#### Statistical Methods in AI (CSE/ECE 471)

Lecture-2: ML Workflow, Data Representations, Supervised Learning, Intro to Classification

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#### Announcements

- IMPORTANT: All assignments/projects will need to be in Python.
- Tutorial on Python, Pandas, Jupyter notebook this Saturday. Bring your laptops.
- Ask questions.

#### Announcements

- Assignments Questions involving equations/ mathematical derivation
  - Write up in latex [overleaf.com] → submit
  - Write neatly on paper → scan as photo/pdf
     [camscanner] → submit
- TA office hours, locations will be announced shortly.

### Lecture Outline

- ML Workflow
- Intro to Supervised Learning
  - Taxonomy
  - Data Representations
  - Models

### Machine Learning



Algorithmic methods that use data to improve their knowledge of a task

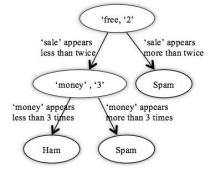
Task: Detect spam email





Data: Labelled emails (in inboxes of other users as well!)

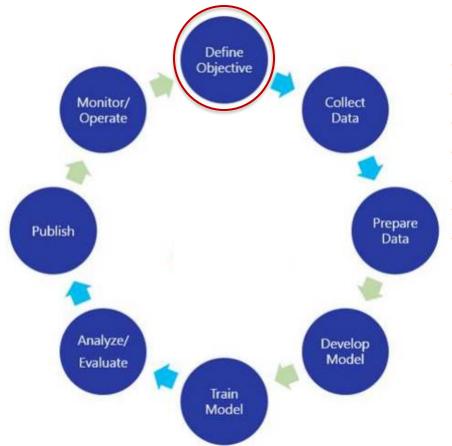
Knowledge:



Improve → 85% reduction of spam emails in Inbox over 3 months

Algorithmic method: Decision Tree

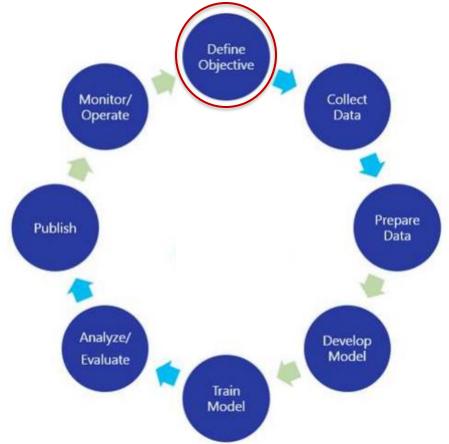
### Workflow of a Machine Learning Problem



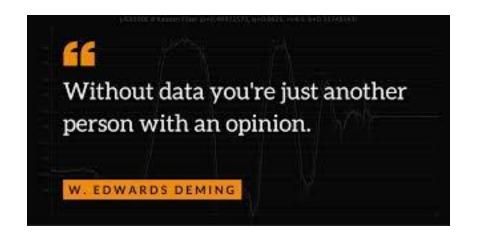
- Detect spam email
- Predict value of a stock
- Predict effect of advertising on sales
- Drive car 'safely' without human intervention
- Translate text from one language to another
- Sentiment Analysis

- ...

## Workflow of a Machine Learning Problem



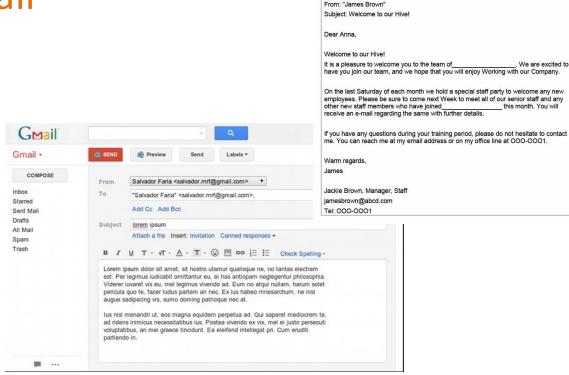
## No Data, no ML!





- Detect spam email





**Business Email Sample** 

. We are excited to

this month. You will

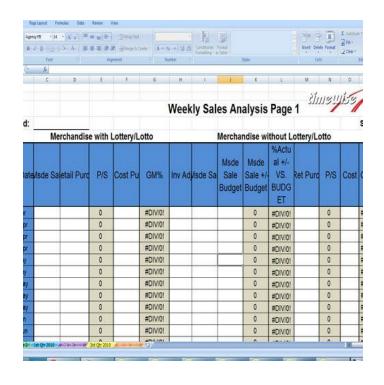
To: "Anna Jones" <annajones@buzzle.com.>

CC: All Staff

- Predict value of a stock



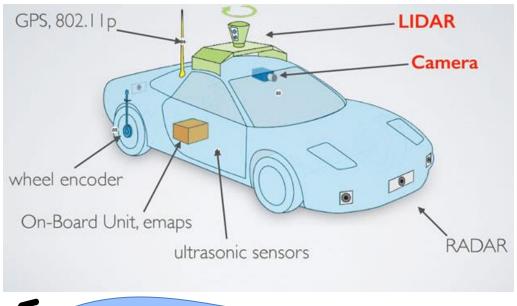
- Predict effect of advertising on sales

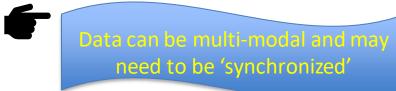






- Drive car safely without human intervention





- Translate text from one language to another





### Raw data

Not all of it relevant

```
The state of the s
      Received: "by luna.mailgun.net with HTTP; Fri, 26 Feb 2016 20:12:03 +0000",
      stripped-signature: "",
      Message-Id: "<20160226201203.54979.26875@mailgun.com>",
      from: "Sample Email <me@mailgun.com>",
      sender: "me@mailgun.com",
      recipients: "anton@mailgunhq.com",
      Subject: "Test Message",
      Content-Transfer-Encoding: "7bit",
      attachments: [ ],
      To: "anton@mailgunhq.com",
      stripped-html: "Testing some Mailgun awesomness!",
      content-id-map: { },
      stripped-text: "Testing some Mailgun awesomness!",
      From: "Sample Email <me@mailgun.com>",
+ message-headers: [...],
      Mime-Version: "1.0",
      Content-Type: "text/plain; charset="ascii"",
      body-plain: "Testing some Mailgun awesomness!",
      subject: "Test Message"
```

#### Raw data

- Often not directly usable
  - Filter (needed data)
  - Transform (to numerical data)

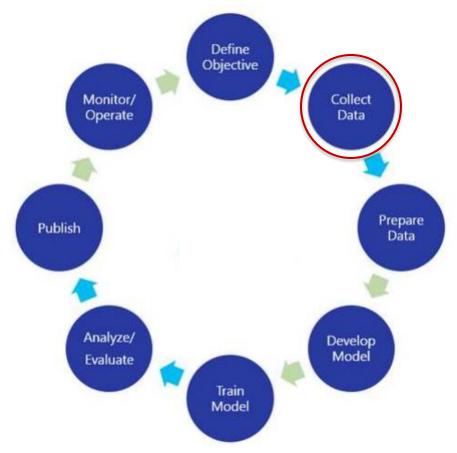
```
C https://api.mailgun.net/v2/domains/mailgun.com/messages/WyJlMTFiZ
 Received: "by luna.mailgun.net with HTTP; Fri, 26 Feb 2016 20:12:03 +0000",
 stripped-signature: "".
 Message-Id: "<20160226201203.54979.26875@mailgun.com>",
 from: "Sample Email <me@mailgun.com>",
 sender: "me@mailgun.com",
 recipients: "anton@mailgunhq.com",
 Subject: "Test Message",
 Content-Transfer-Encoding: "7bit".
 attachments: [ ],
 To: "anton@mailgunhq.com",
 stripped-html: "Testing some Mailgun awesomness!",
 content-id-map: { },
 stripped-text: "Testing some Mailgun awesomness!",
 From: "Sample Email <me@mailgun.com>",
+ message-headers: [...],
 Mime-Version: "1.0",
 Content-Type: "text/plain: charset="ascii"".
 body-plain: "Testing some Mailgun awesomness!",
 subject: "Test Message"
```

### Raw data

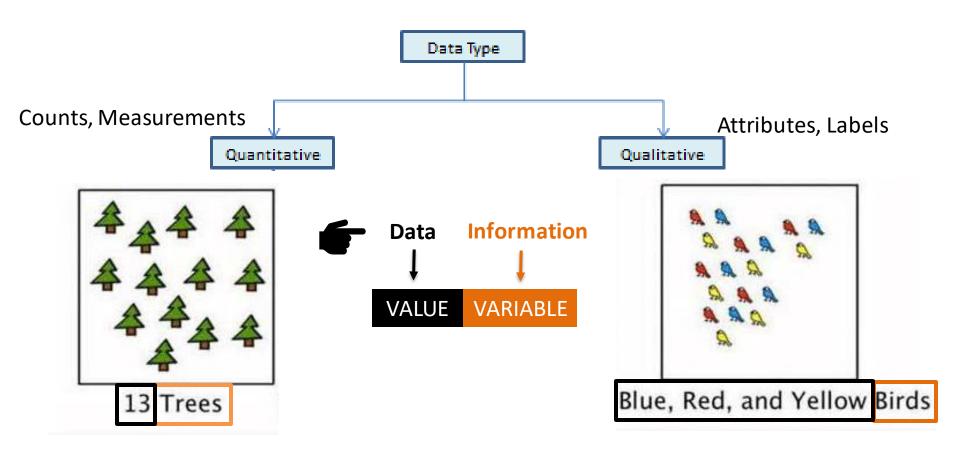
- May be too much in quantity
  - Limitations on system end (compute, storage)

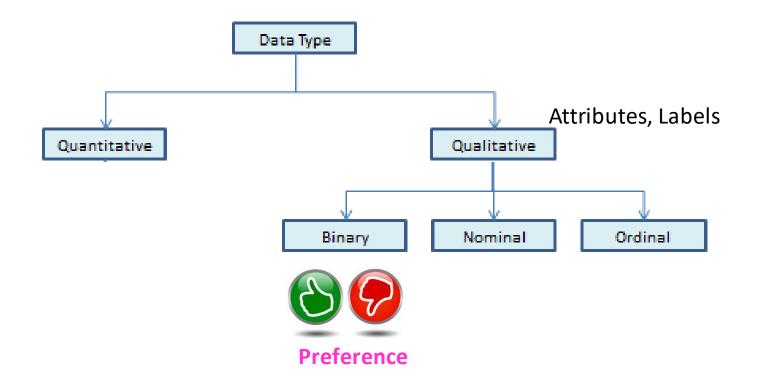


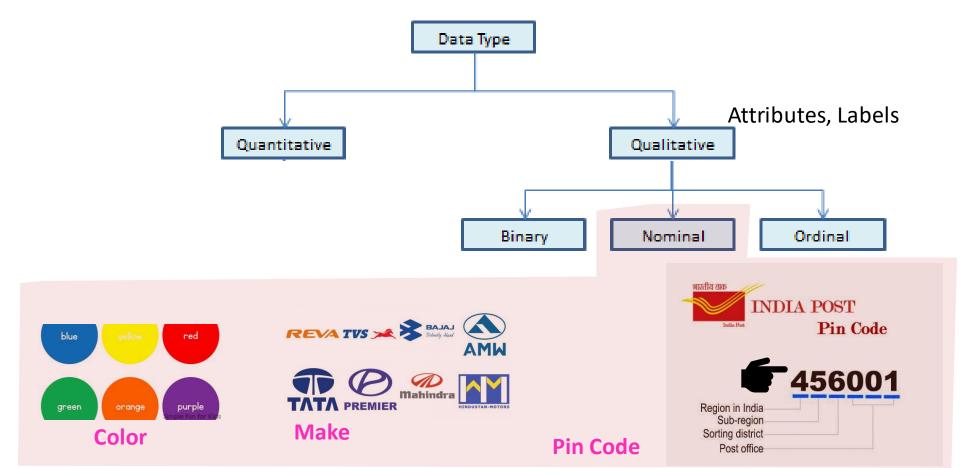
## Workflow of a Machine Learning Problem

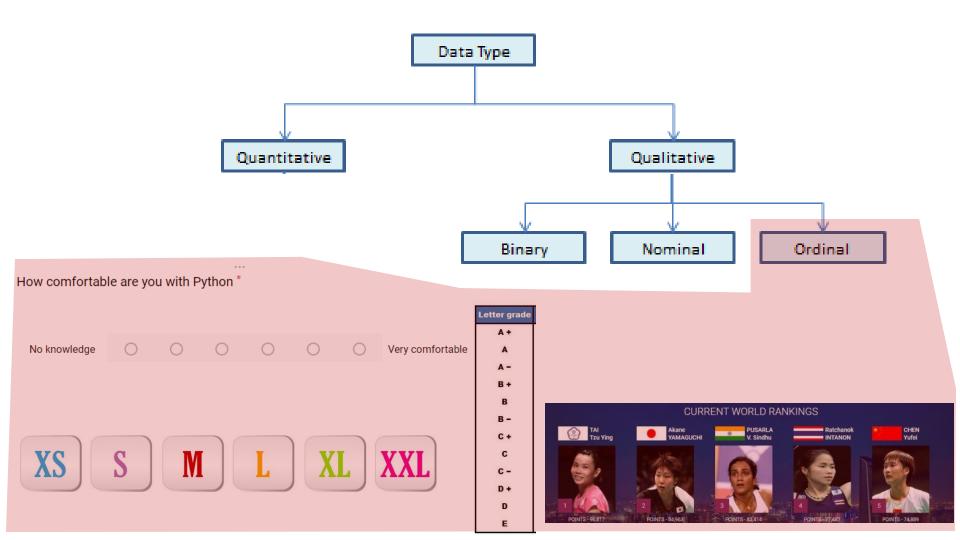


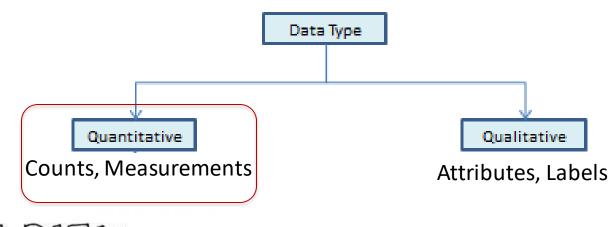
# Taxonomy of data variables



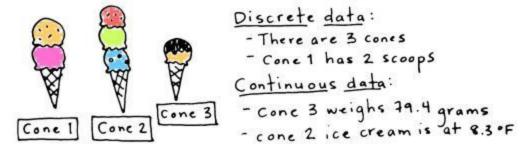


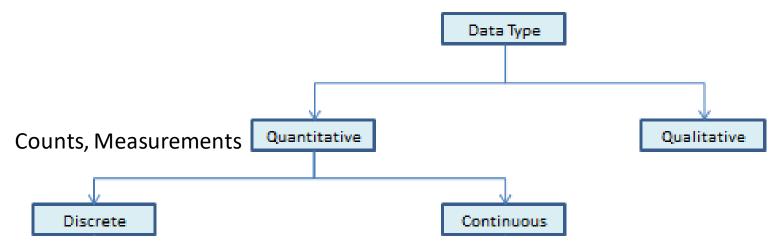




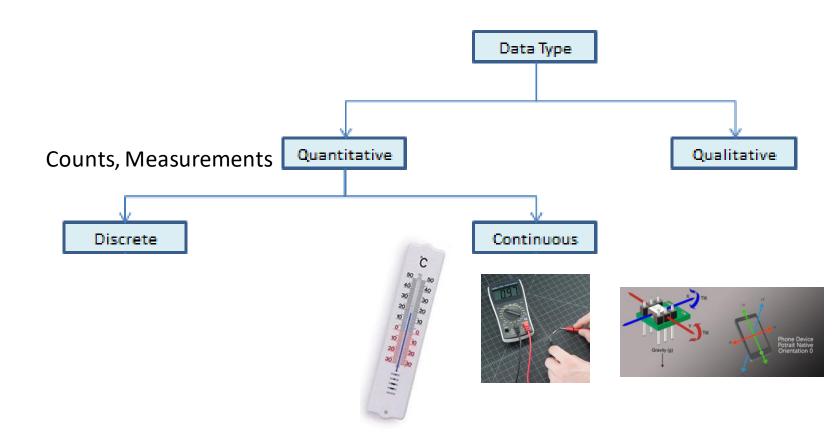


#### QUANTITATIVE DATA:

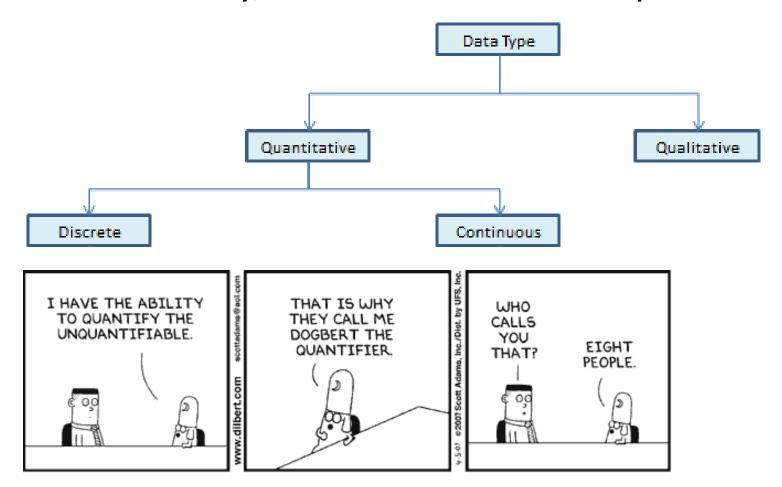




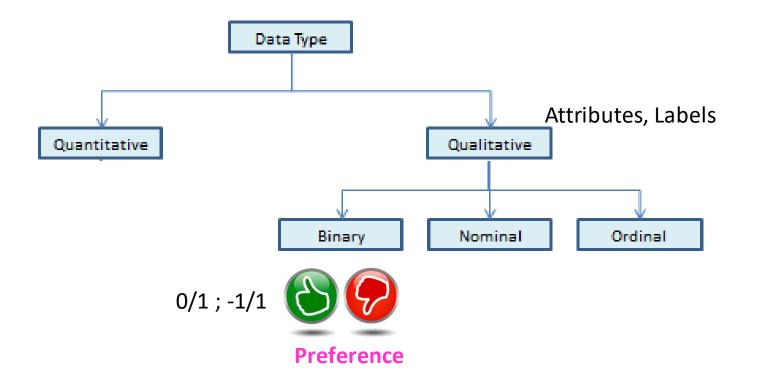
- # of CPU cores
- # of courses taken in a semester
- # of times word 'sale' appears in a doc



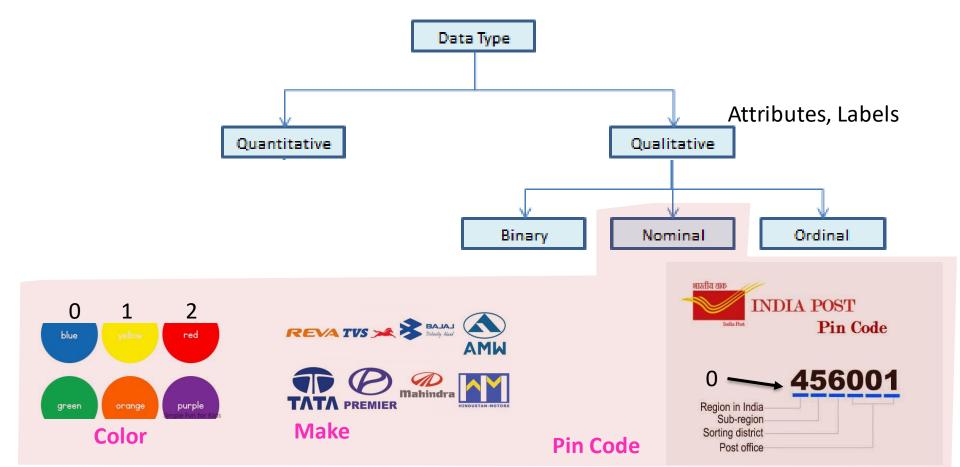
#### Ultimately, all data needs to be quantitative

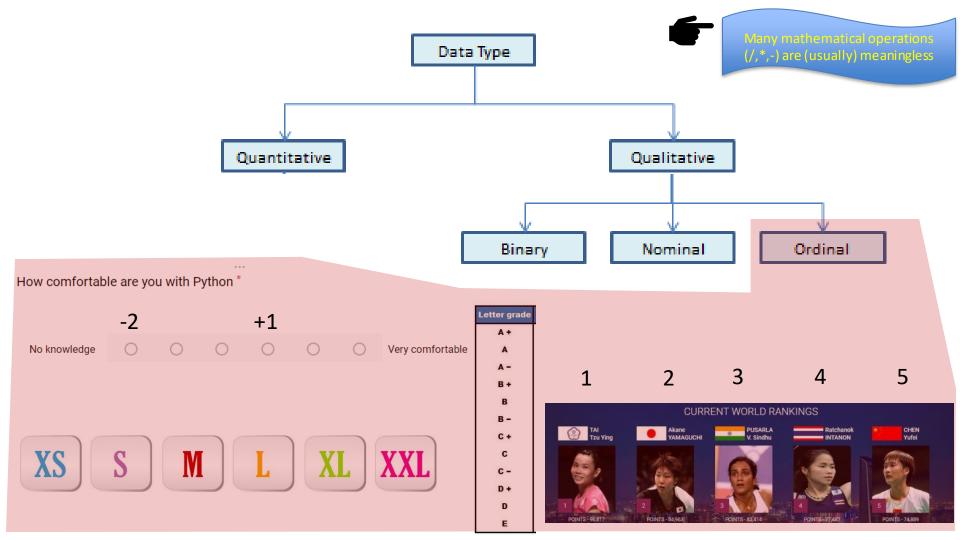


#### Taxonomy of data: Qualitative → Quantitative



#### Taxonomy of data: Qualitative → Quantitative





## Example: Contact Lenses dataset

No patient id

Age is not a number!

Age	Spectacle prescription	Astigmatism	Tear production rate	Recommended lenses	
Young	Myope	No	Reduced	None	
Young	Myope	No	Normal	Soft	
Young	Myope	Yes	Reduced	None	
Young	Myope	Yes	Normal	Hard	
Young	Hypermetrope	No	Reduced	None	
Young	Hypermetrope	No	Normal	Soft	
Young	Hypermetrope	Yes	Reduced	None	
Young	Hypermetrope	Yes	Normal	hard	
Pre-presbyopic	Myope	No	Reduced	None	
Pre-presbyopic	Myope	No	Normal	Soft	
Pre-presbyopic	Myope	Yes	Reduced	None	
Pre-presbyopic	Myope	Yes	Normal	Hard	
Pre-presbyopic	Hypermetrope	No	Reduced	None	
Pre-presbyopic	Hypermetrope	No	Normal	Soft	
Pre-presbyopic	Hypermetrope	Yes	Reduced	None	
Pre-presbyopic	Hypermetrope	Yes	Normal	None	
Presbyopic	Myope	No	Reduced	None	
Presbyopic	Myope	No	Normal	None	
Presbyopic	Myope	Yes	Reduced	None	
Presbyopic	Myope	Yes	Normal	Hard	
Presbyopic	Hypermetrope	No	Reduced	None	
Presbyopic	Hypermetrope	No	Normal	Soft	
Presbyopic	Hypermetrope	Yes	Reduced	None	
Presbyopic	Hypermetrope	Yes	Normal	None	

# Example: PlayTennis dataset

Outlook	Temperature	Humidity	Windy	Play
Sunny	Hot	High	False	No
Sunny	Hot	High	True	No
Overcast	Hot	High	False	Yes
Rainy	Mild	Normal	False	Yes
	***	***	***	***

Outlook	Temperature	Humidity	Windy	Play
Sunny	85	85	False	No
Sunny	80	90	True	No
Overcast	83	86	False	Yes
Rainy	75	80	False	Yes
***		***	***	***

# Sometimes data can be missing

Outlook	Temperature	Humidity	Windy	Play
Sunny	85	85	False	No
Sunny	80		True	No
Overcast	83	86	False	Yes
Rainy	75	80	False	Yes
***	***	444	***	(800)

→ Unknown or unrecorded

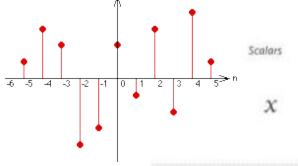
### ... or incorrect

		DBAName	AKAName	Address	City	State	Zip		
	t1	John Veliotis Sr.	Johnnyo's	3465 S Morgan ST	Chicago	IL	60608	X	Conflicts
	t2	John Veliotis Sr.	Johnnyo's	3465 S Morgan ST	Chicago	L	60609		Commoto
	t3	John Veliotis Sr.	Johnnyo's	3465 S Morgan ST	Chicago	IL	60609		
	t4	Johnnyo's	Johnnyo's	3465 S Morgan ST	Cicago	IL	60608		
_	Does not obey			1	Cor	nflict			
→ data distribution									

# Data imputation

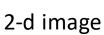
Approaches that aim to estimate missing data

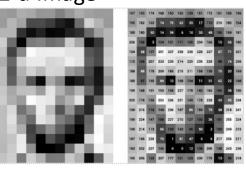
# **Data Representations**

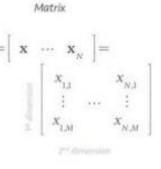


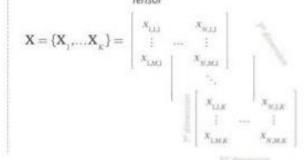
$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_N \end{bmatrix}$$

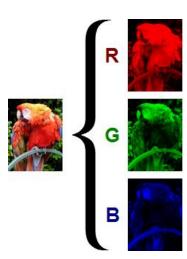






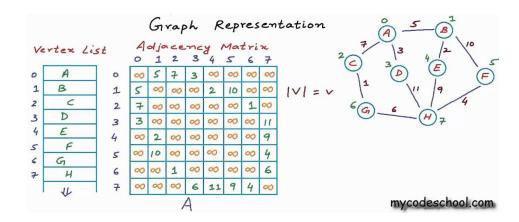






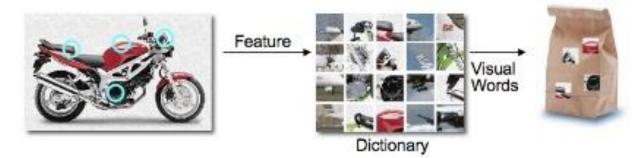
# Data Representations





# Feature Extraction (FE)

Def: Feature Extraction (FE) is any algorithm that transformation raw data into features that can be used as an input for a learning algorithm.

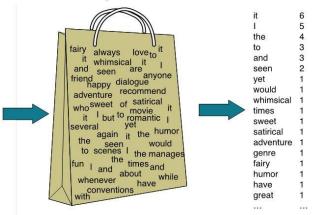


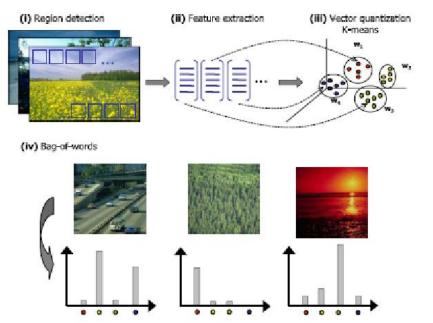
- Examples
- OConstruct bag-of-words vector from an email
- ORemove stopwords in a sentence
- OApply PCA projection to high-dimensional data

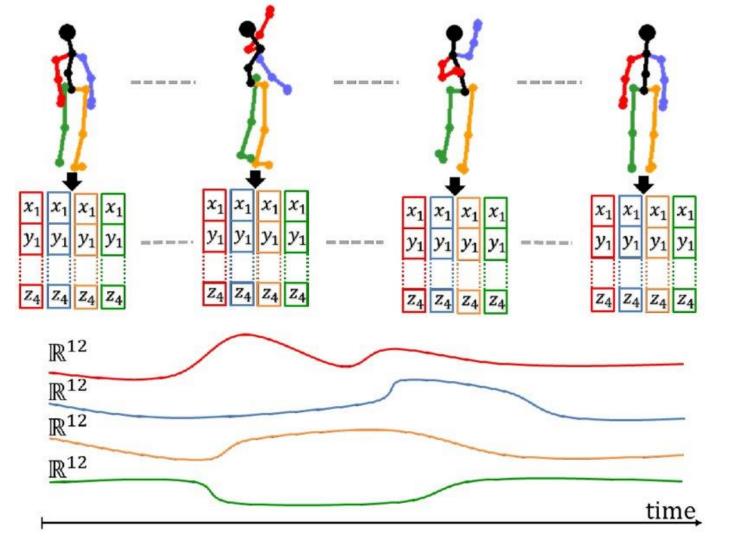
#### The Bag of Words Representation

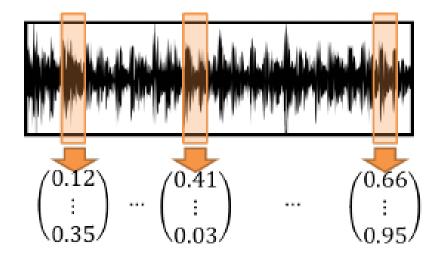
I love this movie! It's sweet, but with satirical humor. The dialogue is great and the adventure scenes are fun... It manages to be whimsical and romantic while laughing at the conventions of the fairy tale genre. I would recommend it to just about anyone. I've seen it several times, and I'm always happy to see it again whenever I have a friend who hasn't seen it yet!

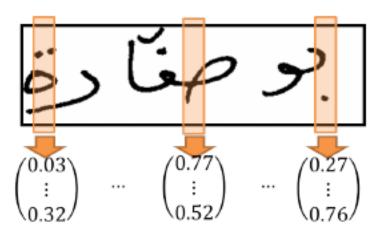
15



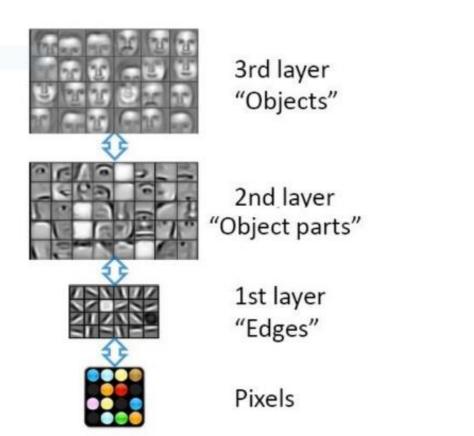


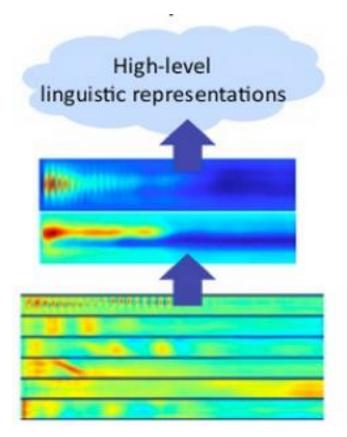




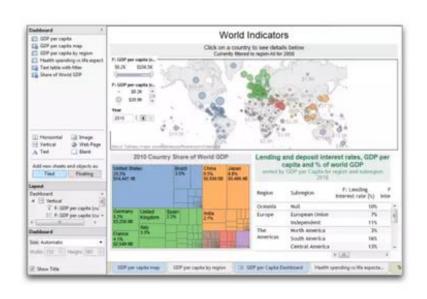


#### Feature-based, Hierarchical Data Representations

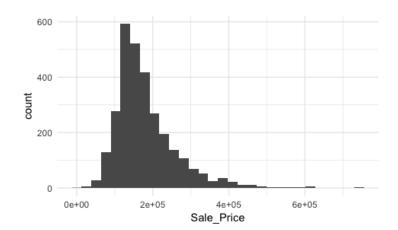




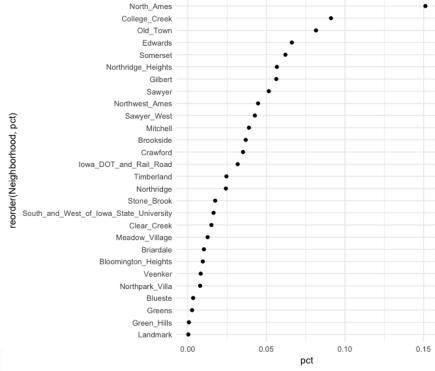
# Gazing at Data: Data visualization data exploration data presentation

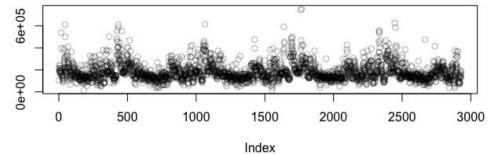


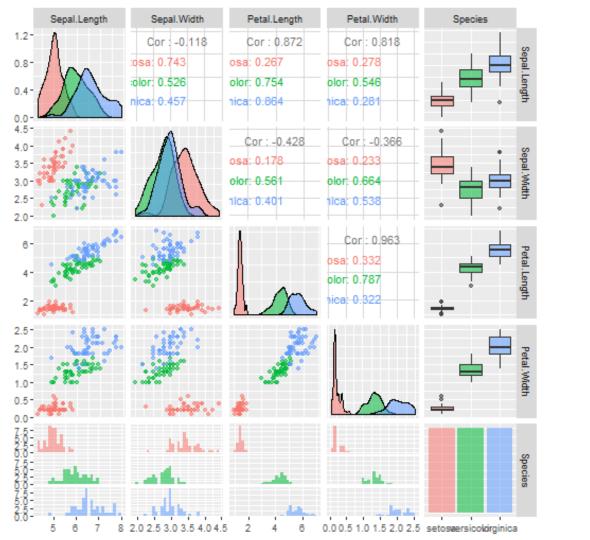




ames\$Sale\_Price

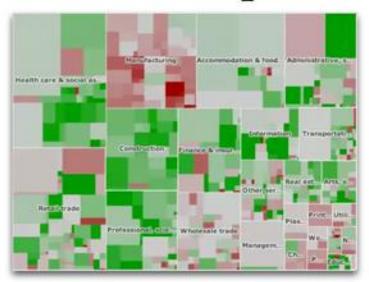






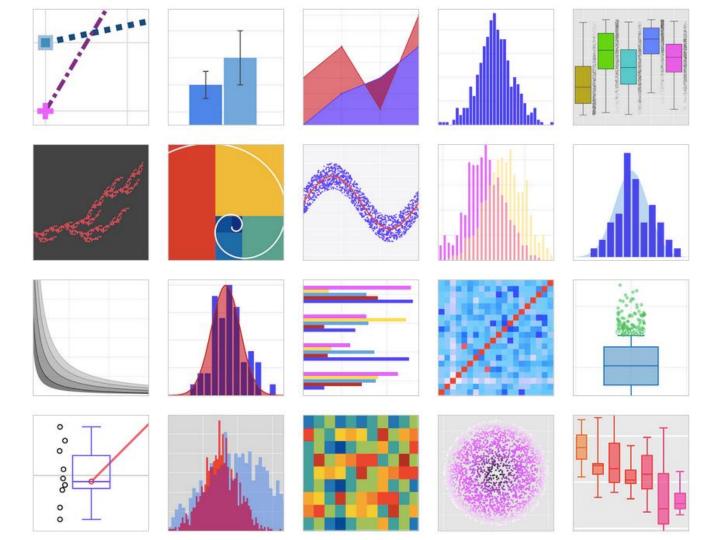
## Data visualization

## treemap

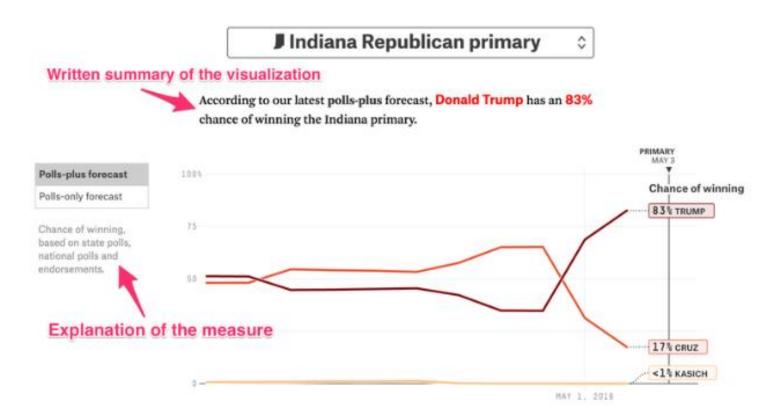


### leaderboard





## Data Exploration and Visualization



## Data Exploration and Visualization

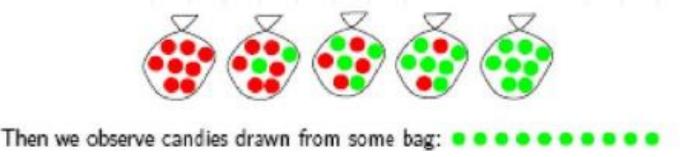
.. To be covered in next week's tutorial

In good information visualization, there are no rules, no guidelines, no templates, no standard technologies, no stylebooks ... You must simply do whatever it takes.

—Edward Tufte

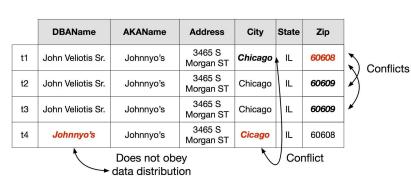
## Data – a probability-based perspective

The basis for Statistical Learning Theory



- Domain described by random variables (r.v.)
  - X = {apple, grape}
  - $b_i \in [1,5]$
- Data = Instantiation of some or all r.v.'s in the domain

## Data: a probabilistic perspective





#### **Proposed Cleaned Dataset**

	DBAName	Address	City	State	Zip
t1	John Veliotis Sr.	3465 S Morgan ST	Chicago	IL	60608
t2	John Veliotis Sr.	3465 S Morgan ST	Chicago	IL	60608
t3	John Veliotis Sr.	3465 S Morgan ST	Chicago	IL	60608
t4	John Veliotis Sr.	3465 S Morgan ST	Chicago	IL	60608

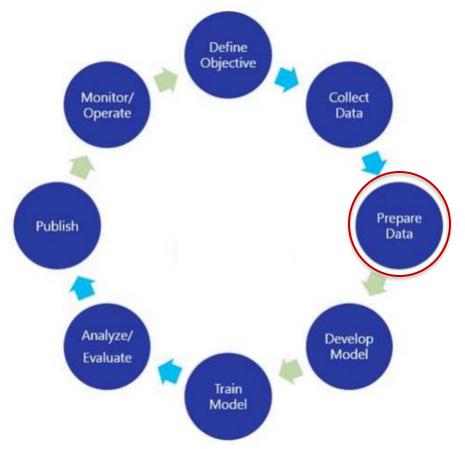
#### Marginal Distribution of Cell Assignments

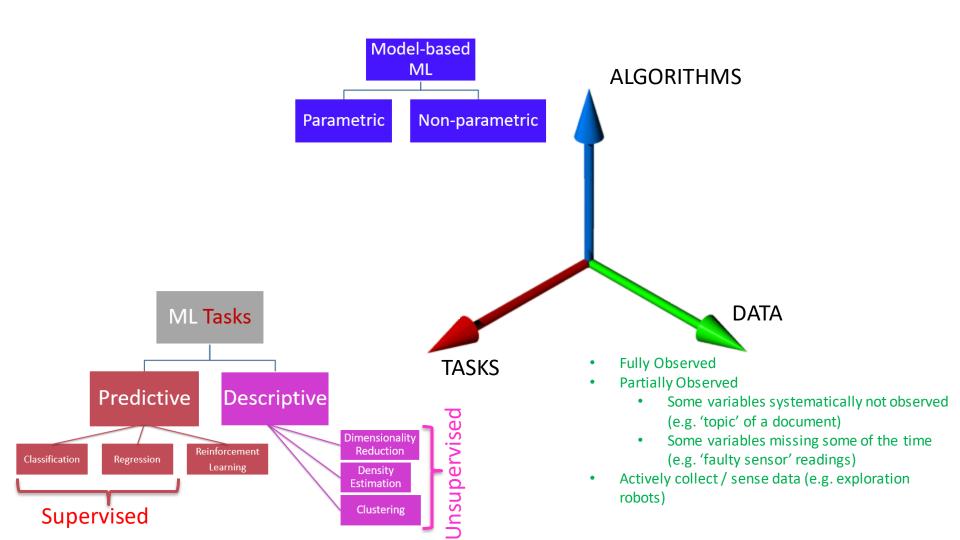
Cell	Possible Values	Probability
10 7:-	60608	0.84
t2.Zip	60609	0.16
t4.City	Chicago	0.95
	Cicago	0.05
HA DOANI	John Veliotis Sr.	0.99
t4.DBAName	Johnnyo's	0.01

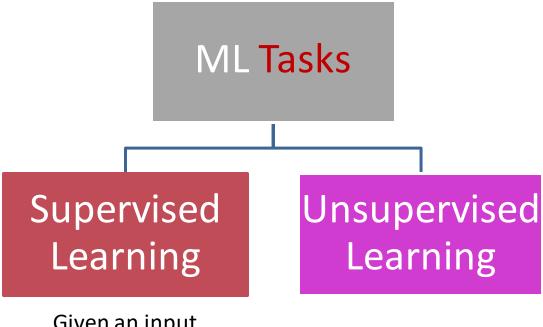
## Other important aspects of data

- Mode of collection
  - Passive ('sense')
  - Active ('explore, sense, repeat')

- Statistical assumptions on data
  - i.i.d (independent and identically distributed)
  - Online (e.g. time-series data)

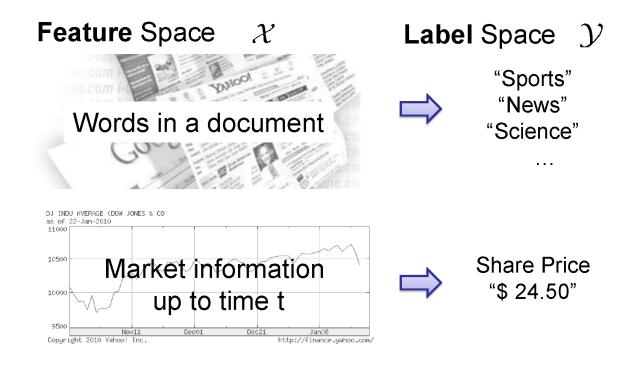






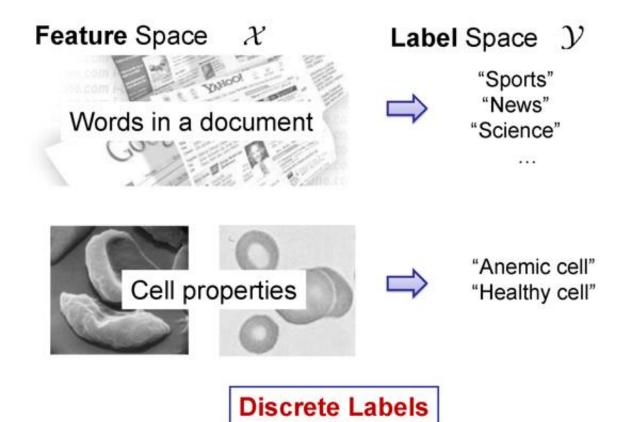
Given an input, estimate output

#### ML::Tasks → Predictive

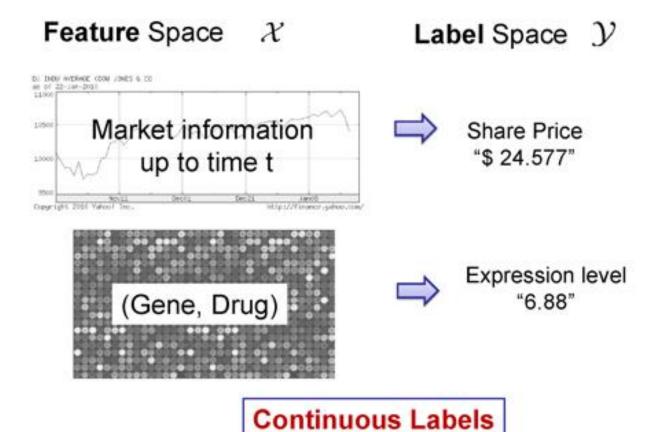


**Task:** Given  $X \in \mathcal{X}$ , predict  $Y \in \mathcal{Y}$ .

#### ML::Tasks → Predictive → Classification



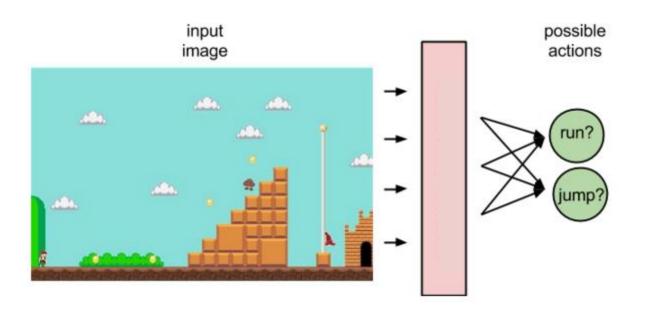
#### ML::Tasks → Predictive → Regression



#### ML::Tasks → Predictive → Reinforcement Learning

Feature Space  $\mathcal{X}$ 

**Label** Space  ${\mathcal Y}$ 

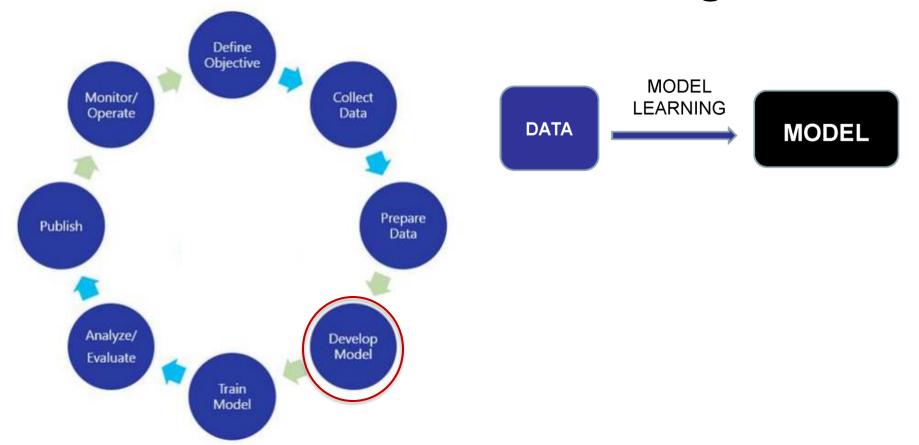


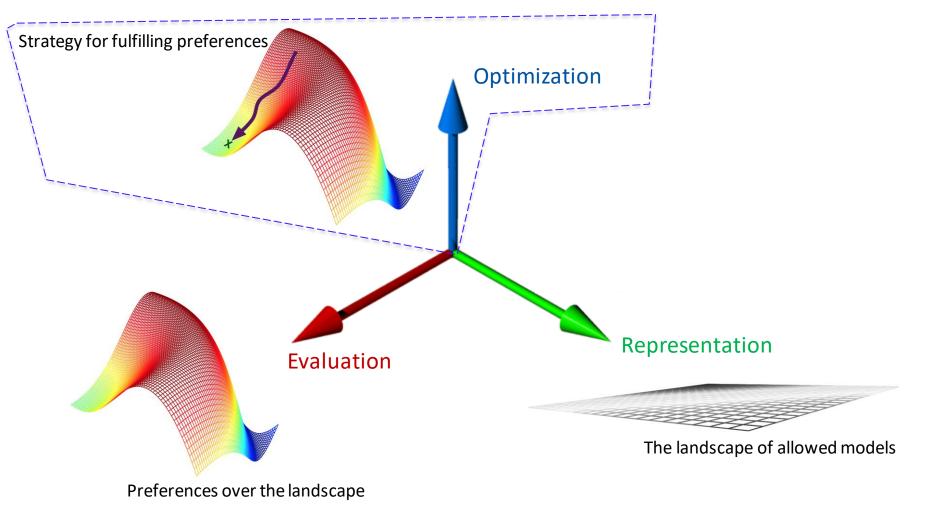
#### ML::Tasks → Predictive → Reinforcement Learning

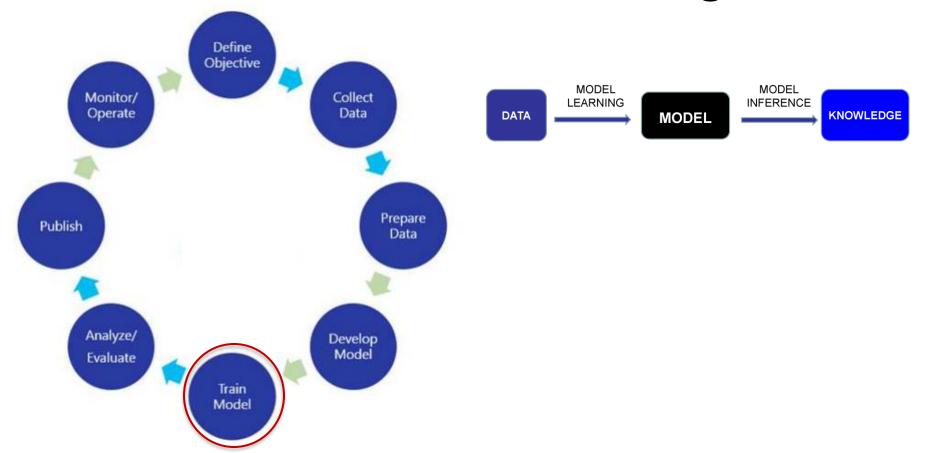
Feature Space X

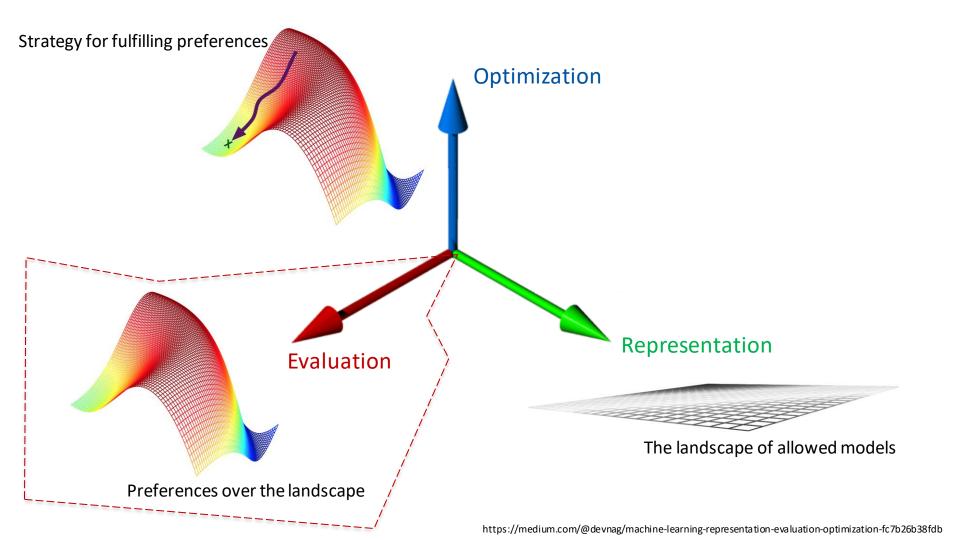
**Label** Space  ${\mathcal Y}$ 

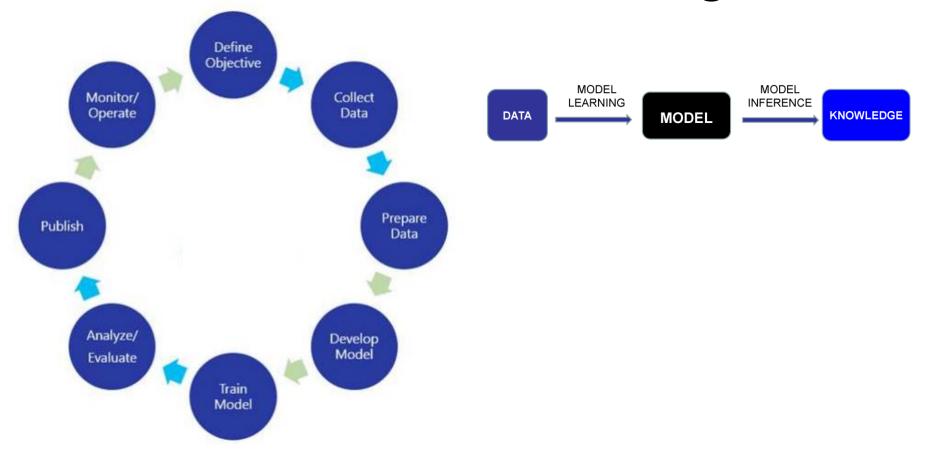


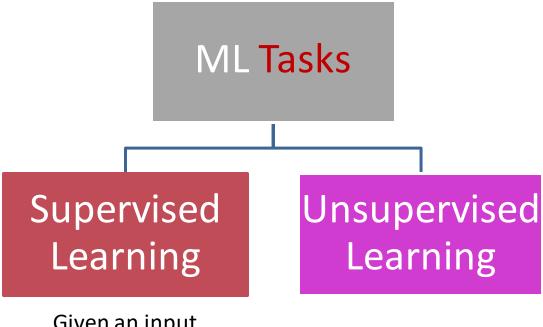












Given an input, estimate output

## Supervised Learning



#### Data

Space of inputs (or, predictors):  $\mathcal{X}$ 

 $\triangleright$  e.g.  $x \in \mathcal{X} \subset \{0, 1, \dots, 2^{16}\}^{64}$  is a string of pixel intensities in an  $8 \times 8$  image.

 $\triangleright$  e.g.  $x \in \mathcal{X} \subset \mathbb{R}^{33,000}$  is a set of gene expression levels.

$$x_1 = \begin{bmatrix} & & & \\ & & & \\ & & & \end{bmatrix}$$

$$x_1 =$$
  $x_2 =$  .

#### Space of outputs (or, responses): $\mathcal{Y}$

- $\triangleright$  e.g.  $y \in \mathcal{Y} = \{0,1\}$  is a binary label (1 = ``cat'')
- $\triangleright$  e.g.  $y \in \mathcal{Y} = [0, 200]$  is life expectancy

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A pair (x, y) is a *labeled* example.

 $\triangleright$  e.g. (x,y) is an example of an image with a label y=1, which stands for the presence of a face in the image x

#### Space of outputs (or, responses): $\mathcal{Y}$

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- $\triangleright$  e.g.  $y \in \mathcal{Y} = [0, 200]$  is life expectancy

A pair (x, y) is a *labeled* example.

 $\triangleright$  e.g. (x,y) is an example of an image with a label y=1, which stands for the presence of a face in the image x

Dataset (or training data): examples  $\{(x_1, y_1), \dots, (x_n, y_n)\}$ 

▶ e.g. a collection of images labeled according to the presence or absence of a face



Classification Regression Reinforcement Learning

We'll focus on these two



Classification

Regression

Reinforcement

Learning

