#### **LECTURE 6**

# Data Cleaning and EDA

Exploratory data analysis and its role in the data science lifecycle.

#### Data 100/Data 200, Fall 2021 @ UC Berkeley

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### Pandas and Jupyter Notebooks

- Introduced DataFrame concepts
  - Series: A named column of data with an index
  - **Indexes**: The mapping from keys to rows
  - **DataFrame**: collection of series with common index
- Dataframe access methods
  - Filtering on predicts and slicing
  - o **df.loc**: location by index
  - df.iloc: location by integer address
  - groupby & pivot aggregating data



## Today





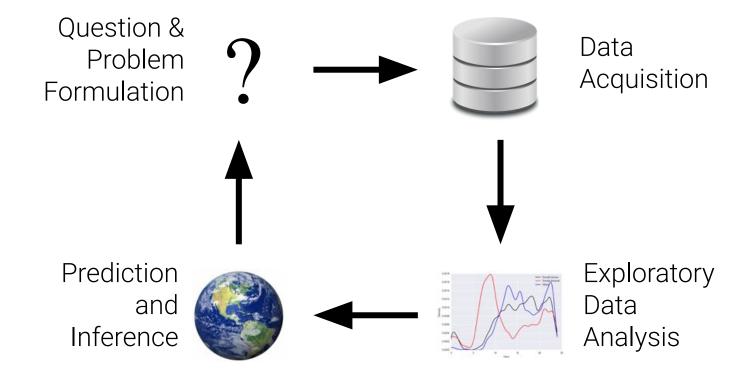
## Congratulations!



You have **collected** or **been given** a box of data?

What do you do next?









## Data Acquisition



## Exploratory Data Analysis

## Topics For This Lecture

- Understanding the Data
  - Data Cleaning
  - Exploratory Data Analysis (EDA)
  - Basic data visualization
- Common Data Anomalies
  - ... and how to fix them





Data Cleaning

## Exploratory Data Analysis





### Data Cleaning

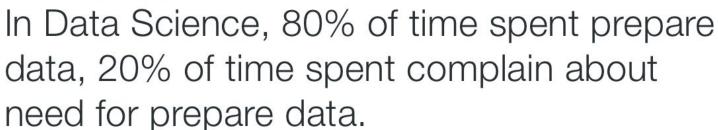
- The process of transforming raw data to facilitate subsequent analysis
- Data cleaning often addresses issues
  - structure / formatting
  - missing or corrupted values
  - unit conversion
  - encoding text as numbers
  - 0 ...
- Sadly, data cleaning is a big part of data science...





#### Big Data Borat





 $\Diamond$ 

**Following** 













Data Cleaning

## Exploratory Data Analysis



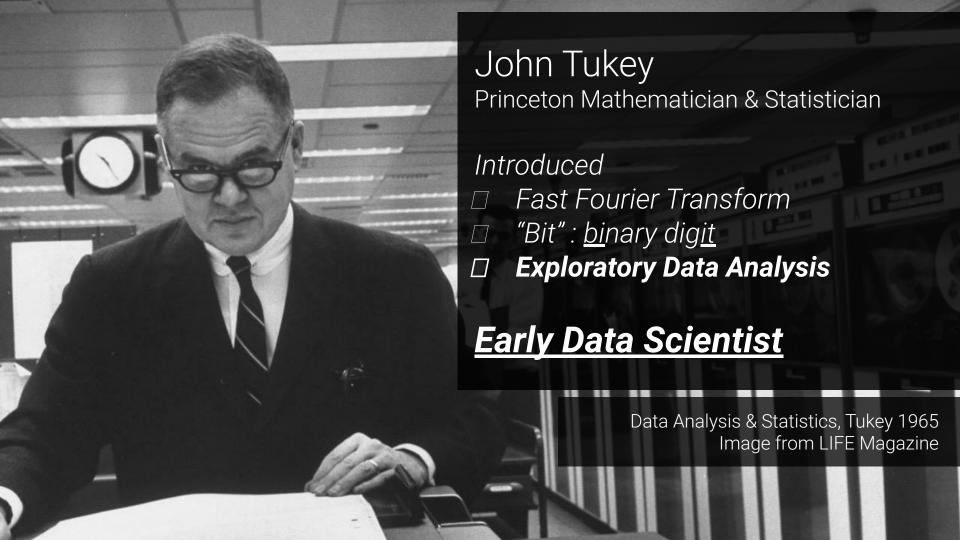


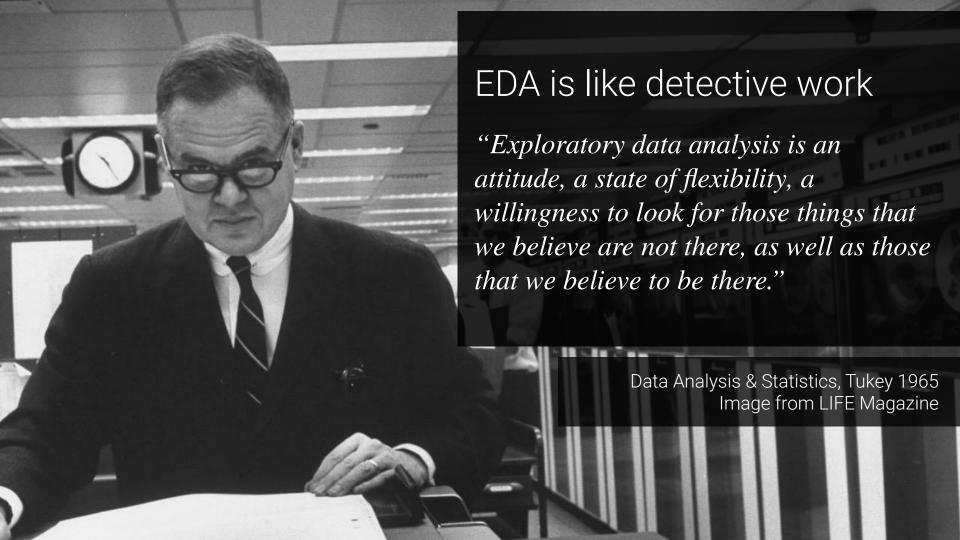
## Exploratory Data Analysis (EDA)

#### "Getting to know the data"

- The process of transforming, visualizing, and summarizing data to:
  - Build/confirm understanding of the data and its provenance
  - Identify and address potential issues in the data
  - Inform the subsequent analysis
  - discover potential hypothesis ... (be careful)
- EDA is an open-ended analysis
  - Be willing to find something surprising







## File Formats and Structure



# What should we look for?



### Key Data Properties to Consider in EDA

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



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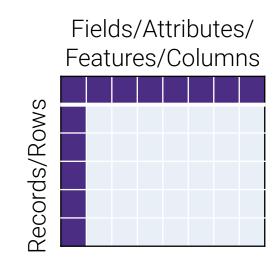


### Rectangular Data

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* (what are the differences?)





### Rectangular Data

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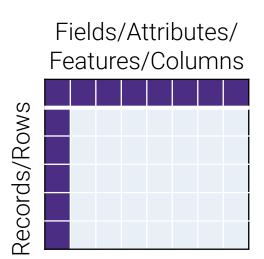
- Regular structures are easy manipulate and analyze
- A big part of data cleaning is about transforming data to be more rectangular

Two kinds of rectangular data: *Tables and Matrices* (what are the differences?)

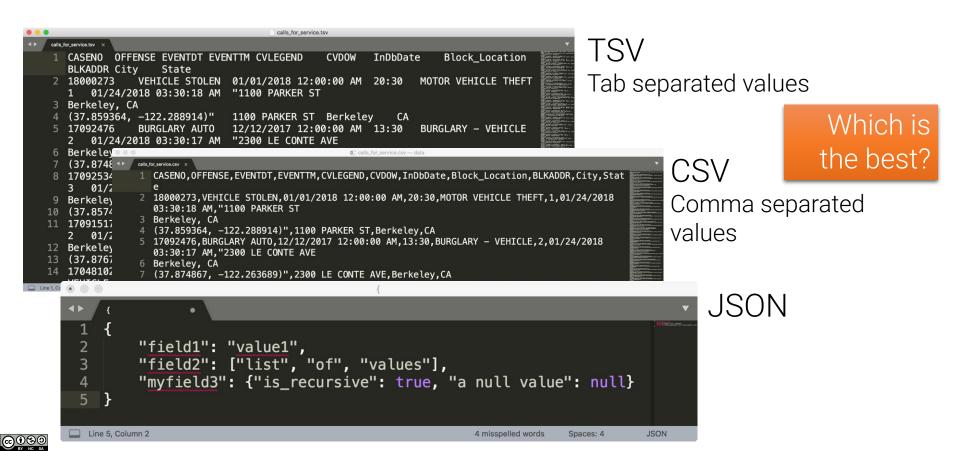
- 1. Tables (a.k.a. data-frames in R/Python and relations in SQL)
  - Named columns with different types
  - Manipulated using data transformation languages (map, filter, group by, join, ...)

#### 2. Matrices

- Numeric data of the same type
- Manipulated using linear algebra

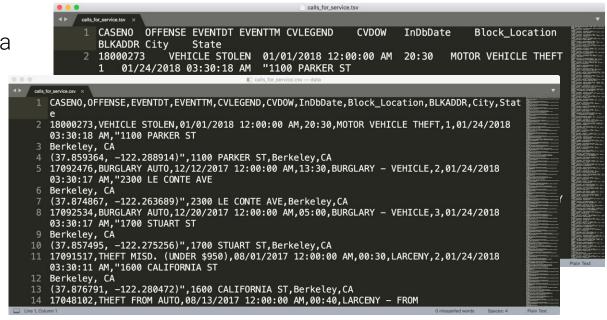


#### How are these data files formatted?



#### Comma and Tab Separated Values Files

- Tabular data where
  - Records are delimited by a newline: "\n", "\r\n"
  - Fields are delimited by ',' (comma) or '\t' (tab)
- Very Common!
- Issues?
  - Commas, tabs in records
  - Quoting
  - 0 ..





## JavaScript Object Notation (JSON)

```
## Company Colors of the color of the c
```

- Widely used file format for nested data
  - Very similar to python dictionaries
  - Strict formatting "quoting" addresses some issues in CSV/TSV
- Issues
  - Not rectangular
  - Each record can have different fields
  - Nesting means records can contain tables complicated



## Extensible Markup Language - XML (another kind of nested data)

```
<catalog>
  <plant type='a'>
    <common>Bloodroot</common>
    <botanical>Sanguinaria canadensis/botanical>
    <zone>4</zone>
    dight>Mostly Shady</liqht>
    <price>2.44</price>
    <availability>03/15/2006</availability>
    <description>
     <color>white</color>
                                        Nested structure
     <petals>true</petals>
    </description>
    <indoor>true</indoor>
  </plant>
</catalog>
```



### Log Data

# Is this a csv file? tsv? JSON/XML?

```
169.237.46.168 - - [26/Jan/2014:10:47:58 -0800] "GET /stat141/Winter04 HTTP/1.1" 301 328 "http://anson.ucdavis.edu/courses/" "Mozilla/4.0 (compatible; MSIE 6.0; Windows NT 5.0; .NET CLR 1.1.4322)"
```

```
169.237.6.168 - - [8/Jan/2014:10:47:58 -0800] "GET /stat141/Winter04/ HTTP/1.1" 200 2585
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```



# Keys and Joins



#### Primary Key

### Structure: Keys

- Often data will reference other pieces of data
- Primary key: the column or set of columns in a table that determine the values of the remaining columns
  - Primary keys are unique
  - Examples: SSN, ProductIDs, ...

Purchases.csv

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

#### 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

#### Primary Key

#### Customers.csv

<u>CustID</u>	Addr
171345	Harmon
281139	Main



#### Primary Key

- Structure: Keys
- Often data will reference other pieces of data
- Primary key: the column or set of columns in a table that determine the values of the remaining columns
  - Primary keys are unique
  - Examples: SSN, ProductIDs, ...
- Foreign keys: the column or sets of columns that reference primary keys in other tables.
- You will need to join across tables

		1 010110303.034
<u>OrderNum</u>	<u>ProdID</u>	Quantity
1	42	3
1	999	2
2	42	1

Foreign Key		Orders.csv
<u>OrderNum</u>	<u>CustID</u>	Date
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Products.csv

Purchases CSV

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Primary Key

<b>X</b> C031	Officis.Csv
<u>CustID</u>	Addr
171345	Harmon

281139

Customers csv

Main ..



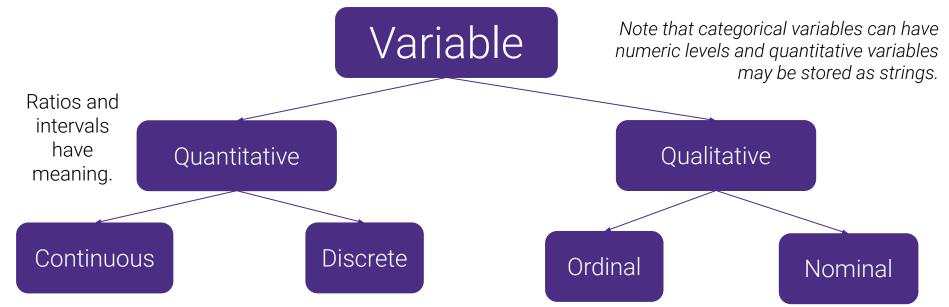
### Questions to ask about Structure

- Are the data in a standard format or encoding?
  - Tabular data: CSV, TSV, Excel, SQL
  - Nested data: JSON or XML
- Are the data organized in "records"?
  - No: Can we define records by parsing the data?
- Are the data nested? (records contained within records...)
  - Yes: Can we reasonably un-nest the data?
- Does the data reference other data?
  - Yes: can we join/merge the data
- What are the fields in each record?
  - How are they encoded? (e.g., strings, numbers, binary, dates ...)
  - What is the type of the data?



# Variable Types





Could be measured to arbitrary precision.

#### **Examples:**

- Price
- Temperature

Finite possible values

#### **Examples:**

- Number of siblings
- Yrs of education

Categories w/ levels but no consistent meaning to difference

#### **Examples:**

- Preferences
- Level of education

Categories w/ no specific ordering.

#### **Examples:**

- Political Affiliation
- CallD number



## What is the type of variable?

	Quantitative Continuous	Quantitative Discrete	Qualitative Ordinal	Qualitative Nominal
CO <sub>2</sub> level (PPM)				
Number of siblings				
GPA				
Income bracket (low, med, high)				
Race				
Number of years of education				
Yelp Rating				



# Granularity, Scope, and Temporality

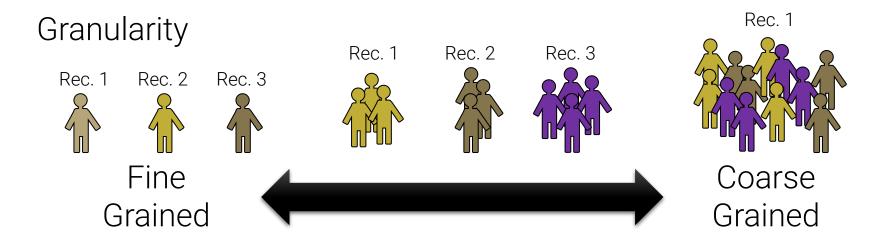


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- What does each record represent?
  - Examples: a purchase, a person, a group of users
- Do all records capture granularity at the same level?
  - o Some data will include summaries (aka rollups) as records
- If the data are coarse how was it aggregated?
  - Sampling, averaging, ...



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### Scope

- Does my data cover my area of interest?
  - **Example:** I am interested in studying crime in California but I only have Berkeley crime data.
- Is my data too expansive?
  - **Example:** I am interested in student grades for DS100 but have student grades for all statistics classes.
  - Solution: Filtering ⇒ Implications on sample?
    - If the data is a sample I may have poor coverage after filtering ...
- Does my data cover the right time frame?
  - More on this in temporality ...



## Revisiting the Sampling Frame

- The sampling frame is the population from which the data was sampled.
  - Note that this **may not be** the **population** of interest.
- How complete/incomplete is the frame (and its data)?
- How is the frame/data situated in place?
- How well does the frame/data capture reality?
- How is the frame/data situated in time?



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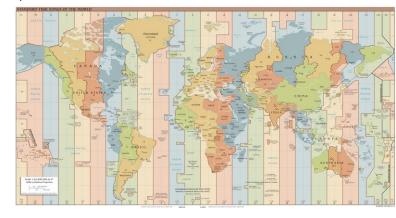
### Temporality

- Data changes when was the data collected?
- What is the meaning of the time and date fields?
  - When the "event" **happened**?
  - When the data was collected or was entered into the system?
  - Date the data was copied into a database (look for many matching timestamps)
- Time depends on where! (Time zones & daylight savings)
  - Learn to use datetime python library
  - Multiple string representation (depends on region): 07/08/09?
- Are there strange null values?
  - January 1<sup>st</sup> 1970, January 1<sup>st</sup> 1900
- Is there periodicity? Diurnal patterns



#### Unix Time / POSIX Time

- Time measured in seconds since January 1<sup>st</sup> 1970
  - Minus leap seconds ...
- Unix time follows Coordinated Universal Time (UTC)
  - International time standard
  - Measured at 0 degrees latitude
    - Similar to Greenwich Mean Time (GMT)
  - No daylight savings
  - Time codes
- Time Zones:
  - San Francisco (UTC-8)
     without daylight savings



https://en.wikipedia.org/wiki/Coordinated\_Universal\_Time



## Faithfulness and Missing Values



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#### 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?
  - o E.g., age and birthday don't match
- Was the data entered by hand?
  - Spelling errors, fields shifted ...
  - Did the form require 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.



## Signs that your data may not be faithful

- Missing Values/Default values?
  - What do they look like?
    - **.** " ",
    - **=** 0,
    - **■** -1, 999, 12345,
    - NaN, Null,
    - **1970, 1900**



### What to do with the Missing Values?

- Drop records with missing values
  - Probably most common
  - Caution: check for biases introduced by dropped values
    - Missing or corrupt records might be related to something of interest
- Imputation: (Inferring missing values)
  - Mean Imputation: replace with an average value
    - Which mean? Often use closest related subgroup mean.
  - Hot deck imputation: replace with a random value
    - Choose a random value from the subgroup and use it for the missing value.
- Prof. Gonzalez Suggestion:
  - Drop missing values but check for induced bias (use domain knowledge)
  - Directly model missing values during future analysis



## Signs that your data may not be faithful

- Missing Values or default values
- Truncated data (early excel limits: 65536 Rows, 255 Columns)
  - Soln: be aware of consequences in analysis ⇒ how did truncation affect sample?
- Time Zone Inconsistencies
  - Soln 1: convert to a common timezone (e.g., UTC)
  - Soln 2: convert to the timezone of the location useful in modeling behavior.
- Duplicated Records or Fields
  - Soln: identify and eliminate (use primary key) ⇒ implications on sample?
- Spelling Errors
  - Soln: Apply corrections or drop records not in a dictionary ⇒ implications on sample?
- Units not specified or consistent
  - Solns: Infer units, check values are in reasonable ranges for data
- Others...



## Summary



## Summary: How do you do EDA/Data Cleaning?

- Examine data and metadata:
  - What is the date, size, organization, and structure of the data?
- Examine each field/attribute/dimension individually
- Examine pairs of related dimensions
  - Stratifying earlier analysis: break down grades by major ...
- Along the way:
  - Visualize/summarize the data
  - Validate assumptions about data and collection process
  - Identify and address anomalies
  - Apply data transformations and corrections
  - Record everything you do! (why?)

