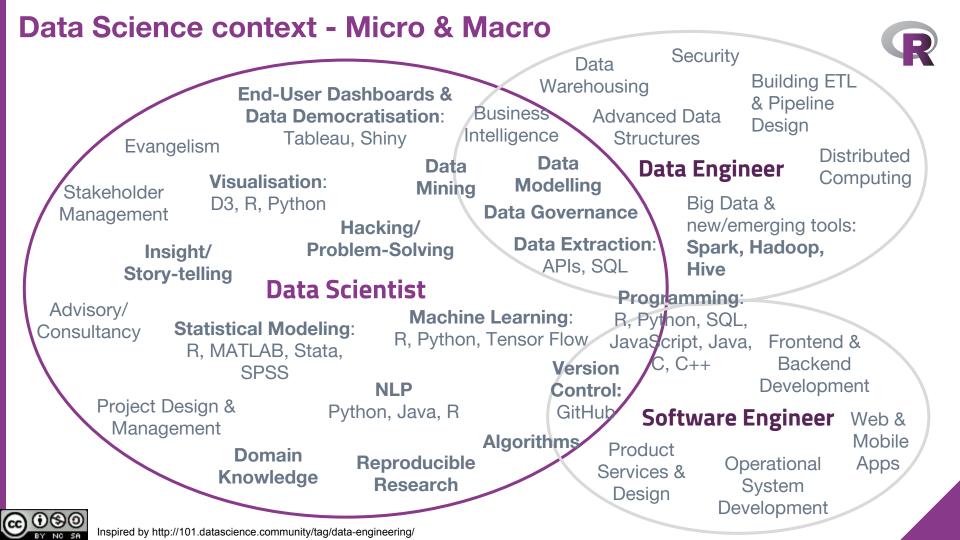
#### Data Scientist in London Salaries

Job Title	Location	Salary
Ocado Data Scientist	London, England (UK)	£38,172
Tesco Data Scientist	London, England (UK)	£45,296
Unilever Data Scientist	London, England (UK)	£36,878
Dunnhumby Data Scientist	London, England (UK)	£70,432
Imagini Holdings Data Scientist	London, England (UK)	£52,744
Virgin Media Data Scientist	London, England (UK)	£47,424
First Derivatives Data Scientist	London, England (UK)	£20,104
King Senior Data Scientist	London, England (UK)	£63,333



### Skills







### Workflow





#### **Data Science - Workflow**

Data

Visualise

#### Receive or Identify: -business problem -research question -hypothesis -project brief Design Work Collect

Munge

& Wrangle

#### Create:

Analyse

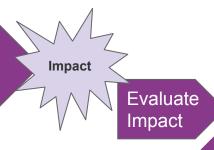
& Model

- -output
- -insight/finding
- -product
- -deliverable

Effectively: -inform/affect -communicate

-implement

-get adoption







#### Receive or Identify:

- -business problem
- -research question
- -hypothesis
- -project brief

Design Work

Collect

Munge
& Wrangle

Data

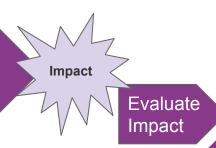
Analyse
& Model

Visualise Computational data skills are Necessary but **not** Sufficient!!

#### Create:

- -output
- -insight/finding
- -product
- -deliverable

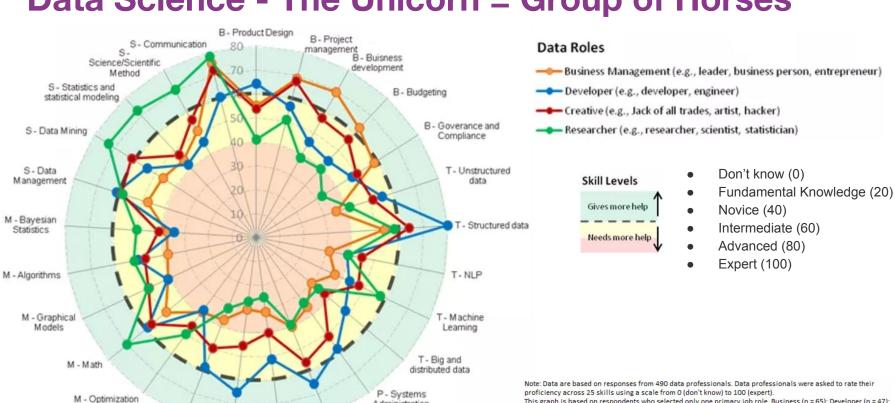
Effectively:
-inform/affect
-communicate
-implement
-get adoption







#### **Data Science - The Unicorn = Group of Horses**



Administration

P - Database

P - Cloud Administration

Management

This graph is based on respondents who selected only one primary job role. Business (n = 65); Developer (n = 47); Creative (n = 25); Researcher (n = 101)



P - Front-end

Programming - Back-end

Programming



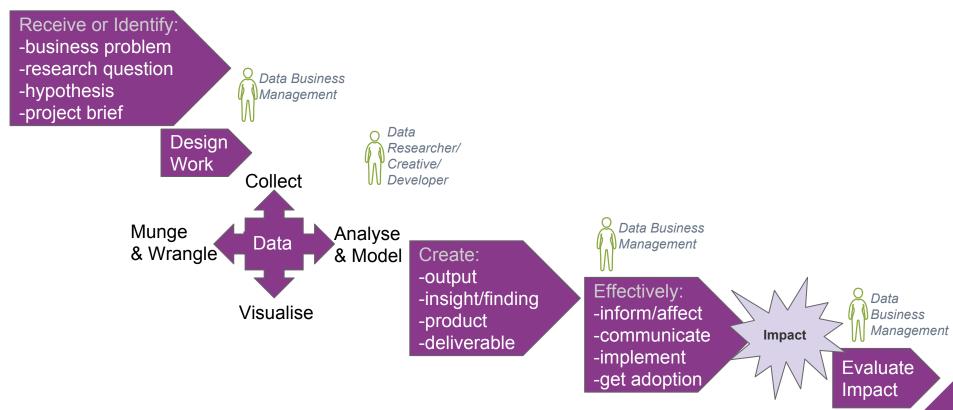
#### **Data Science - The Unicorn = Group of Horses**

- Leader, Business person, Entrepreneur | Data Business Management are the
  product and profit-focused data scientists. They're leaders, managers, and
  entrepreneurs, but with a technical bent. A common educational path is an engineering
  degree paired with an MBA.
- Developer, Engineer | Data Developers are focused on writing software to do analytic, statistical, and machine learning tasks, often in production environments. They often have computer science degrees, and often work with so-called "big data".
- Generalist, Hacker | Data Creatives are eclectic jacks-of-all-trades, able to work with a broad range of data and tools. They may think of themselves as artists or hackers, and excel at visualization and open source technologies.
- Researcher, Scientist, Statistician | Data Researchers apply their scientific training, and the tools and techniques they learned in academia, to organizational data. They may have PhDs, and their creative applications of mathematical tools yields valuable insights and products.





#### **Data Science Workflow**





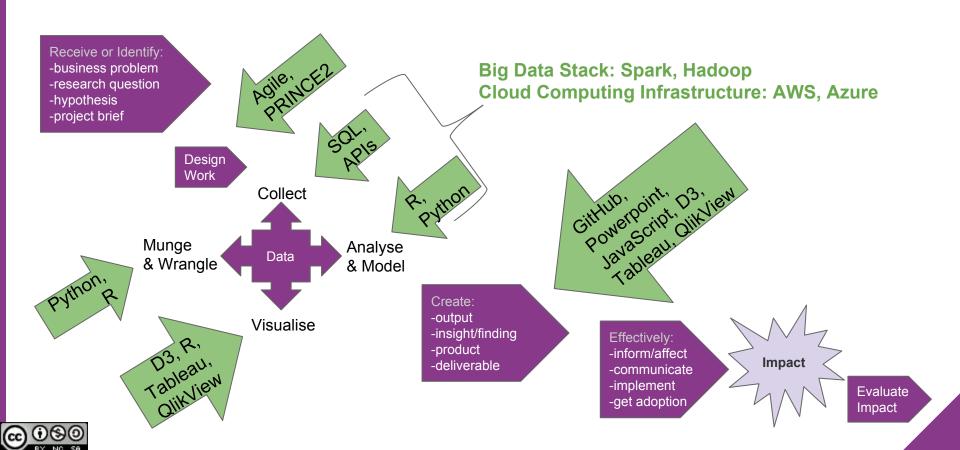


## Prepare





#### **Data Science Workflow - where tools fit**





#### **To Do - Interview Questions**

#### Core

- Walk me through your CV
- Why do you want this job
- Why are you suitable for this job
- Why are you looking to move from your current role/situation
- What are your salary expectations

#### **Possible**

- Greatest (data science-related) achievement
  - Explain any gaps in your CV
  - What are your weaknesses

#### **Vacancy-Specific**

Are you aware of the domain-specific challenges relating to <eCommerce/Finance/Gov/Health etc>?

#### Things to Ask

- What are the next steps of the process?
- What is the provenance/background to this vacancy?
- Where does this role sit in the organisation structure, e.g. under Chief Tech Officer, Chief Marketing Officer?
- Why did you (to interviewer) take this job/What have you enjoyed about this job?
- What is the organisation's vision/strategy/position w.r.t Data Science/Tech, e.g. investing, exploring?



#### To Do

**Fundamentals** 

Useful

If Possible













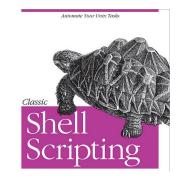














#### **Example hand-picked live Data Science roles**

(at time of writing)



Data Scientist Entry-Level https://www.linkedin.

https://www.linkedin. com/jobs/view/36166 5375/



Data Analyst Associate (Junior) | Excel & SQL

https://www.linkedin.com/jobs/view/364310099/



Junior Data Scientist Entry-Level SQL & R Reasonable

Requirements

https://www.linkedin.com/jobs/view/340146959/



British RedCross Data Scientist Mid-Senior

Reasonable Requirements

£34,136 - £47,450 + London weighting https://www.linkedin.com/jobs/view/322581997/



Data Scientist Mid-Senior

Reasonable Requirements

https://www.linkedin.com/jobs/view/297108433/



Junior/Graduate Data Scientist Excel & SQL (Access)

Easy Requirements

https://www.linkedin.com/jobs/view/357961672/



Insight Analyst - Data Science Entry-Level

Reasonable Requirements

https://www.linkedin.com/jobs/view/304630939/

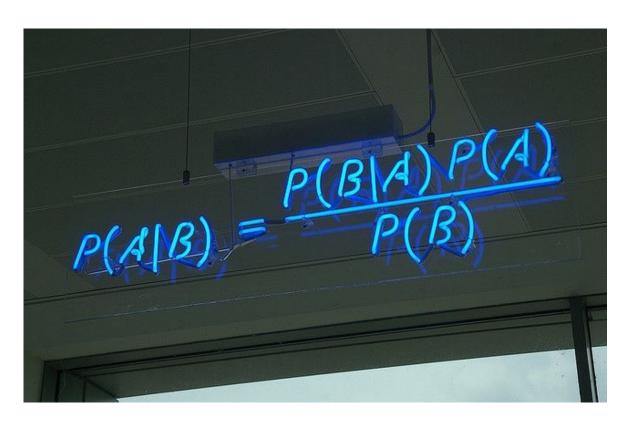


### Let's Go!





#### Data Science is awesome!!!





### Appendix



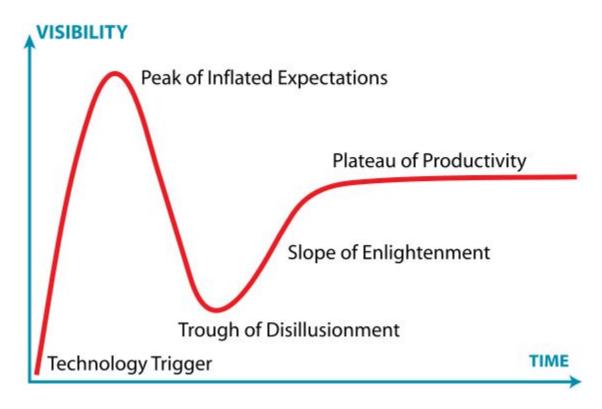


## Hype





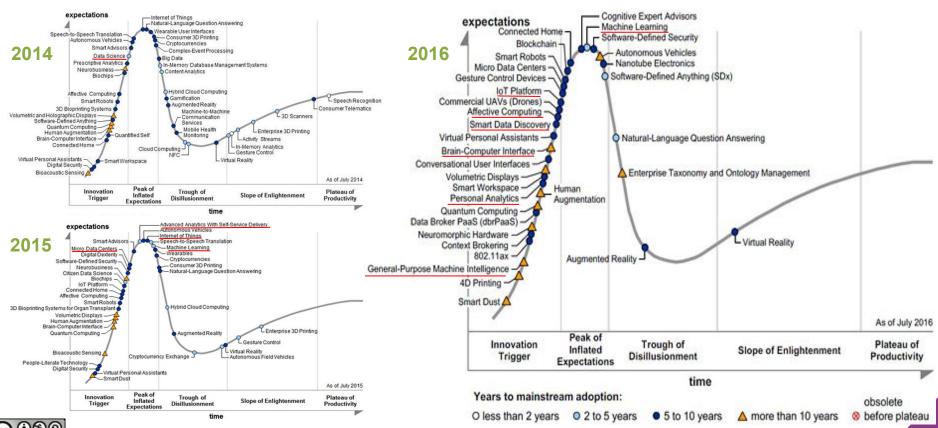
#### **Hype Cycle**







#### **Hype Cycle - Silver Bullets?**







### Pitfalls



#### **Data Science Pitfalls - Noise**



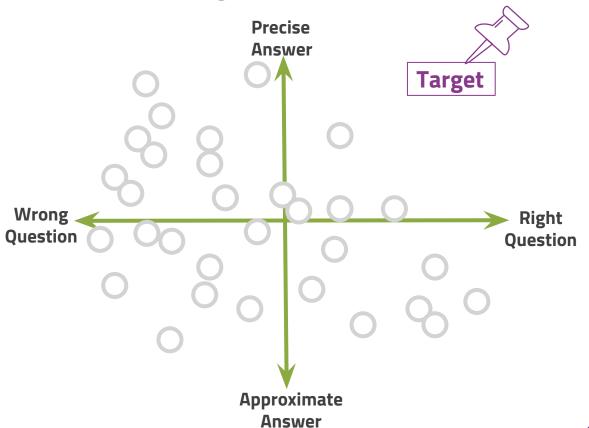


#### **Data Science Pitfalls - Wrong Question**

"Far better an **approximate answer to the right question**, which is often vague, than an exact answer to the wrong question, which can always be made precise."

—John W. Tukey

The Future of Data Analysis (July 1961)





#### Data Science Pitfalls - Blind Leading the Blind?



#### **Data Scientist**

The Co-op

Manchester, England £60,000 to £80,000 p.a

We're looking to recruit an experienced Data Scientist to join our team, and unearth information from our data assets that'll help Co-op make smarter decisions for our members. You'll work as an internal data consultant with people from right across the Co-op family to find simple solutions to complex problems. And if you can bring us the combination of data expertise & great people Skills we need, we can offer you a unique opportunity to help build our data capabilities from the ground up. To be successful as a Data Scientist at Co-op, you'll need to be an established data expert. Statistical models will be your speciality, but in addition to tried and tested technical ability, it's important you've got the skills to put these techniques to good use in our business.

#### What you'll do:

- Understand Co-op objectives, and Create algorithms that deliver positive impacts for members
- Be capable and confident in tackling the most complex data-driven and analytical problems
- Work across several simultaneous projects with partners from various Co-op

  businesses
- Take on different roles on different projects (expert adviser/statistical guru/lead developer)
- Write code when you need to in order to tackle and troubleshoot the most complex tasks
- Make sure new problem solving approaches comply with existing Co-op rules and values
- Plan projects over a 12 month horizon, and manage budgets using Agile principles
- Contribute to the Coaching & mentoring of a pool of less experienced colleagues
- Design solutions that create outputs which are accessible to non-technical audiences
- Be a great advocate for Data Science, promoting our services to Co-op colleagues



#### Data Science Pitfalls - Blind Leading the Blind?



### Data Scientist Telegraph Media Group London, GB

#### Requirements:

- Robust experience in SQL (Hive QL or any other flavour) with databases spanning tens of millions of records
- Solid fundamental understanding of statistics, and experience with statistical analysis using R or Python from university degree or previous role
- Be knowledgeable about the digital ecosystem
- You are a Communicative person that values building strong relationships with colleagues and stakeholders and have the ability to explain complex topics in simple terms
- A Self-starter who's comfortable working autonomously
- Solid commercial skills and business awareness

#### To be successful in this role, you will need:

- Outstanding logical thinking, a very strong background in a quantitative discipline such
  as mathematics, physics, engineering, operations research, data science or similar, and
  a scientific approach to problem solving
- Demonstrate ability to manage stakeholder, show diplomacy and be an effective influencer
- Good technical skills (at least some practical experience with statistical/pattern recognition/machine learning methods)
- A sharp focus on getting results and insights using the most practical and fastest approach (even if it's not the most challenging or interesting for you)
- Ability to perform in a high pressure, fast paced environment and be comfortable with a high level of ambiguity
- Ability to Communicate complex concepts concisely and clearly, including to senior level executives
- High energy and a can do, no-excuses attitude
- curiosity and passion for making an impact
- Good interpersonal skills

#### **Behavioural Competencies:**

Bright, driven, love challenges and want to change the editorial world