## Descriptive and Exploratory Analysis

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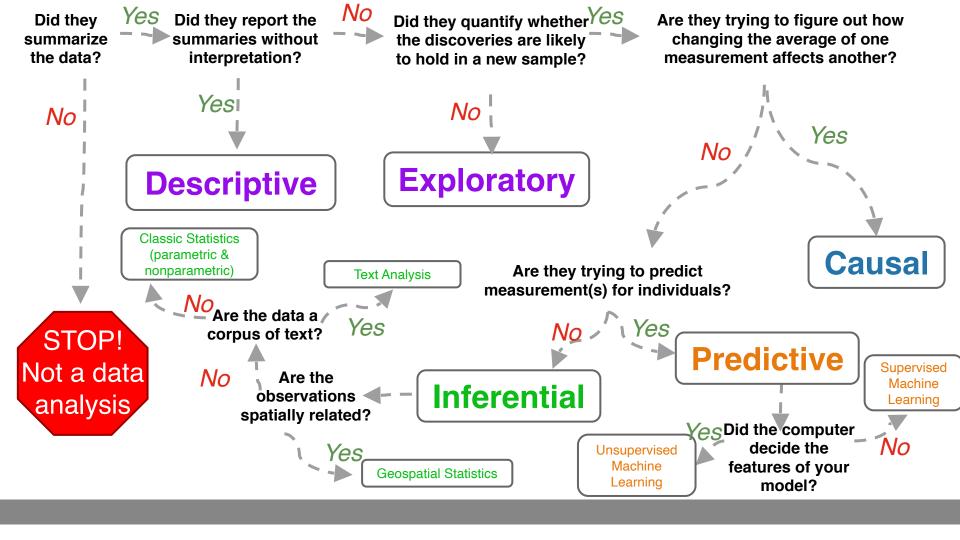


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"Data science is the process of formulating a quantitative question that can be answered with data, collecting and cleaning the data, analyzing the data, and communicating the answer to the question to a relevant audience."

To do this, you have to look at, describe, and explore the data



### Summary: Analytical Approaches

- 1. Descriptive (and Exploratory) Data Analysis are the first step(s)
- 2. Inference establishes relationships
  - a. Classic Statistics
  - b. Geospatial Analysis
  - c. Text Analysis
- 3. Machine Learning is for prediction
  - a. Supervised
  - b. Unsupervised
- 4. Experiments best way to establish causality

### **Exploring Analyses**

General question: What impacts politics in America?

Data Science question: Is there a relationship between the sentiment of political words in South Park and America's presidential approval rating?

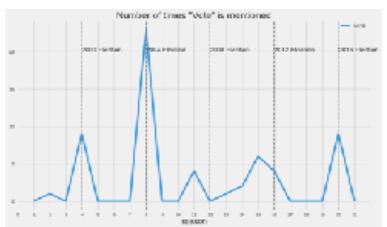
**Descriptive** 

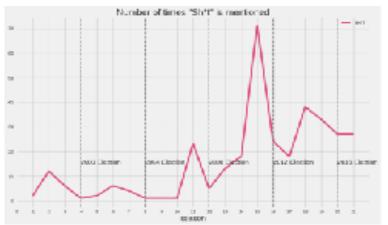
**Exploratory** 

**Inferential** 



Classic Statistics (parametric & nonparametric)





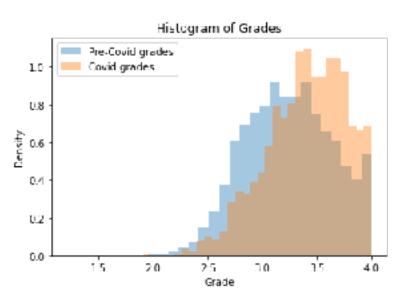
General question: How has COVID-19 impacted students?

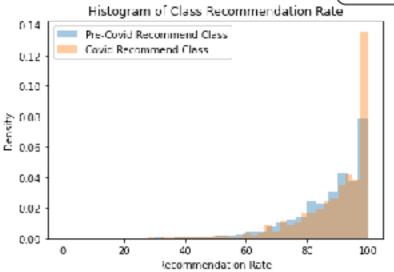
Data Science question: At UCSD, is there a difference between students' grades and how they rate their classes before COVID-19 and during remote learning, due to COVID-19?

### **Descriptive**

**Exploratory** 

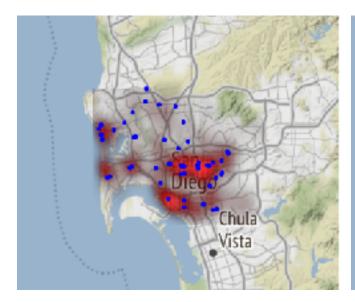
### **Inferential**

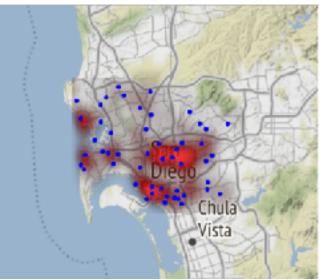




Classic Statistics (parametric & nonparametric) General question: Why isn't police response time always the same?

Data Science question: Where should police cars be stationed, accounting for crime levels and time of day, to make police response times equitable throughout San Diego?





### **Descriptive**

**Exploratory** 

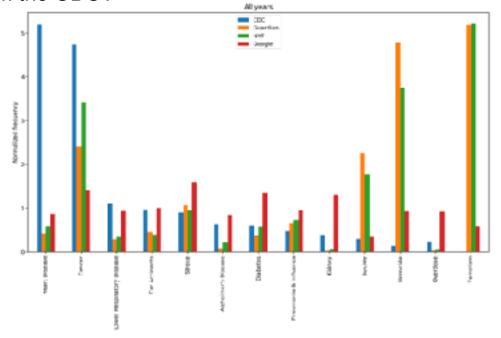
**Predictive** 

**Inferential** 

Geospatial Analysis

General question: What gets too much attention in the news?

Data Science Question: Is there a relationship over time between cause of death terms in the *NYT*, The Guardian, and Google trends data relative to data from the CDC?



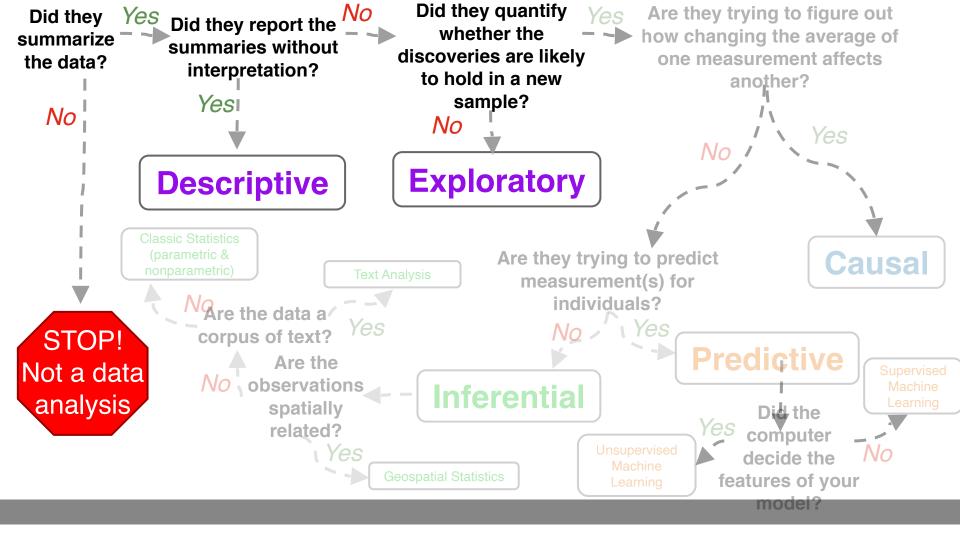
### **Descriptive**

### **Exploratory**

### **Inferential**

**Text Analysis** 

Classic Statistics (parametric & nonparametric)



### **Descriptive**

**Descriptive**: The goal of descriptive analysis is to understand the components of a data set, describe what they are, and explain that description to others who might want to understand the data.

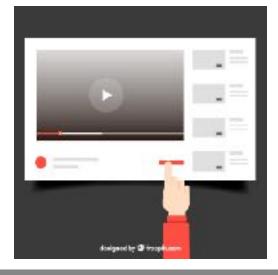
 Problem: Understanding whether users are nice or mean on Youtube

 Data science question: Are the words that people use in their comments more frequently positive words (great, awesome, nice, useful) or negative words (bad, stupid,

lame, awful)?

Type of analysis: Descriptive analysis

To answer this you would calculate <u>statistics</u> about YouTube comments



### **Statistics**

"the science that deals with the collection, classification, analysis, and interpretation of numerical facts or data"

statistic

"A quantity computed from a <u>sample</u>"

### statistic

### "A quantity computed from a sample"



For our YouTube analysis, we could take a <u>random sample</u> of comments from YouTube and calculate the following statistic: the number of positive and the number of negative words in each review.

Source: dictionary.com



### All comments on YouTube

During the second quarter of 2020, almost 2.13 billion comments on YouTube videos were removed due to violation of the platform's community guidelines. - J Clement on

We want to learn something about this...

Sampling Inference

....but we can only *actually* collect data from this

Sample

1million comments from 2020

### Best sampling practices:

- Always think about what your population is
- Collect data from a sample that is representative of your population
- If you have no choice but to work with a dataset that is not collected randomly and is biased, be careful not to generalize your results to the entire population



You'd want to be sure you sample randomly across *all* YouTube comments, making sure not to get more comments from one genre over another, or one location over another, etc.

### Examples of bad sampling:

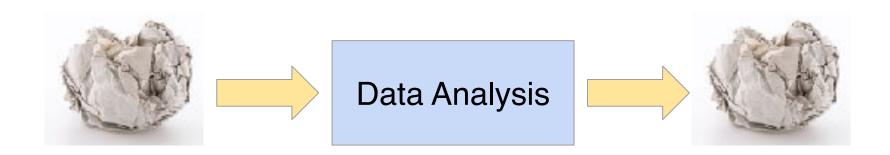
- Surveying subscribers of a gun-related magazine for research on Americans' attitudes toward owning guns
- Randomly sampling Facebook users for what TV shows people like



To understand *all* YouTube comments, you wouldn't just want to sample from one YouTube channel, or videos in a single language.

It's *always* worth spending time at the <u>beginning</u> of a project to determine whether or not the data you have are garbage. Be certain they are actually able to help you answer the question you're interested in.

### GIGO: Garbage In. Garbage Out.





For the survey data I collected from you all, which of the following best describes the population I could generalize findings back to.

**A** Undergraduates

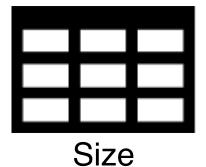
**B** Undergraduates in the US

C Undergraduates at UCSD

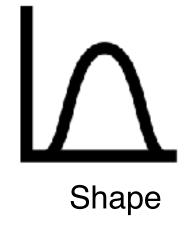
**D** Students aged 18-25

**E** UCSD COGS108 students

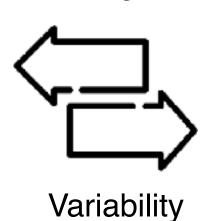
# Descriptive Analysis



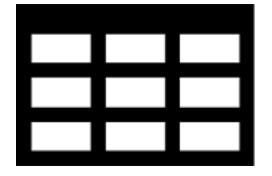












Size

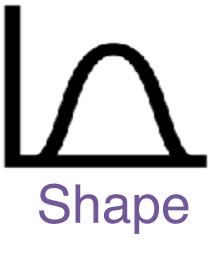
How many <u>observations</u> (rows) and <u>variables</u> (columns) you have is an important first step. You should always be aware of the size of your dataset





Missingness It's critical to know how many observations have missing data for variables of interest in your data. Knowing why their missing is also important.

**Descriptive** 



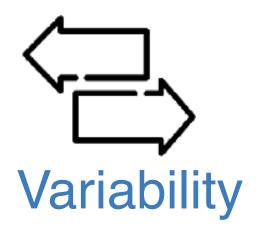
It's critical to know the distribution of the variables in your dataset. Certain statistical approaches can only be used with certain distributions.

**Descriptive** 

# Central Tendency

Knowing the mean, median, and/or mode can help you get an idea of what a typical value is for your variable(s) of interest





The central tendency tells you part of the story. The variability in the values in your observation helps fill in the rest.



# Which of the following is NOT something accomplished by a descriptive analysis?

A Describes typical values in your dataset

**B** Determines the size of your dataset

C Establishes causal relationships between variables

**D** Identifies missing data

**E** Determines how variable values in your dataset are

### Descriptive Statistics & Summary

"We must suppress some of the truth to communicate the truth... In short, the techniques of descriptive statistics are designed to match the salient features of the data set to human cognitive abilities."

-I.J. Good (1983)

# Descriptive Analyses are often included as "Table 1" in academic publications

Chanderinic	Ramibisumab Mandidy (94 - 900)	Sevacionnals Monthly 34 - 356	tanibicanab as Membel 94 = 290	threadourest as Membel (N = 300)
Age — cu. plu				
50-50 rt	2 (8.7)	1.0034	6 (3.0)	3 (0.7)
60-60 pt	10 (0.14)	20 (0.3)	au pave	34 pt.a)
20-29 pt	302 (01.8)	84 (29.4)	115 (1846)	100 (0.00)
80-85 yr	142 (41.2)	191 (52.4)	125 (42.7)	140 (973)
990 yr	32 (7.1)	23 (8.0)	26 (6.7)	18 (6.0)
Mean — yr	29.247.4	85.1-7.3	384a3.8	79.143.6
Section (Si)				
Fertale	383 (90.8)	380 868 96	185 (62.2)	184 (GLT)
Irlate	115 (20.2)	106 (37.1)	113 (27.5)	106 (18.7)
Face — no. (2001)				,
Wite	207 (38.7)	381 (58.5)	100 (20.0)	294 (55.0)
Other	4(0.3)	5(17)	2 (0.7)	60.00
History of reynousidal infanction — no. (%)	360139	40 (14.0)	80 (88.0)	58 (17.E)
History of strain — na. (95)	16.067)	19 (0.3)	22 (3.4)	18 (9.5)
History of purplent belief is situal; — no. (%)	12 (40)	25 (8.7)	12 (4.0)	15 (6.5)
Blood prossure — mm Hg				
Syraic	114418	115-19	126607	136e37
Distrole	75a16	35-10	7648	75a30
Visual usually some and Sheller equivalent.				
65-85 letters, 25/65-40 ea. (16)	311 (96.9)	94 (00.00	116 (19.0)	130 (14.3)
53-67 lations, 28,90-80 — so, (90)	25 (32.5)	118 (41.5)	100 (16.2)	115 (29.7)
36-52 lates to 28/308-190 no. (%)	67 (23.5)	53 (38.5)	58 (19.5)	51 (19.3)
23-37 larsen, 28,908-320 eo. (%)	25 (0.3)	21, (7.3)	16 (5.4)	20 (0.7)
Missan poorte	901a14.9	68.2619.7	61.5a21.2	60.4a11.4
lotal # idensis at fines — and	458,1184	465,106	4884198	461.175
Resinal thickness plan subfeveal-fluid thickness at fevea — pm	251,4123	254,123	3474112	3534315
Forest senter involvement — 60. (90)				
Choroidal neovasoularization	176 (58.5)	100 (51.5)	376 (59.2)	110 (GLG)
Rule	85 (28.2)	81, (28.5)	77 (23.0)	72 (24.0)
Hemorrhage	22 (8.6)	24 (8.4)	24 (8.1)	25 (8.3)
Other	15 (6.0)	20 (7.0)	15 (5.0)	18 (0.0)
No choroidal recovarcularization or not possible to griefs	2 (0.7)	8 (2.0)	6 (2.0)	3 (0.7)

<sup>4</sup> Rus-minus values are means ±50.

### **Descriptive**

If Race was self-reported.

That the bear in the fees includes the netral, substantified, characterization, and retrait pages of special elevation.

### **Descriptive**

Table 1. Baseline Characteristics of the Patients.*							
Characteristic		Ranibizumab Monthly (N = 301)	Bevacizumab Monthly (N = 286)	Ranibizumab as Needed (N = 258)	Bevacizumab as Needed (N = 300)		
Age — ro. (%)							
50-59 yr		2 (0.7)	1 (0.3)	6 (2.0)	2 (0.7)		
60-69 yr		33 (11.0)	28 (9.1)	31 (10.4)	34 (11.3)		
70-79 yr	Shape	102 (33.9)	84 (29.4)	115 (38.6)	103 (34.3)		
80-89 yr	Chape	142 (47.2)	150 (52.4)	126 (42.3)	142 (47.3)		
≥90 yr		22 (7.3)	23 (8.0)	20 (6.7)	19 (6.3)		
Mean — yr	Central	79.2: 7.4	<b>vari</b>	abilit	79.3±7.6		
Sez — no. (%)	Central				7		
Female		183 (60.8)	180 (62.9)	185 [62.1]	184 (61.3)		
Male	tendency	118 (39.2)	106 (37.1)	113 (37.9)	116 (38.7)		
Race — nc. (%)ĵ	terraction						
White		297 (98.7)	281 (98.3)	296 (99.3)	294 (98.0)		
Other		4 (1.3)	5 (1.7)	2 (0.7)	6 (2.0)		

<sup>\*</sup> Plus-minus values are means ±\$D.

### Size

Zooming in on this we see variables stratified by Age, Sex, and Race

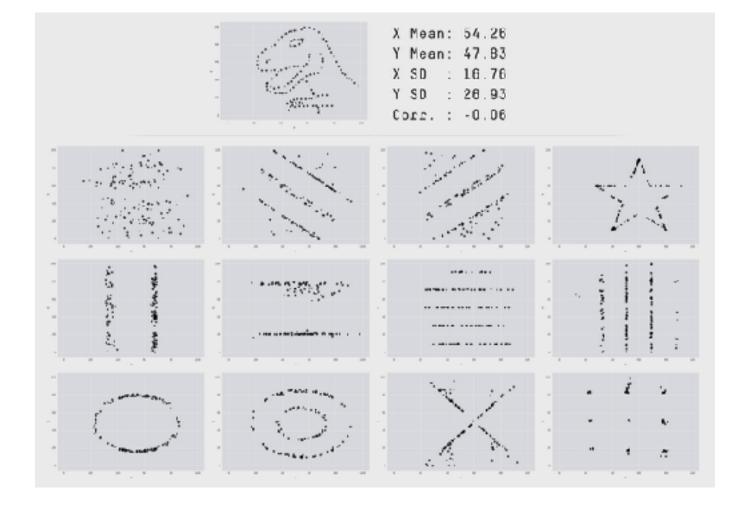
<sup>†</sup> Race was self-reported.

<sup>\$</sup> Total thickness at the forea includes the retina, subretinal fluid, choroidal neovascularization, and retinal pigment epithelial elevation.

**Descriptive** 

Descriptive Statistics & Summary

Calculating descriptive statistics, understanding what they tell you about your data, and reporting them are critical steps in every analysis.



**Exploratory**: The goal is to find unknown relationships between the variables you have measured in your data set. Exploratory analysis is open ended and designed to verify expected or find unexpected relationships between measurements.



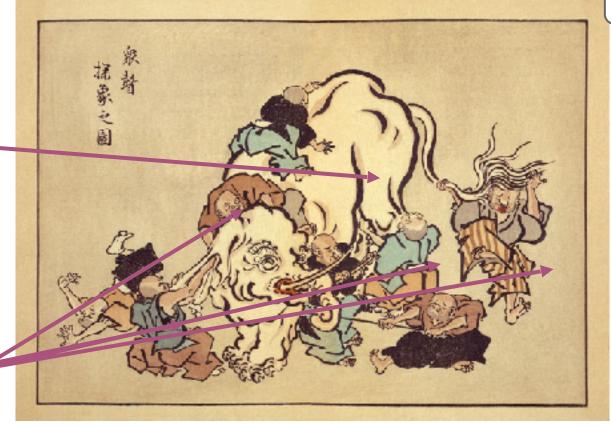
Exploratory Data Analysis (EDA) detective work answering the question: "What can the data tell us?"

### Why EDA?

- Understand data properties
- Discover Patterns
- Generate & Frame Hypothesis
- Suggest modeling strategies
- Check assumptions (sanity checks)
- Communicate results (present the data)

.....and if you don't, you'll regret it

The dataset



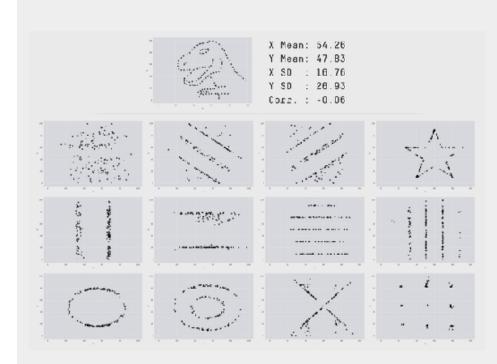
You

The general principles of exploratory analysis:

- Look for missing values
- Look for outlier values
- Calculate numerical summaries
- Generate plots to explore relationships
- Use tables to explore relationships
- If necessary, transform variables

### Start raw

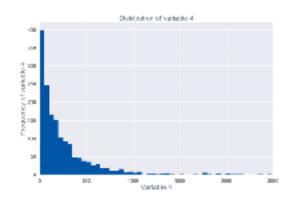
- Examine raw data in the most direct way you can reasonably do so
- View a random sample of the data
- Plots, especially subsets of variables and dimensionality reduction
- Helpful for seeing weirdness, missingness, outliers, min/max/typical values



#### EDA Approaches to "Get a Feel for the Data"

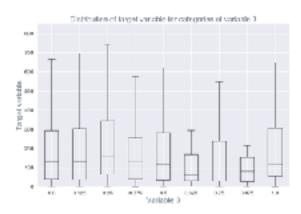
Understanding the relationship between variables in your dataset







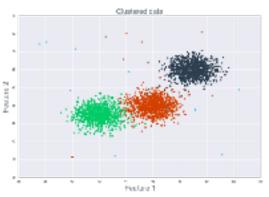
understanding a single variable i.e.: histogram, densityplot, barplot



### **Bivariate**

understanding relationship between 2 variables

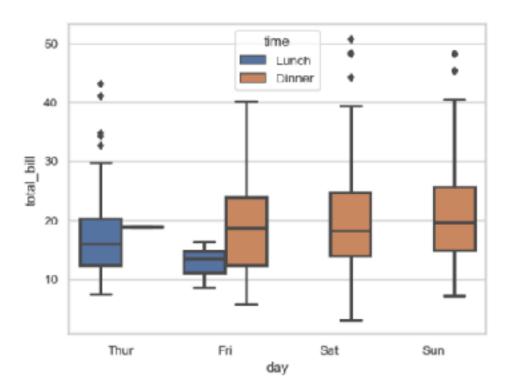
i.e.: boxplot, scatterplot, grouped barplot, boxplot

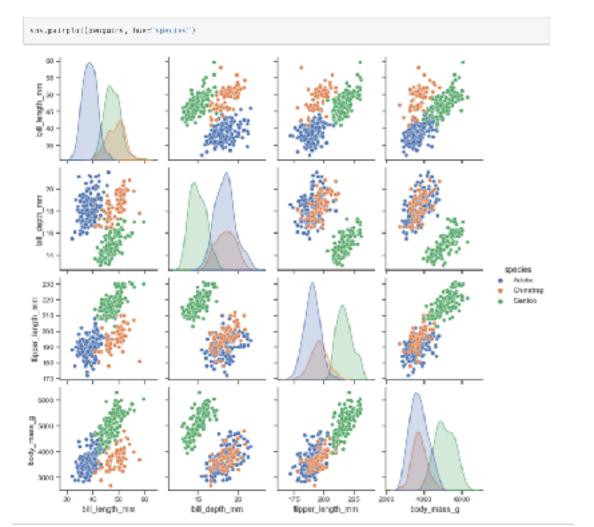


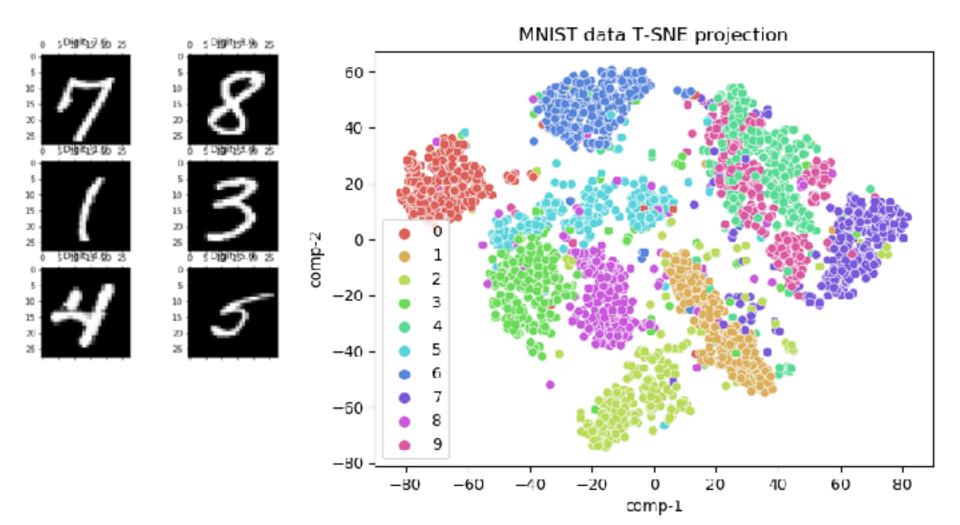
### **Dimensionality Reduction**

projecting high-D data into a lower-D space

i.e.: PCA, ICA, Clustering







- 1. Hyperparameters really matter
- 2. Cluster sizes in a UMAP plot mean nothing
- 3. Distances between clusters might not mean anything
- 4. Random noise doesn't always look random.
- 5. You may need more than one plot

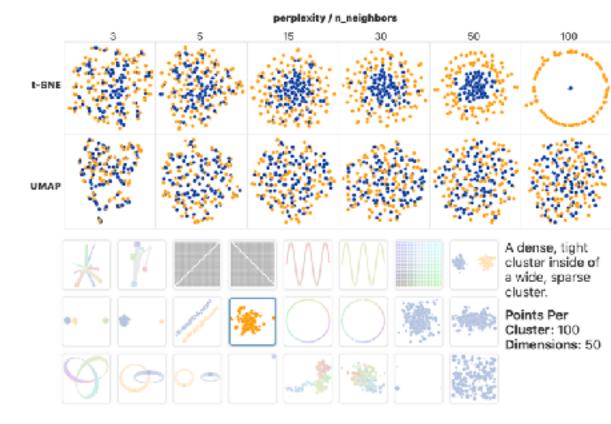


Figure 7: Comparison between UMAP and t-SNE projecting various toy datasets.