You will spend nearly 70% of your time doing this! Organize & pre-process your Data

Rutgers Libraries - NB Data Science Workshop Series

Pratiksha Sharma Oct 13, 2022

Fall 2022 Hours

Pratiksha Sharma - Data Science Graduate Specialist

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Topics: Data Science, Tableau, Python, SQL & NoSQL Databases

Office Hours (by appointment):

Thursday 12:30 - 01:00 pm (on days when workshop ends at 12:30 pm)

Thursday 01:00 - 01:30 pm (on days when workshop ends at 01:00 pm)

General Consultation: Request an appointment via email

Location:

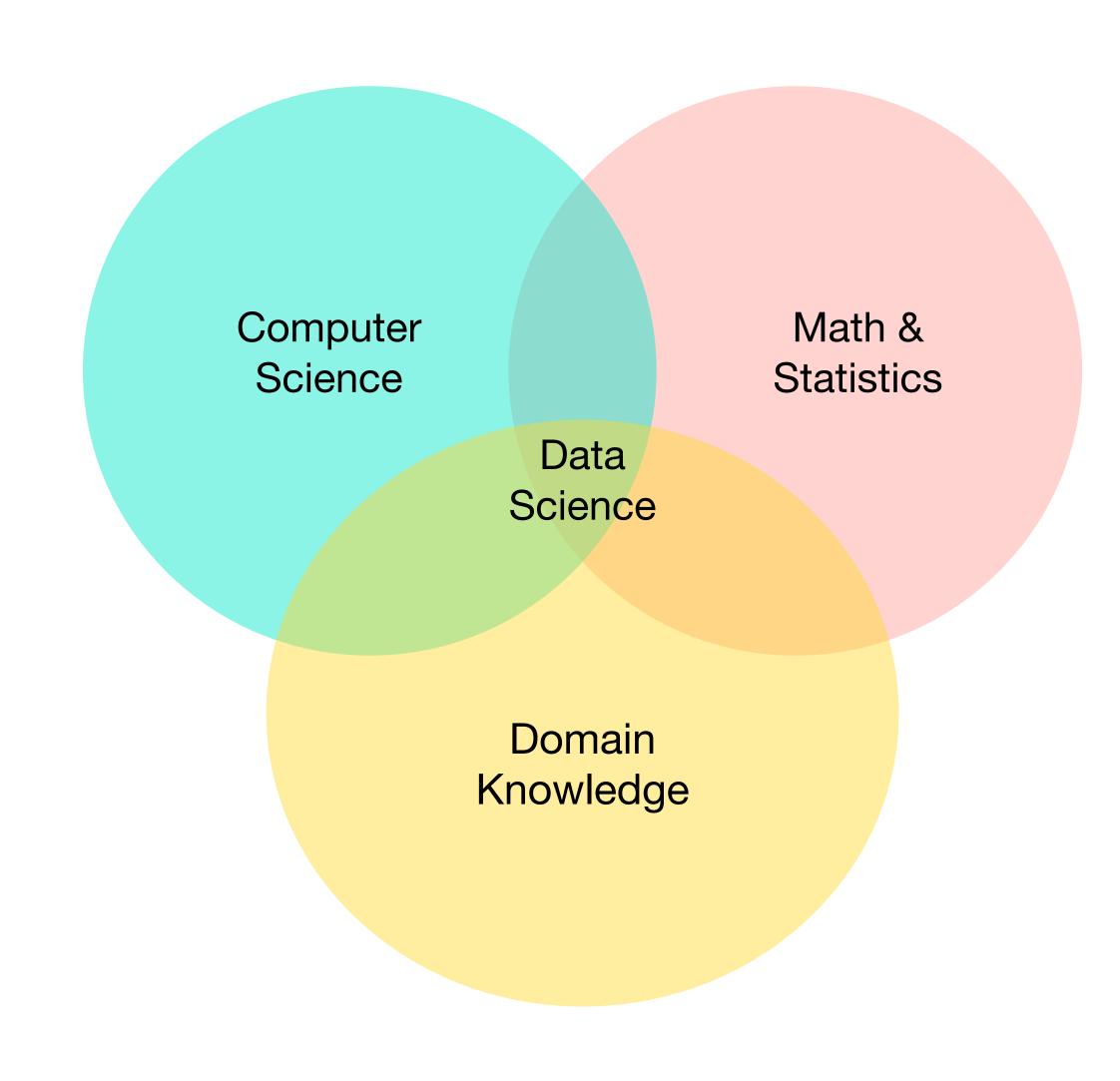
Zoom Meeting Link

Meeting ID: 926 5210 0393

Passcode: 772895

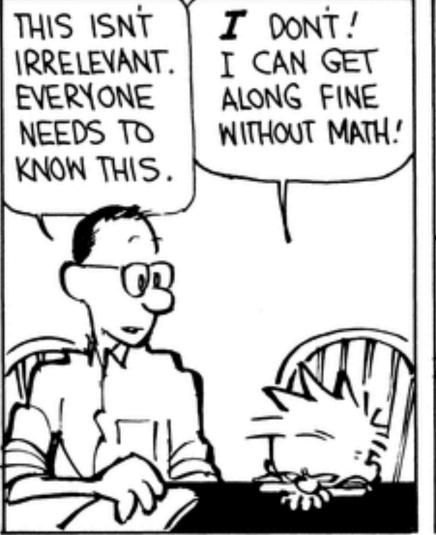
Quick Look: Data Science What's all the fuss about?

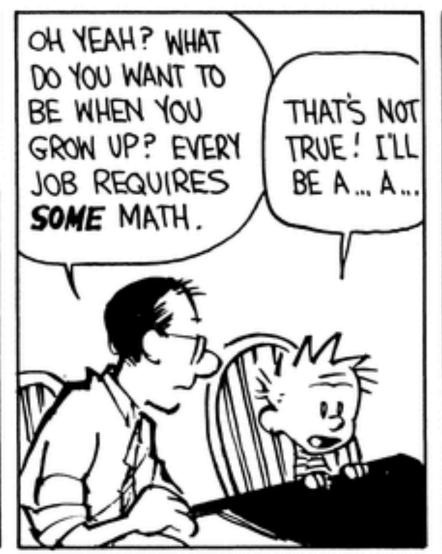
- A combination of Maths & Statistics, Computer Science and Domain Knowledge.
- This workshop is a combination of Computer Science and Maths & Statistics!
- You don't need to be an expert but data science is a part of everyday life!



Motivation









Source: Bill Waterson | Universal Press Syndicate

Before we begin..

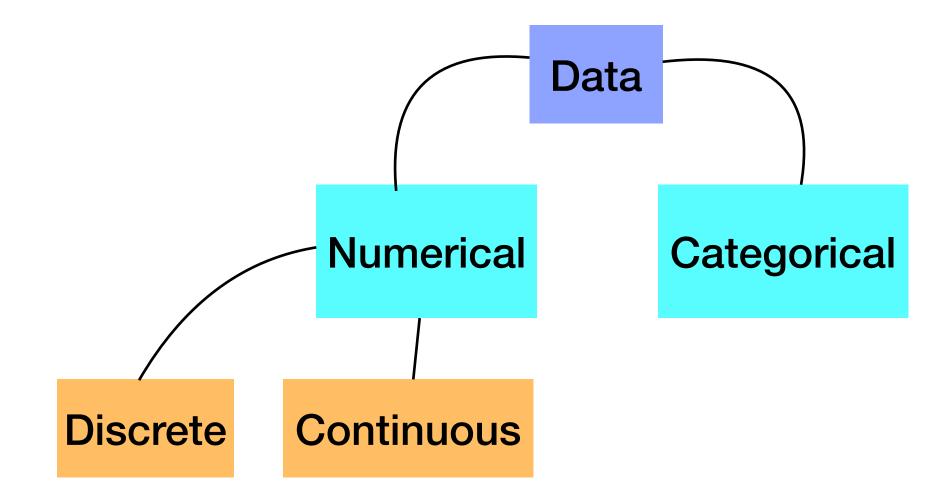
- We will be using MS Excel in this workshop
- Particularly some examples from the Analysis ToolPak: <u>Load Analysis ToolPak</u> for Excel
- You are encouraged to participate & follow along!
- Data would be provided in the Zoom chat & will be available later on with the workshop materials on: <u>Rutgers Libguides Data Science Workshops</u>

Data Cleaning, organizing & pre-processing We will cover..

- Quick Look: Data & its types
- Current challenges
- Specific Examples: Problems & Solutions
- Linear Regression: Example with MS Excel Analysis ToolPak

Data & its types

Data can be categorized into two types:



Data & its types Gold! Data Ratio **Nominal** Weight RUID Zipcode Price/Income **Ordinal** Interval **Temperature Shoe Size** Rank Age (not Kelvin)

Current Challenges

Cross-country survey data:

Customer ID	Name	Region	Phone
100	Jim Pembry	North America	732-790-6574
200	Gavin Andrews	Australia	61(460)882410
300	Shruti Gupta	India	8765607003
400	Van Dyke	UK	(+)442079357865
500	Heon Wang	China	8.62165E+11

• Do you see major problems?

Current Challenges

- The data that is generated today, has 3 main properties:
 - 1. Variety:

Data is being generated and collected from various sources & various formats

2. Volume

Data is being generated in huge quantities; to the extent that there is a whole field of research & study dedicated to storage & retrieval of this massive amount of data

3. Velocity

Data is not just being generated in huge quantities, but at an unprecedented rate; to the extent that there is a whole field of research & study dedicated to high velocity collection of data

Current Challenges

- Untreated/Raw data is often disintegrated, misleading & in most cases, not useful
- Data of this kind cannot be necessarily modeled; think about creating wrong models that work for wrong data
- Unprocessed data may restrict you from performing the necessary analysis for achieving your goals: think about visualizing categorical data, outliers skewing a distribution plot, etc.
- With the data that is being generated today, manual treating may not help
- Sophisticated methods are required to deal with bigger problems

Specific Examples: Problems & Solutions Faulty values

- The dataset on the right has faulty values
- Causes: Incorrect recording device, human error etc.
- Possible Solutions:
 - 1. Delete the outliers/faulty rows
 - 2. Substitution
 - 1. Mean when the data is usually similar with no strong low or high values
 - 2. Median when data has peculiar low or high values that have a potential of providing a misleading mean
 - 3. Mode think about categorical data, frequency data

tudent heights	
158.48	
143.1	
155.92	
155.57	
147.02	
271	
155.37	
148.04	
150.98	
162.82	
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155.75	
151.13	
146.93	
164.79	
157.14	
158.84	
60.46	
145.00	
153.26	
154.06	
152.37	
158.12	
158 11	
281.5	
151.77	
147.96	
162.05	
162.91	
150.2	

Integrating different variety

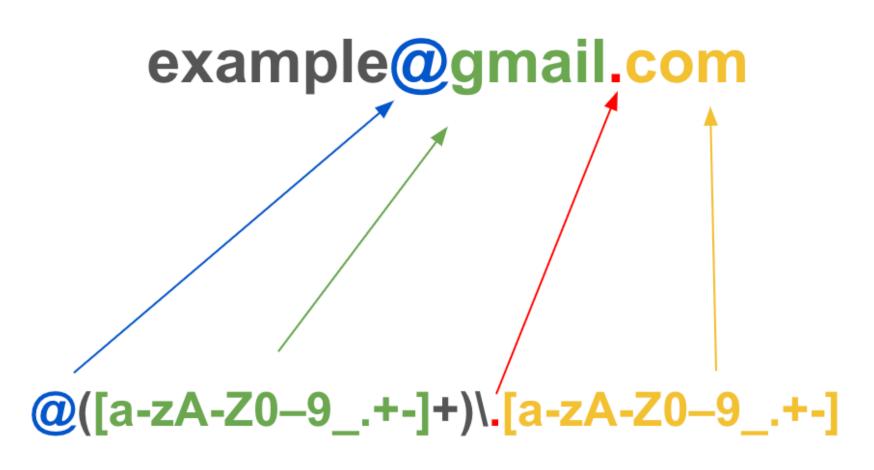
Come back to this example. How can you help?

Customer ID	Name	Region	Phone
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- Use the region to pick a pattern for text matching, capture important information & retain in desirable format
- The solution: Regex

Specific Examples: Problems & Solutions Integrating different variety

- Regex stands for Regular Expression
- It is a 'pattern matching' technique
- You have data in various 'exact' forms but in some similar sort of fashion
- One can use little patterns to provide a general format of how data could look like
- Regex is much more powerful than this!
- Let's jump to a movies dataset -



Look at a particular dataset

- Movies Dataset has:
 - 1. Columns with a lot of NA values
 - 2. Rows with irrelevant data
 - 3. Disintegrated formats of data capture
 - 4. Categorical to numeric conversion: Dummy conversion, one-hot encoding

Look at a particular dataset

One-hot encoding:

Label Encoding

Food Name	Categorical #	Calories
Apple	1	95
Chicken	2	231
Broccoli	3	50

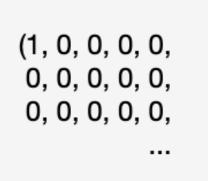
One Hot Encoding

Apple	Chicken	Broccoli	Calories
1	0	0	95
0	1	0	231
0	0	1	50

GenreVector W

(1, 1, 1, 0, 0,	
0, 0, 0, 0, 0,	
0, 0, 0, 0, 0,	

(0, 0, 0, 1, 1,	
0, 0, 0, 0, 0,	
0, 0, 0, 0, 0,	



(0, 0, 1, 0, 0,

Look at a particular dataset

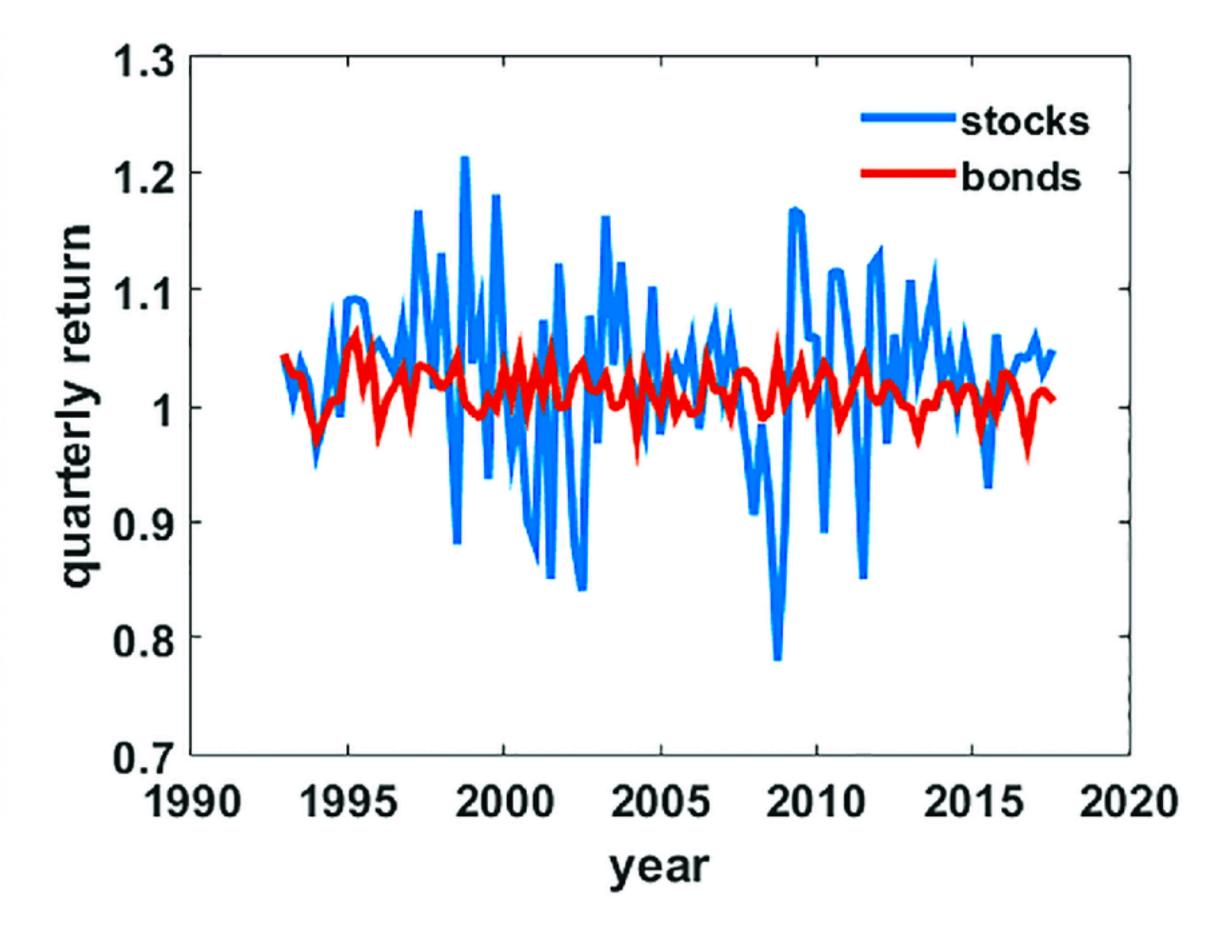
- Regex matching for more complex data
- Thinking about using a combination of different methods: first regex, then frequency filling for NA values etc.
- Think about a case where the data itself is correct; but problematic
- Normalisation
- SMOTE

Look at a particular dataset

Normalisation:



Image Source: Guy Metcalfe | researchgate.net



SMOTE: Problem of Imbalanced Data

- Data: 997 cat pictures, 3 dog pictures
- You make a model: (the specifics of the model are out of scope)
- You make a model with superb accuracy!
- A model that looks at every image & says this is a CAT!
- Accuracy = 99.7 % AMAZING?!
- What about the Dog class?
- Enter: SMOTE

SMOTE: Problem of Imbalanced Data

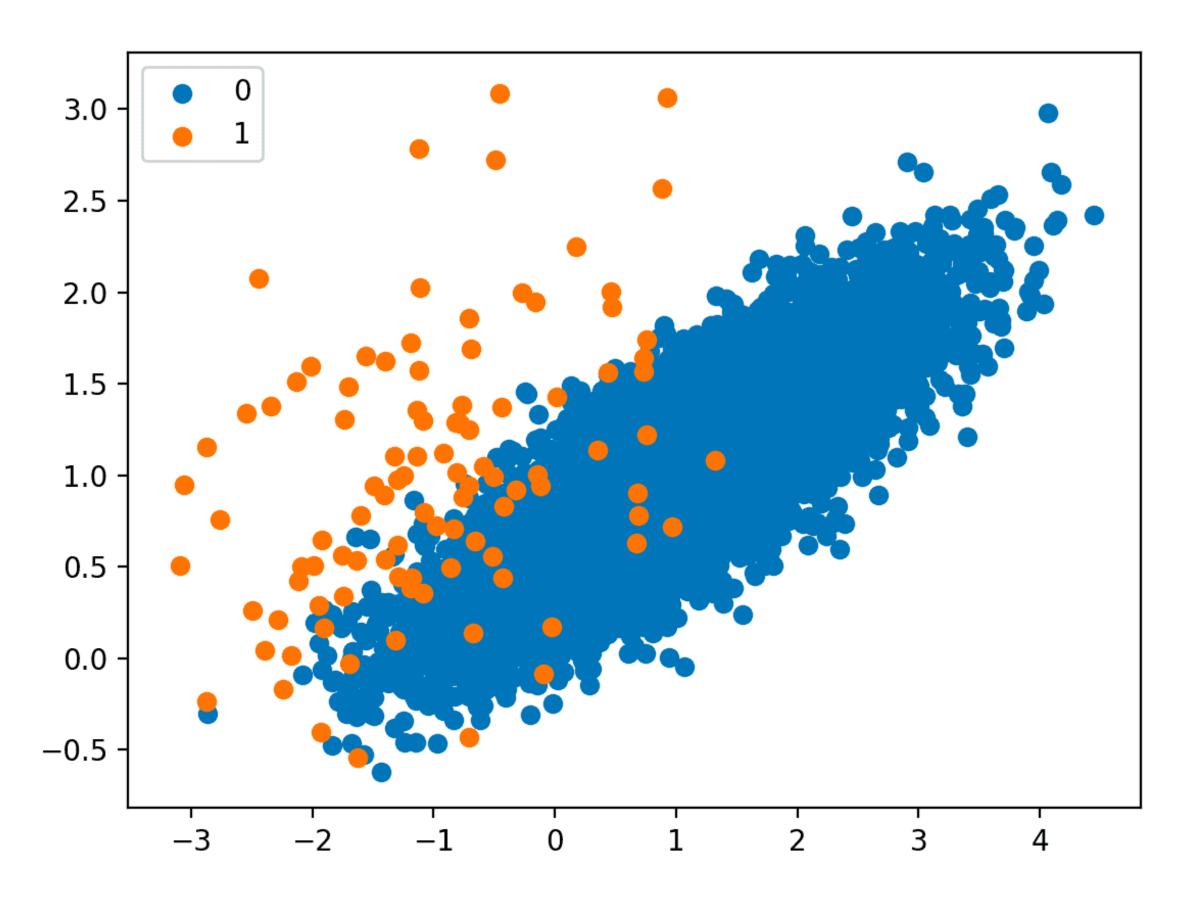


Image Source: https://machinelearningmastery.com/undersampling-algorithms-for-imbalanced-classification/

Problem of Imbalanced Data

- SMOTE: Synthetic Minority Oversampling Technique
- For imbalanced classification cases, SMOTE has the goal to increase more examples of underrepresented data i.e. minority data
- Some applications: Fraud Classification, Rare Virus finding etc.
- SMOTE uses a technique where it will look for data points most similar to a
 data point from a minority class, and will generate synthetic samples between
 the actual and the neighboring data points
- Repeat SMOTE to say double, or triple current representation of minority data to make a better model

Feature Engineering

- Sometimes, you have data that by itself does not make a lot of sense or does not necessarily give you a good model
- Transformations don't improve performance too
- What do you do?
- Feature engineering concentrates on combining one or more multiple features to create a higher, better performing feature

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Linear Regression

- Some feauture engineering
- Discussion of dealing with categorical data
- Linear Regression model with Numeric data

Takeway:

Modeling is not as straightforward as it seems!

Upcoming Workshops

https://libcal.rutgers.edu/nblworkshops

- The Power of Visual Storytelling: Learning Tableau Public: Oct 20
- "Make your computer work for you!" Learn how to use Python to program your tasks- Part 1: Oct 27
- "Make your computer work for you!" Explore popular Data Science libraries in Python - Part 2: Nov 03

Feedback Form

https://rutgers.libwizard.com/f/graduate_specialist_feedback