Master in Data Science - Introduction

Israel Herraiz

March 16, 2016

Contents

- 1 The Data Scientists
- 2 The Data Science Process
- 3 Program of the Master in Data Science
- 4 Recommendations to follow the master
- 5 This week's session

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Why data science?

The intelligente use of data:

- has become a source of competitive advantage
 - Better knowledge about the market
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The goal

Data-Driven Decision Making

What is a data scientist?

Key skills

- Fitting in an organization, leading projects in a heterogeneous environment, aligning with strategy
- Data-Analytic Thinking
- How to extract knowledge from data

The three facets of a data scientist

- Functional (domain knowledge)
- Analytical (how to extract knowledge from data)
- Technical (how to implement the data science process)

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The three facets of a data scientist

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Question

- Are the three facets equally important?
- Which facet is the most important?

Skills demanded in job postings

Analytical skills

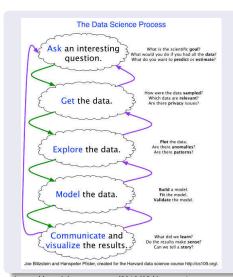
- Machine Learning
- Statistics
- Bias towards engineering and scientific backgrounds

Technical skills

- R or SAS for statistical programming
- Python
- SQL
- Hadoop and Big Data (covered only partially)
- Visualization technologies (Tableau, Qlik, D3.js)
- Excel

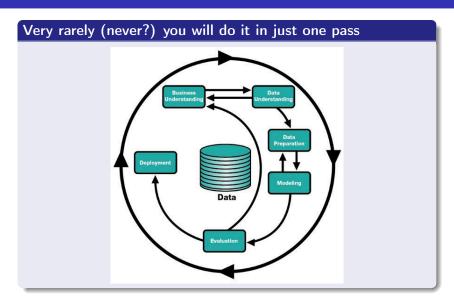
- 1 The Data Scientists
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The Data Science process



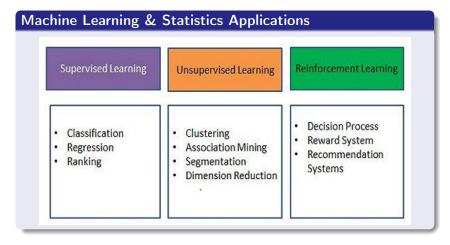
- 1 Ask a question
 - Domain expertise
- 2 Get and prepare the data
 - Python, Pandas
- 3 Explore the data
 - Pandas, Matplotlib,R, Spark, Tableau
- 4 Model the data
 - R, Spark
- 5 Communicate the results
 - D3.js, Tableau

But it is always iterative...



https://en.wikipedia.org/wiki/Cross_Industry_Standard_Process_for_Data_Mining

What kind of questions can we answer to?



http://www.kdnuggets.com/2015/09/questions-data-science-can-answer.html

What kind of questions can we answer to?

Classifications of questions

- Is this A or B?
 - Binary classification
- Is this *A*, *B*, *C* or *D*?
 - Multi-class classification
- Is this normal or weird?
 - Anomalies detection
- How much or how many?
 - Regression
- How is this data organized?
 - Unsupervised learning
 - Dimensionality reduction
- What should I do now?
 - Recommendation systems

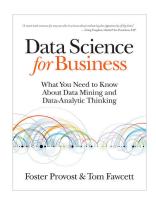
- 11 The Data Scientists
- 2 The Data Science Process
- 3 Program of the Master in Data Science
- 4 Recommendations to follow the master
- 5 This week's session

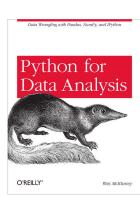
The program

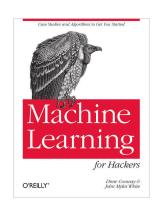
- Intro to data science
 - Setup environment, working with the command line, intro to Git
 - Intro to SQL
- Data analysis
 - Python for Data Analysis
 - Intro to Statistical Programming (R)
- Machine Learning and Statistics (R)
- Big Data
 - Very brief intro to Hadoop and Spark, using Python
- Visualization and Business Intelligence
 - Dashboards development with Tableau
 - Brief intro to D3.js

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Recommended books







Share promiscuously

You need visibility

- Use the social networks to position yourself as a data scientist
 - Crucial: Profile in LinkedIn and share, share, share!
 - Share frequently about your progress in the master
- Share your code in Github or Bitbucket
 - Try to use it frequently, so it always shows recent activity
 - Don't worry if you don't know Git, we will see in the first session
- Don't mix (too much) your personal postings with your data scientist postings
 - For instance, use Facebook for your personal network, and Twitter for your professional postings

Recommended readings

Amadeus Data Scientists series

http://www.amadeus.com/blog/tag/data-scientist/

Guía de las profesiones de Internet

http://www.avanzaentucarrera.com/llegaraser/profesiones-y-p

Key ideas to remember during the master

■ Extracting useful knowledge from data to solve business problems can be treated systematically by following a process with reasonably well-defined stages.

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- If you look too hard at a set of data, you will find something but it might not generalize beyond the data you are looking at.
- Formulating data science solutions and evaluating the results involve thinking carefully about the context in which they will be used.
- It's not about the technologies. Technology will always change very fast. Learn the concepts, apply them with technology. Be open to learning new technologies (and sometimes it will also imply learning new concepts). The only constant is change.

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Intro to data science

We have just completed this part.

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The environment

Setting up the environment with VirtualBox and Fedora 23

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The environment

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Starting with Git

Intro to Git, and the importance of sharing our source code

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Getting familiar with the command line

Many times the shell command line is enough to answer to a lot of questions