

Data Science Workshop

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RSS Data Science Section EFSPI Statistical Leaders' Meeting 4th July 2017

Introduction – 15 mins

LUNCH

- Survey analysis 15 mins
- Case studies 10 mins
- Group work: four themes 30 mins
 - •The Internet of Things
 - •Big Data: EHRs
 - Decision science
 - •Automation and artificial intelligence
- Report back 25 mins
- Discussion 25 mins







Data Science Section Remit

To be a professional body that represents data scientists in the UK. The section will organise meetings for a broad range of attendees and generate outputs that are aimed at:

- Promoting good practice by addressing what good Data Science looks like (with exemplars) and what it does not look like.
- Promoting the statistical aspects of Data Science / re-enforcing the statistical framework
- Being a trusted voice on Data Science for employers, including inputting to consultation exercises
- Supporting the Data Science community throughout the UK
- Supporting the pipeline and career development of data scientists and statisticians by elevating skill sets to work in the modern world
- Supporting important emerging topics such as ethics, privacy, algorithmic responsibility and personalization lifting the quality of the conversation
- Fostering multi-disciplinary connections and the exchanging of ideas



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DSS Committee Members

Fran Bennett – Mastodon C Simon Briscoe (Council representative) David van Dyk – Imperial / ASA DS Chapter Andrew Garrett (Chair) - ICON Martin Goodson – Evolution AI Mark Girolami – Turing Institute / Imperial Ioanna Manolopoulou - UCL Giles Pavey – ex Dunnhumby/Tesco Harry Powell – Barclays Richard Pugh (Meetings Secretary) – Mango Solutions Matthew Upson (Secretary) – Cabinet Office Leone Wardman - ONS James Weatherall (Vice Chair) - AZ





DSS Launch event

The Industrialisation and Professionalisation of DS (19th June)

- 12 Questions presented, with three formal responses
- An example topic
- President's response
- Q&A

YouTube: https://m.youtube.com/watch?v=5aH3vVvtOfc

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DSS Social Media

- RSS website: landing page
- Twitter: @RSS_DSS
- GitHub: <u>https://github.com/rssdatascience</u> LinkedIn:
- https://www.linkedin.com/company-beta/111500048/
- Slack: https://rssdatascience.slack.com



ROYAL

STATISTICAL

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Please sign up!

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Personal definitions of data science



There are a wide range of perspectives

Gaining Knowledge and Insights from Data

Data Science is an interdisciplinary field of expertise about processes and systems to extract knowledge or insights from data in various forms, either structured or unstructured, in order to address various kinds of technical, scientific and business needs

Data-driven science based on maths, computer science and domain knowledge

Combination of computational and statistical expertise to access and analyse data

Data visualisation, modelling, simulation and AI technologies are applied in Data science

A multidisciplinary field, merging math/stat skills with computer science and

Evidence that we have failed as a statistical discipline

Database setup/programming, CRF design, data management

A blend of statistics, IT and mathematics for big data



Visualisation skills - usually focussed on a specific domain

Data science is recognised in most organisations



Broad range of contributions from data scientists



Most believe a more mature data science capability is needed



Future look: Insights, IoT, visualisation & decision science



Opportunities for data science throughout development



STATISTICAL SOCIETY DATA EVIDENCE DECISIONS

ROYAL

Where is the gap?



There are a wide range of perspectives

we definitely lack people able to assemble or transform the diverse datasets; we also need more associates knowledgeable or experts in Machine learning type of methods

Complexity of the big data topic and variety of potential applications makes it challenging to focus and join forces between computationally oriented and statistically oriented staff

Statisticians with an interest in non-traditional data sources people with an interest in non-traditional data sources who understand anything about statistics, uncertainty, randomness

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STATISTICAL

DATA EVIDENCE DECISIONS



Develop experienced DS teams gathering expertise in technology/mathematics/computer sciences while being open minded and being able to embark and lead DS projects with other scientists (biologists --> clinicians) or internal partners

Organisational boundaries

This is a multidimensional activity needing staff with different skills. Challenge is to have the right balance in the team

Limited resources/competencies in the critical areas like AI, wearable/sensor technologies



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Patient Flows in EHR data OncologyFlo



ROYAL STATISTICAL SOCIETY DATA | EVIDENCE | DECISIONS

Unsupervised machine learning – Insights into healthcare



"Seven Ages of Man" healthcare clustering



What is robotic process automation (RPA)

 Software that automates repetitive, rules-based tasks to free up your best people to be your best people

| Robotics (RPA) | Cognitive Automation | | Artificial Intelligence |
|--|---|--|--|
| TY. | ~\$C\$ | | £⊕ |
| "Mimics Human Actions" Used for rules based processes Enables: Faster processing time Higher volumes Reduced errors | "Mimics/Augments Quantitative Human Judgment" Used for judgement based processes Machine learning capability Interprets human behavior | "Augments Human Intelligence" Used for predictive decisioning Dynamically self- adaptable and managing | "Mimics Human Intelligence" Turing Test Definition – "A test for intelligence in a computer, requiring that a human being should be unable to distinguish the machine from another human being by using the replies to questions put to |

Safety data collection via Robotic Process Automation



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Group work – three key questions

- 1. Brainstorm: what are the main opportunities and challenges
- 2. What are the top 3 areas we should address as statistical leaders
- 3. What immediate action should we take next?





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The End

Thank you!



