LECTURE 4

Data Wrangling and EDA, Part I

Exploratory Data Analysis and its role in the data science lifecycle

Data 100/Data 200, Spring 2023 @ UC Berkeley

Narges Norouzi and Lisa Yan

Content credit: Lisa Yan, Josh Hug



Interlude

Real World Data Scientist:

Timnit Gebru

- Gender Shades (with Joy Buolamwini, 2300+ citations)
- <u>Datasheets for Datasets</u> (580+ citations)

The characteristics of these datasets fundamentally influence a model's behavior: a model is unlikely to perform well in the wild if its deployment context does not match its training or evaluation datasets, or if these datasets reflect unwanted societal biases.

We need EDA because many datasets don't come with comprehensive datasheets/codebooks!









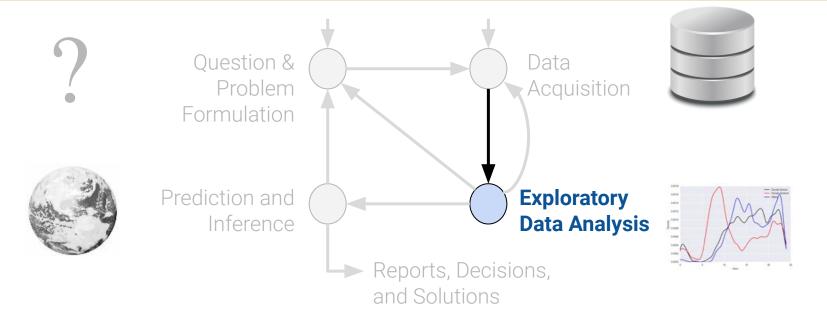
EDA Guiding Principles

The Next Step



Plan for first few weeks





(Weeks 1 and 2)

Exploring and Cleaning Tabular Data From datascience to pandas



(Weeks 2 and 3)

Data Science in Practice

EDA, Data Cleaning, Text processing (regular expressions)





Exploring Tabular Data

Lecture 04, Data 100 Spring 2023

Exploring Tabular Data

Record Granularity

Variable Types

Multiple Files

More EDA/Wrangling

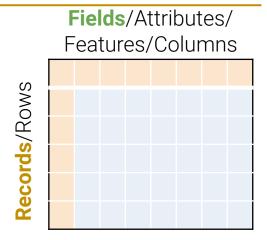


Rectangular Data

We prefer rectangular data for data analysis (why?)

- Regular structures are easy manipulate and analyze
- A big part of data cleaning is about transforming data to be more rectangular

Two kinds of rectangular data: **Tables** and **Matrices**.



Tables (a.k.a. dataframes in R/Python and relations in SQL)

- Named columns with different types
- Manipulated using data transformation languages (map, filter, group by, join, ...)

Matrices

- Numeric data of the same type (float, int, etc.)
- Manipulated using linear algebra

What are the differences?
Why would you use one over the other?



Tuberculosis - United States, 2021



Summary

What is already known about this topic?

The number of reported U.S. tuberculosis (TB) cases decreased sharply in 2020, possibly related to multiple factors associated with the COVID-19 pandemic.

What is added by this report?

Reported TB incidence (cases per 100,000 persons)

increased 9.4%, from 2.2 during 2020 to 2.4 during 2021 but was lower than incidence during 2019 (2.7). Increases

occurred among both U.S.-born and non–U.S.-born persons.

What are the implications for public health practice?

Factors contributing to changes in reported TB during 2020–2021 likely include an actual reduction in TB incidence as well as delayed or missed TB diagnoses. Timely evaluation and treatment of TB and latent tuberculosis infection remain critical to achieving U.S. TB elimination.

CDC Morbidity and Mortality Weekly Report (MMWR) 03/25/2022.

What is **incidence**? Why use it here?

How was "9.4% increase" computed?

Question: Can we **reproduce** these rates using government data?



CDC: source, 2021

Demo Slides

CSV: Comma-Separated Values



Tuberculosis in the US [CDC source].

CSV is a very common tabular file format.

- Records (rows) are delimited by a newline: '\n', "\r\n"
- Fields (columns) are delimited by commas: ', '

Pandas: pd.read_csv (header=...)

Fields/Attributes/Features/Columns

Rows		U.S. jurisdiction	TB cases 2019	
ords/F	0	Total	8,900	
Reco	1	Alabama	87	





Exploring Tabular Data

Record Granularity

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Multiple Files

More EDA/Wrangling

Record Granularity

Lecture 04, Data 100 Spring 2023





(we'll come back to this)





Granularity -- how fine/coarse is each datum

Scope -- how (in)complete is the data

Temporality -- how is the data situated in time

Faithfulness -- how well does the data capture "reality"

Key Data Properties to Consider in EDA



Granularity: How fine/coarse is each datum?



What does each **record** represent?

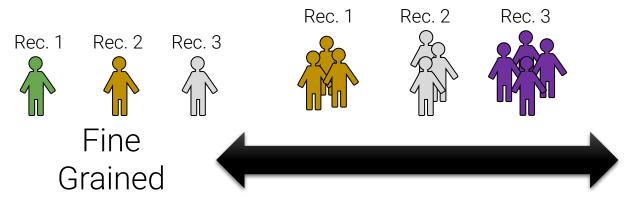
Examples: a purchase, a person, a group of users

Do all records capture granularity at the same level?

Some data will include summaries (aka rollups) as records.

If the data are **coarse**, how were the records aggregated?

Sampling, averaging, maybe some of both...





To the demo!!





Variable Types

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Exploring Tabular Data Record Granularity

Variable Types

Multiple Files

More EDA/Wrangling



(we're back to this)

Variable Type





Structure -- the "shape" of a data file

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Variables are Columns

Let's look at records with the same granularity.

What does each **column** represent?

A **variable** is a **measurement** of a particular concept.

 U.S. jurisdiction
 TB cases 2019
 ...

 1
 Alabama
 87
 ...

 2
 Alaska
 58
 ...

 ...
 ...
 ...
 ...

The U.S. Jurisdiction variable

It has two common properties:

Datatype/Storage type:

How each variable value is stored in memory. dffcolname].dtype

o integer, floating point, boolean, object (string-like), etc.

Affects which pandas functions you use.

Variable type/Feature type:

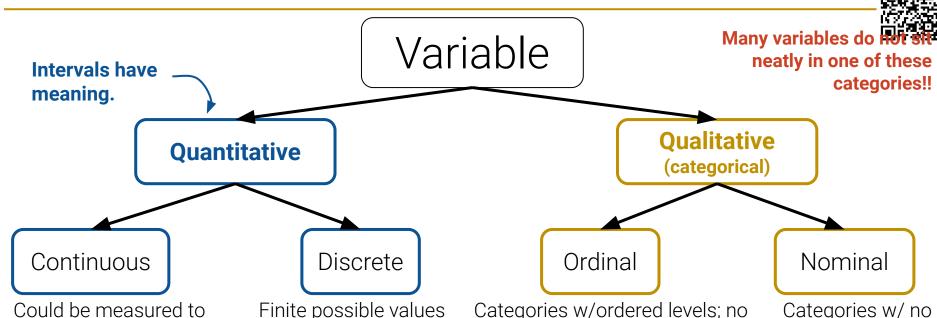
Conceptualized measurement of information (and therefore what values it can take on).

- Use expert knowledge
- Explore data itself
- Consult data codebook (if it exists).

Affects how you visualize and interpret the data.

In this class, "variable types" are conceptual!!

Variable Feature Types



Examples:

- Price
- Temperature

arbitrary precision.

Examples:

- Number of siblings
- Yrs of education

Examples:

Preferences

consistent meaning to difference

Level of education

Examples:

Political Affiliation

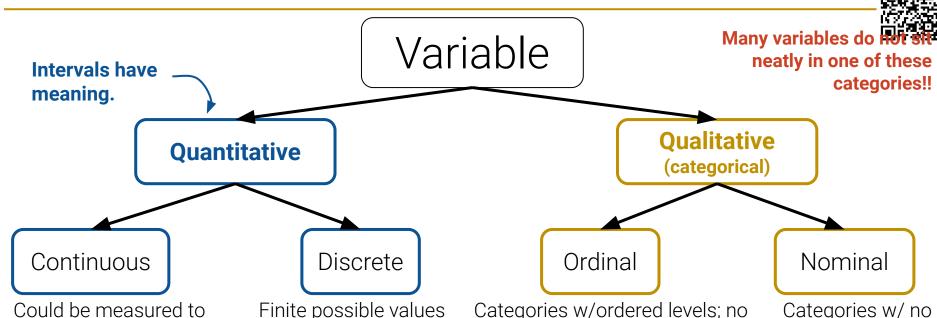
specific ordering.

Cal ID number

Note that **qualitative variables** could have numeric levels; conversely, **quantitative variables** could be stored as strings!



Variable Feature Types



Examples:

- Price
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arbitrary precision.

Examples:

- Number of siblings
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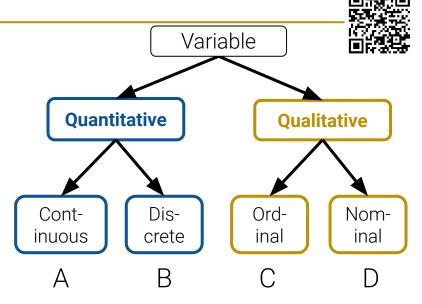
Note that **qualitative variables** could have numeric levels; conversely, **quantitative variables** could be stored as strings!



Variable Types

What is the feature type (i.e., variable type) of each variable?

Q	Variable	Feature Type
1	CO ₂ level (ppm)	
2	Number of siblings	
3	GPA	
4	Income bracket (low, med, high)	
5	Race/Ethnicity	
6	Number of years of education	
7	Yelp Rating	





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Variable Types

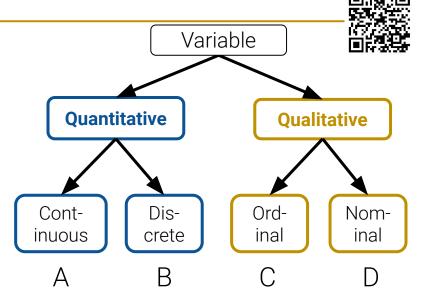
① Start presenting to display the poll results on this slide.



Variable Types

What is the feature type of each variable?

Q	Variable	Feature Type
1	CO ₂ level (ppm)	A. Quantitative Cont.
2	Number of siblings	B. Quantitative Discrete
3	GPA	A. Quantitative Cont.
4	Income bracket (low, med, high)	C. Qualitative Ordinal
5	Race/Ethnicity	D. Qualitative Nominal
6	Number of years of education	B. Quantitative Discrete
7	Yelp Rating	C. Qualitative Ordinal



Many of these examples show how "shaggy" these categories are!!
We will revisit variable types when we learn how to visualize variables.





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Exploring Tabular Data
Record Granularity
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Multiple Files

More EDA/Wrangling

Multiple Files

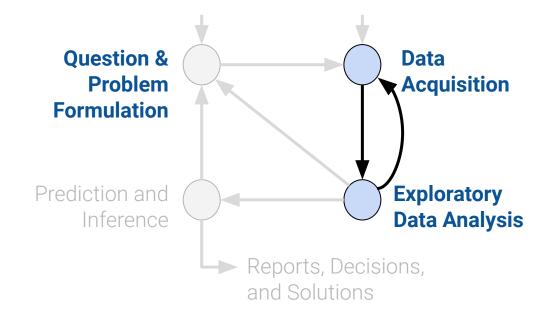
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The Data Science Lifecycle is a Cycle



In practice, EDA informs whether you need more data to address your research question.





What is incidence?



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CDC: **source**, 2021

Defining incidence



From the <u>CDC report</u>: **TB incidence** is computed as "Cases per 100,000 persons using mid-year population estimates from the U.S. Census Bureau."

Incidence is useful when comparing case rates across differently sized populations.

TB incidence =
$$\frac{\text{# TB cases in population}}{\text{# groups in population}}$$
 (group: 100,000 people)
$$= \frac{\text{# TB cases}}{\text{(population/100,000)}}$$
 =
$$\frac{\text{# TB cases}}{\text{population}}$$
 x 100,000 population

We don't have U.S. Census population data in our DataFrame.

We need to acquire it to verify incidence!





Demo Slides



Structure: Primary Keys and Foreign Keys

Customers.csv

Sometimes your data comes in multiple files:

- Often data will reference other pieces of data.
- Alternatively, you will collect multiple pieces of related data.

Use <u>pd.merge</u> to **join** data on **keys**.

		3.5
ustID	Addr	
1345	Harmon	ا
1139	Main	

Orders.csv

<u>OrderNum</u>	<u>CustID</u>	Date
1	171345	8/21/2017
2	281139	8/30/2017

Products.csv

<u>ProdID</u>	Cost
42	3.14
999	2.72

Purchases.csv

<u>OrderNum</u>	<u>ProdID</u>	Quantity
1	42	3
1	999	2
2	42	1



Structure: Primary Keys and Foreign Keys

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- Often data will reference other pieces of data.
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Use pd.merge to join data on keys.

Primary key: the column or set of columns in a table that determine the values of the remaining columns

- Primary keys are unique, but could be tuples.
- Examples: SSN, ProductIDs, ...

Primary Key

Primary Key	C	ustomers.csv	蒙
	<u>CustID</u>	Addr	
	171345	Harmon	
Primary Key	281139	Main	
Timaly Roy		Orders.csv	
<u>OrderNum</u>	<u>CustID</u>	Date	
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Products.csv

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Purchases.csv

		' '	JICI 103C3.C3V
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Structure: Primary Keys and Foreign Keys

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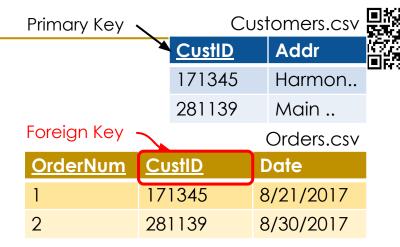
Use pd.merge to join data on **keys**.

Primary key: the column or set of columns in a table that determine the values of the remaining columns

- Primary keys are unique, but could be tuples.
- Examples: SSN, ProductIDs, ...

Foreign keys: the column or sets of columns that reference primary keys in other tables.

More later when we see SQL. Stay tuned!



Products.csv
ProdID Cost
42 3.14
999 2.72

Purchases.csv

<u>OrderNum</u>	ProdID	Quantity
1	42	3
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More EDA/Wrangling

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More next time







Structure -- the "shape" of a data file

Granularity -- how fine/coarse is each datum

Scope -- how (in)complete is the data

Temporality -- how is the data situated in time

What else?



Faithfulness -- how well does the data capture "reality"



Faithfulness: Do I trust this data?



Does my data contain **unrealistic or "incorrect" values**?

- Dates in the future for events in the past
- Locations that don't exist
- Negative counts
- Misspellings of names
- Large outliers

Does my data violate **obvious dependencies**?

E.g., age and birthday don't match

You will explore this more in homework. Stay tuned!

Was the data **entered by hand**?

- Spelling errors, fields shifted ...
- Did the form require all fields or provide default values?

Are there obvious signs of **data falsification**?

 Repeated names, fake looking email addresses, repeated use of uncommon names or fields.



LECTURE 4

Data Wrangling and EDA, Part I

Content credit: Joseph Gonzalez, Deborah Nolan, Joseph Hellerstein, and Lisa Yan

