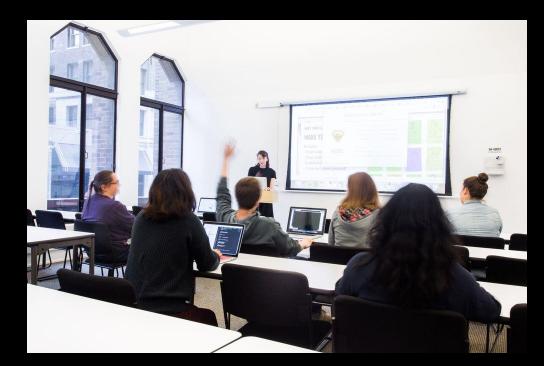


UNIT 1 ORIENTATION, DEV ENVIRONMENT, PYTHON REVIEW, DATA SOURCE PRESENTATION



Orientation

Introduction
Course Overview
Introduction to Data Science
Types of Data



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Introduction

Name

Background in Data or reason for taking class

Something about you ie Hobby, fun fact, current projects, Favorite movies or books



Course Overview



20 session grouped by 4 units, 3-unit demonstration projects, 1 final project.

UNIT 1 PYTHON FOUNDATIONS

- Unit presentation: Python Technical Code Challenges

UNIT 2 EXPLORATORY DATA ANALYSIS AND HYPOTHESIS

- EDA + Chipotle data
- Final presentation: Proposal + Dataset

UNIT 3 MODELS(Linear regression, KNN/ Classification, Logistic regression)

- Final presentation: Initial EDA Brief
- Linear Regression and KNN Practice(Optional)

UNIT 4 TIME SERIES, NLP, MODEL EVALUATIONS

- Final presentation Technical Report
- Final presentation Executive Report

Course Outline

UNIT 1 PYTHON FOUNDATIONS

- Welcome to Data Science
- Config Dev Environment
- Python Foundations
- Lab and Presentation

UNIT 2 EDA AND HYPOTHESIS

- EDA in Pandas
- Working with data: APIs
- Data Visualization
- Statistics in Python
- Experiments and Hypothesis
- Lab and Presentation



UNIT 3 MODELS(Linear Regression, KNN/Classification, Logistic regression)

- Linear Regression
- Train-Test Split & Bias Variance
- KNN / Classification
- Logistic Regression
- Lab and Workshop

UNIT 4 WRANGLING, TIME SERIES, & NLP

- Intro to Time Series
- Intro to Natural Language Processing
- Flex more models, SQL Data management, Data cleaning, more evals
- Review + Lab
- Presentations

Course Outline

All pre-built exercises drills or homework

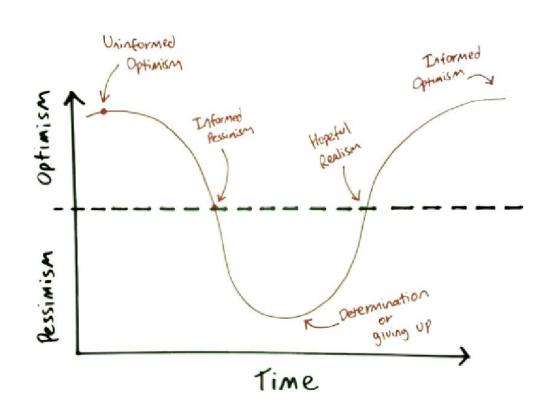
Sublime	
Python	
Chipolte	

Learning

10% skill90% Never give up, Never surrender

Strategies for Class:

- Progress rating
- Road Map/Cheat sheets
- Mentor and Present



Study partners, study groups increase retention, accountability, support, learning perspectives, networking, and can lead to friendships or fun.

Practice skills on easy data sets: Start small, find a data set that is easy and/or something that you like or fascinates you. Get the skills down and focus on harder data problems later.

Feach others a topic that is hard for you, a classmate, a partner, or a friend.

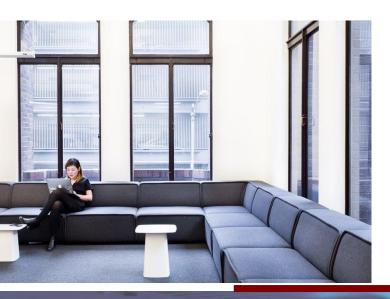
Free time Fake out. When you don't feel like doing a task, put a show on that you have seen 100 times, open your laptop and just start on the task, you will find it can turn into a few hours because well you have already seen that episode and you are not really working because you are watching TV.

Set specific small goals with quick deadlines. It is easier to complete several 1hr tasks then one 8-hour task.

Procrastinate with 2 or 3 other tasks. Rotate when you procrastinate.

Be a learn it all instead of a know it all. There is always something new to learn even with something you feel you have mastered.

"I dedicate 1 to 3 hours a day, determined to master the material."





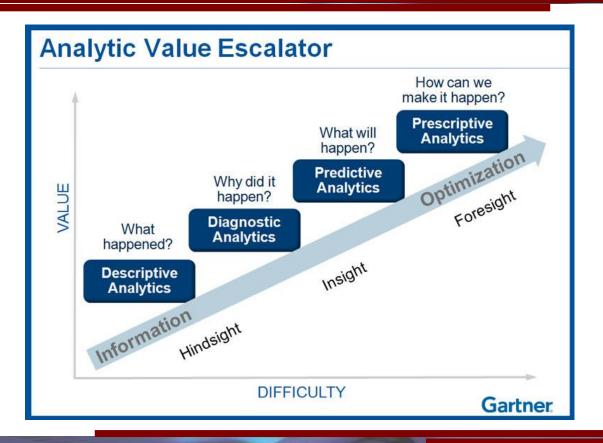
1) Out of the units what is most interesting to you and what seems most challenging?		
2) Which Study tips seem realistic to apply? Do you have additional study tips?		

Introduction to Data Science

What is Data Science and Why do we care? What are the tools of a Data Scientist? What is a Data Scientist workflow look like?

BI or Business Intelligence Business Analyst Data Analyst Data Scientist Data Engineer

A good article on the difference between BI and Data Science https://www.itproportal.com/2016/08/18/10-differences-between-data-science-and-business-intelligence/



"Data Scientists are people with some mix of coding and statistical skills who work on making data useful in various ways."

Data Scientist Type B (for Building):

- Some statistical background, but strong coder or software engineer.
- Primarily concerned with using data "in production": building models which interact with users (by giving recommendations, for example).

Our course is focused primarily on **Type A**.



Data Scientist (n.): Person who is better at statistics than any software engineer and better at software engineering than any statistician.



12:55 PM - 3 May 2012



Data Scientist (2/2): person who is worse at statistics than any statistician and worse at software engineering than any software engineer





Data Scientist (n.): Person who is better at statistics than any software engineer and better at software engineering than any statistician.

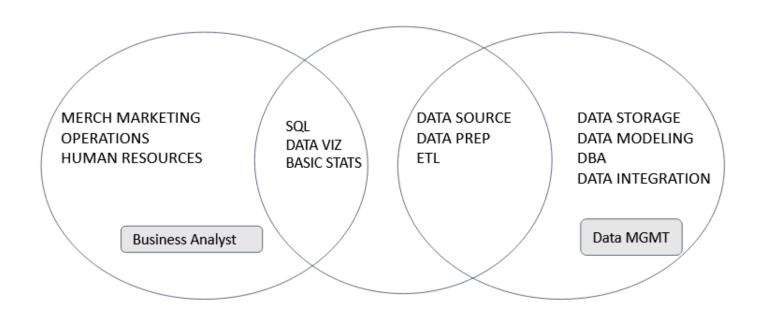


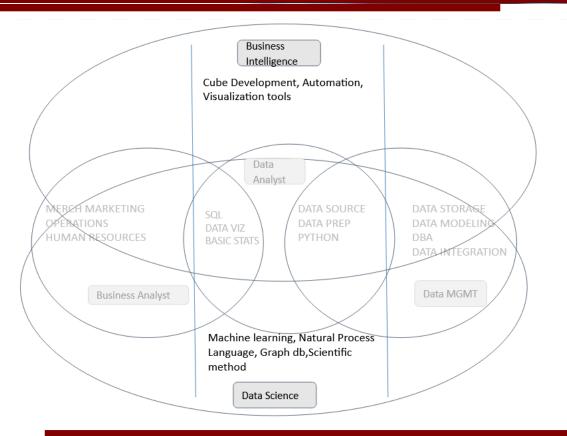
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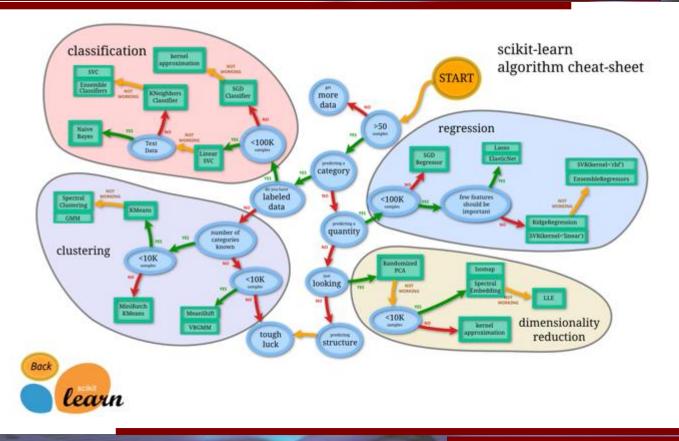


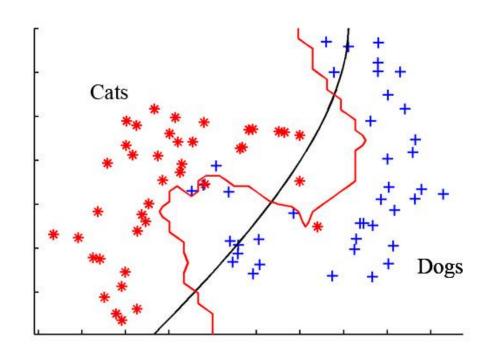
Data Scientist (2/2): person who is worse at statistics than any statistician and worse at software engineering than any software engineer

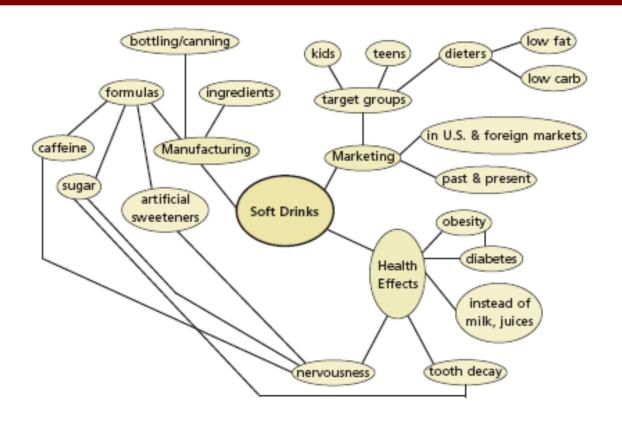


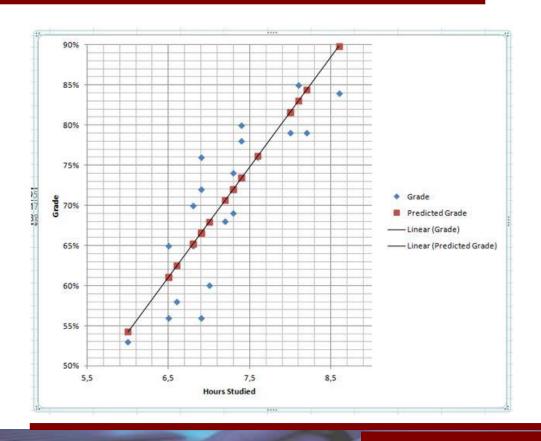














EXAMPLES OF DATA SCIENCE IN ACTION

- Facebook facial recognition in photos
- Netflix/Amazon/Spotify recommendations
- Siri/Echo/Cortana voice recognition assistants
- Building art with Neural Networks https://github.com/jcjohnson/neural-style
- Faceswap https://www.youtube.com/watch?v=UngUWA43q5o
- Stock Market https://www.quantopian.com/ building crowd source hedge funds
- Helping people
 https://www.drivendata.org/ who is a good bet to give money to http://www.datakind.org/projects
- Help find missing children http://www.datakind.org/projects/finding-30000-missing-children
- Find correlations from sickness, grades, and attendance and try to find ways to improve them http://coolculture.org/webfm_send/62
- Additional examples https://www.kaggle.com/wiki/DataScienceUseCases





ORACLE

Data storage, Understanding schemas, tables, fields, relational and non relational databases is a foundation of data analytics











Business knowledge can include understanding: Knowing KPI's, Gather requirements, MetaData, Operational reports, Business acumen, communication and navigating politics and personalities of your business culture



Having a strong understanding of Lookup functions, string and numeric functions is necessary to understand the business and how the currently tackle problems.



Basic Statistics(Central Tendency) Understanding concepts is fine. Understanding long hand even better.



This is a must to understand the basic charts and graphs and be able to tell a story with them.



Structured Query Language: Unless someone is getting all of your data for you and cleaning it all for you, you will want to be proficient in SQL up to Advanced levels.



Python is a general purpose programing language. Allows you to give directions to a computer to tell it what to do.



R is a system for statistical computation and graphics.



SAS **SAS** (Statistical Analysis System) is a software suite developed by **SAS** Institute for advanced analytics, multivariate analyses, business intelligence, data management, and predictive analytics.



SPSS Modeler IBM **SPSS Modeler** is a data mining and text analytics software application from IBM. It is used to build predictive models and conduct other analytic tasks



Jupyter Notebooks allows you to create and share documents that contain live code, equations, visualizations and explanatory text.



Plain text formatter that converts for use in html, used to create documentation within Jupyter



Purpose of git is to manage a project, or a set of files as they change over time. It allows for version control and collaboration.



Command line is a user interface to a computers operating system. It allows you to navigate, manipulate and analyze files, data and more.

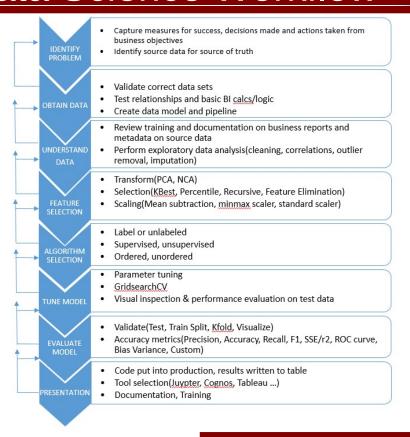


Data Science Workflow

ANALYTICS WORKFLOW



Data Science Workflow



1) What is the difference between a Data Analyst, Data Engineer, and a Data Scientist?
2) In your own words how would you describe the 4 primary algorithm groups in scikit learn?
3) What is the difference between R and Python? Which should you learn?
4) What is the difference between an Analytics workflow and a Data Science workflow?
5) When would you use Jupyter?

Quantitative

Discrete and Continuous

Qualitative

Nominal and Ordinal

----Types of Data Formats----

Flat File

Text Documents

Time Series

Transactional Data

Relational Data

Spatio-Temporal Data

Image Data

Network Data

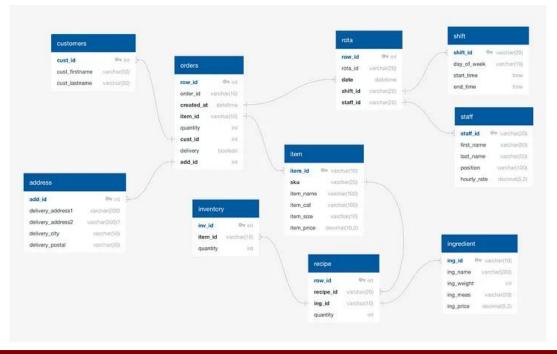
Text Files – data conveyed in written or printed form. Surveys, Social Media Posts, Books, messages in chat. Can be used for word clouds sentiment analysis, give insights, however, may contain errors, bias

Flat File – rows and columns of data typically seen in .csv files. Single files that are not modeled or related.

Transactional Data – Date stamped events (logs): can be represented as time series. Customer bought a coffee at 12PM

Time Series – Similar to Transactional data, however, can be used to as successive measurements over time rather than over transaction. Number of coffees sold between 12 and 1pm or number of customers served between 12 and 1PM.

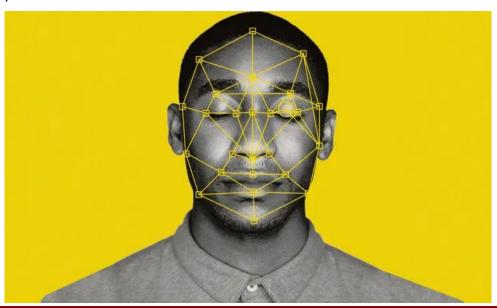
Relational Data- stores tables that relate to on another example would be a customer table that relates to an item table that relates to a transaction table.



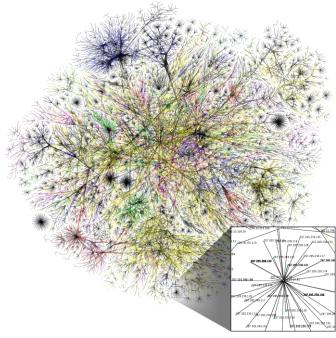
Spatio-Temporal Data – information about space and time. Examples include Weather patterns, predicting earthquakes, determining global warming, evaluation of traffic for city planning and more.



Image Data: photographic or trace objects that represent the underlying pixal data of an area image element. Examples include face recognition, label classifications(dog vs cat), autonomous driving, Quality Control in manufacturing, target identification, threat assessment, Healthcare,



Network Data: Data extracted from connected network devises and is limited to anonymized personal data.



Quantitative[Discrete, Continuous Data]: Expressed by a number and measured by numerical variables, How many, how much, how often.

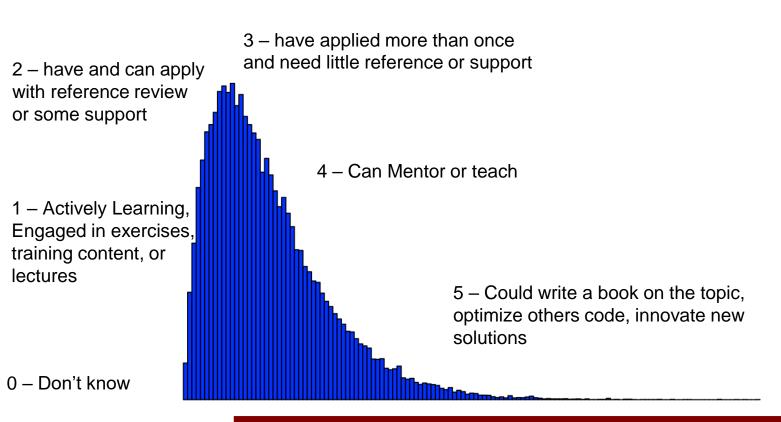
Discrete Data – Number of employees at a company, Number of movies on streaming platform

Continuous Data – Tenure of employees at a company, Length of movies on streaming platform

Qualitative/Categorical[Nominal Data Ordinal Data]: cannot be expressed as a number and cannot be measured. Qualitative consists of words, symbols, or pictures. What is? Who is? Where is?

Ordinal: Place numbers in order but math cannot be done on the order – Heat setting on stove; low, medium, high

Nominal: Used for Labeling variables – Type of food cooked: Thia, Mexican, Indian



- 1) 8 data types were described, name as many as you can recall.
- 2) What is the difference between Qualitative and Quantitative?
- 3) What is the difference between Ordinal and Nominal?
- 4) Are Ordinal and Nominal considered Quantitative or Qualitative?
- 5) What is the difference Discreet and continuous?
- 6) Are Discreet and Continuous Quantitative or Qualitative?

Classroom Exercise

- 1. Download the titanic dataset
- 2. Identify a question
- 3. What are observations made about the data?
- 4. What next steps would you make?

https://github.com/Morrisdata/data/blob/main/titanic.xls

GENERAL ASSEMBLY

Tasks

Familiarize with Git content
Complete any Python pre-work content
Connect with Students
Set up folder structures for ease of Content access
Consider the final project and a data set you are interested in

Matthew Morris

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