

Course Announcements

Major screwups on my part:

- Q6
- Weekly Project Survey (*optional*)

Due Friday (11:59 PM):

- D6
- A3

Machine learning

Jason G. Fleischer, Ph.D.

Asst. Teaching Professor

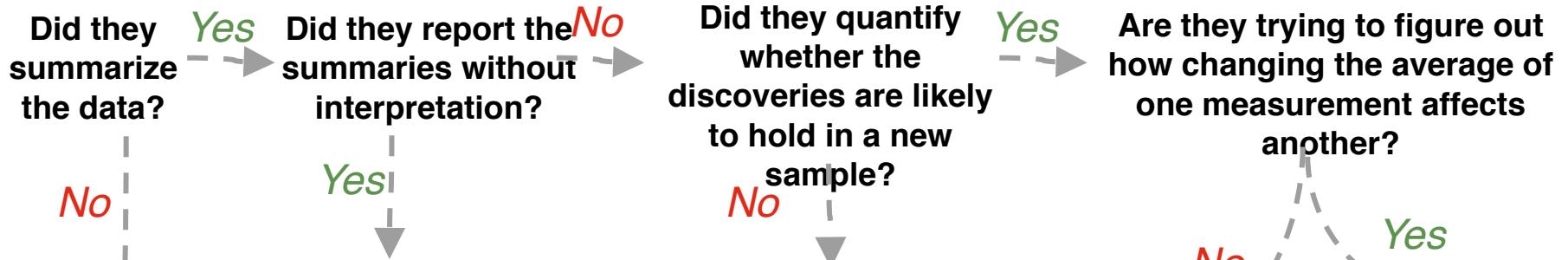
Department of Cognitive Science, UC San Diego

jfleischer@ucsd.edu



@jasongfleischer

<https://jgfleischer.com>



Predictive: apply machine learning techniques to data you have currently

Nonparametric Text Analysis



Are the data a corpus of text?

No

Yes

Exploratory

Are they trying to predict measurement(s) for individuals?

No

Yes

Causal

Inferential

Are the observations spatially related?

No

Yes

Geospatial Statistics

Predictive

Unsupervised Machine Learning

Did the computer decide the features of your model?

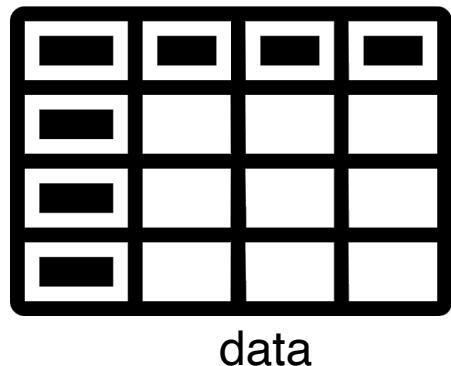
Supervised Machine Learning

- **Problem:** Detecting whether credit card charges are fraudulent.
- **Data science question:** Can we use the time of the charge, the location of the charge, and the price of the charge to predict whether that charge is fraudulent or not?
- **Type of analysis:** Predictive analysis



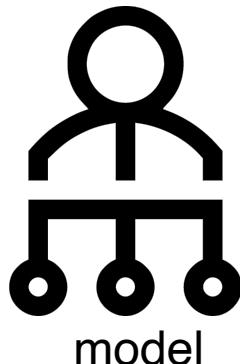
predictive analysis
uses data you have now
to make predictions in
the future

machine learning
approaches are used for
predictive analysis!



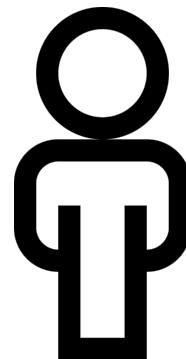
data

train →



model

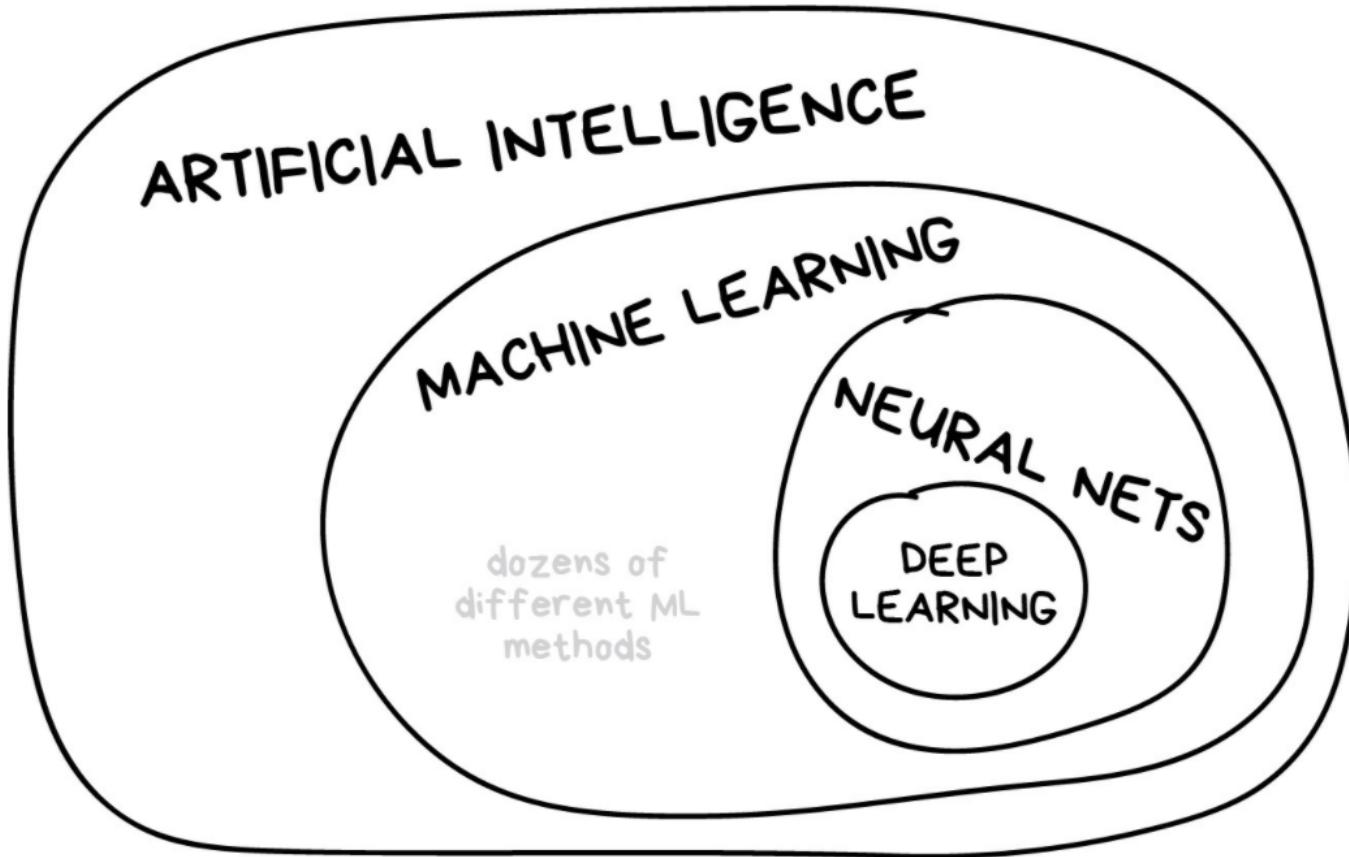
predict →



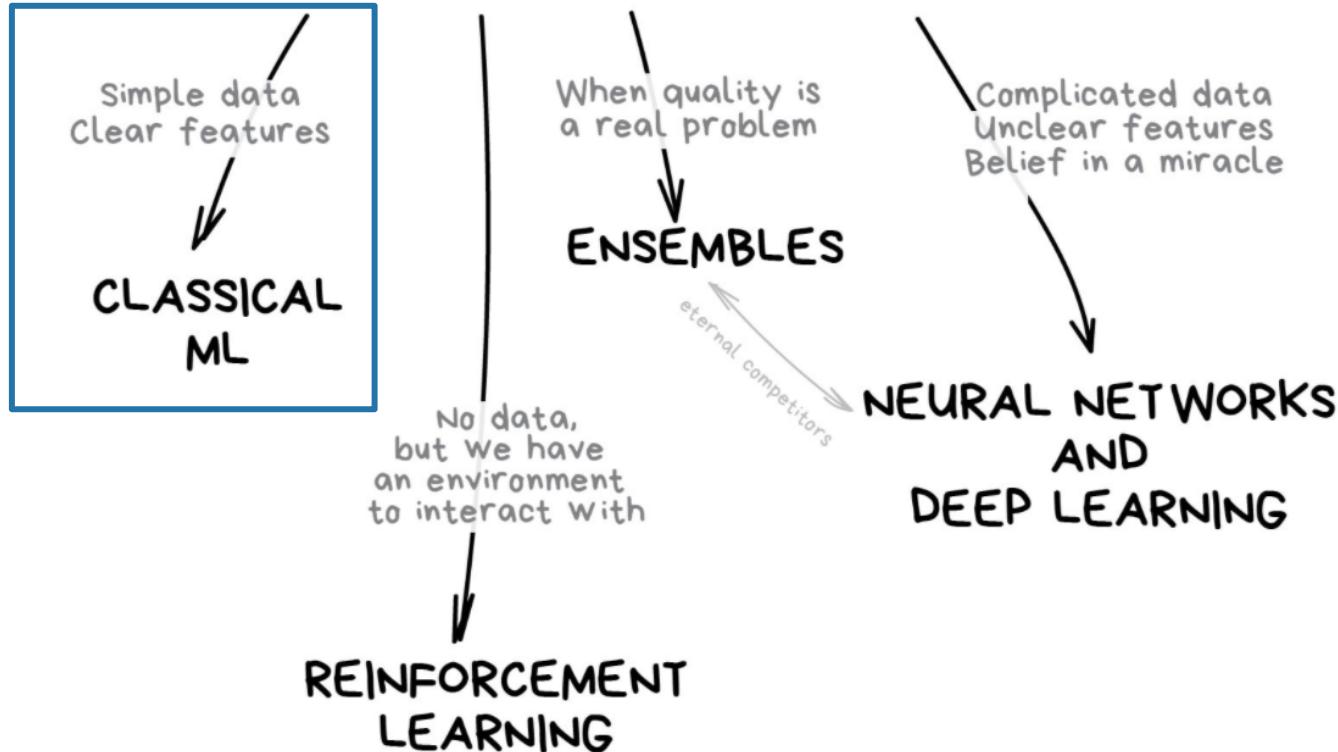
What is machine learning?

“Machine learning is the science of getting computers to act without being explicitly programmed”

- Andrew Ng, Stanford, ex-Google, chief scientist at Baidu, Coursera founder, Stanford Adjunct Faculty



THE MAIN TYPES OF MACHINE LEARNING





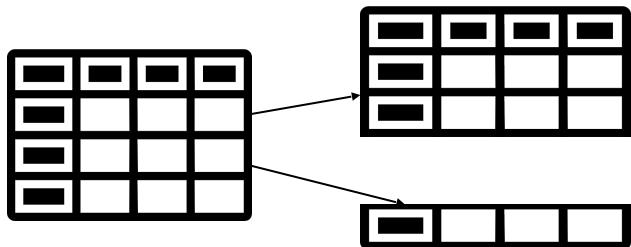
Prediction Questions

Which of these questions is most appropriate for machine learning?

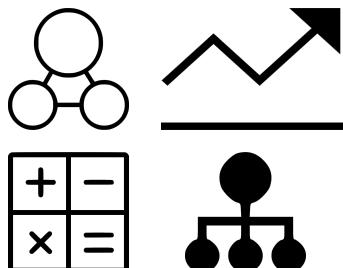
- A** How common is watching Sesame Street in the US?
- B** What is the effect of watching Sesame Street on children's brains?
- C** What is the relationship between early childhood educational programming and success in elementary school?
- D** Can we use information about one's early childhood to predict their success in elementary school?
- E** How does Sesame Street cause an increase in educational attainment?

Machine Learning Generalizations

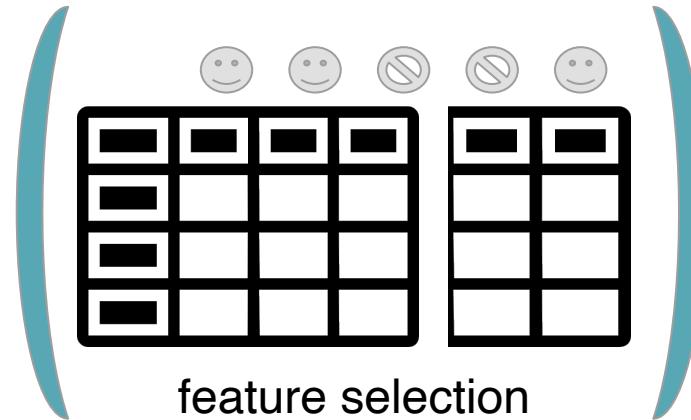
Basic Steps to Prediction



data
partitioning



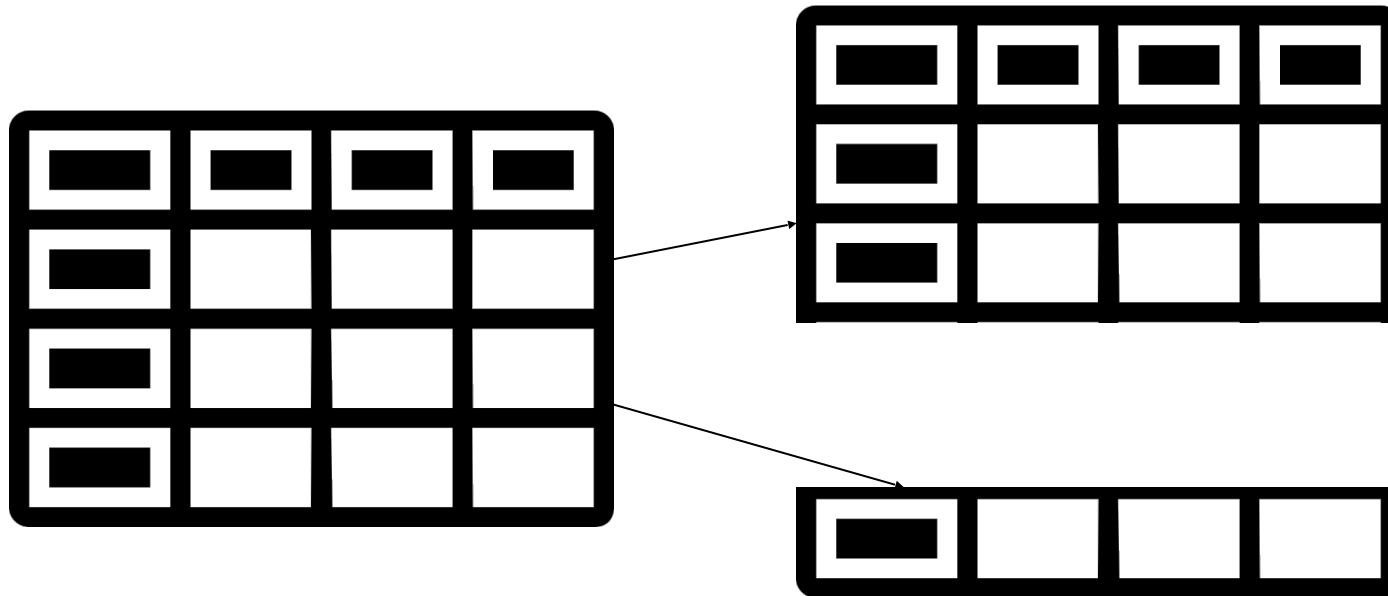
model selection



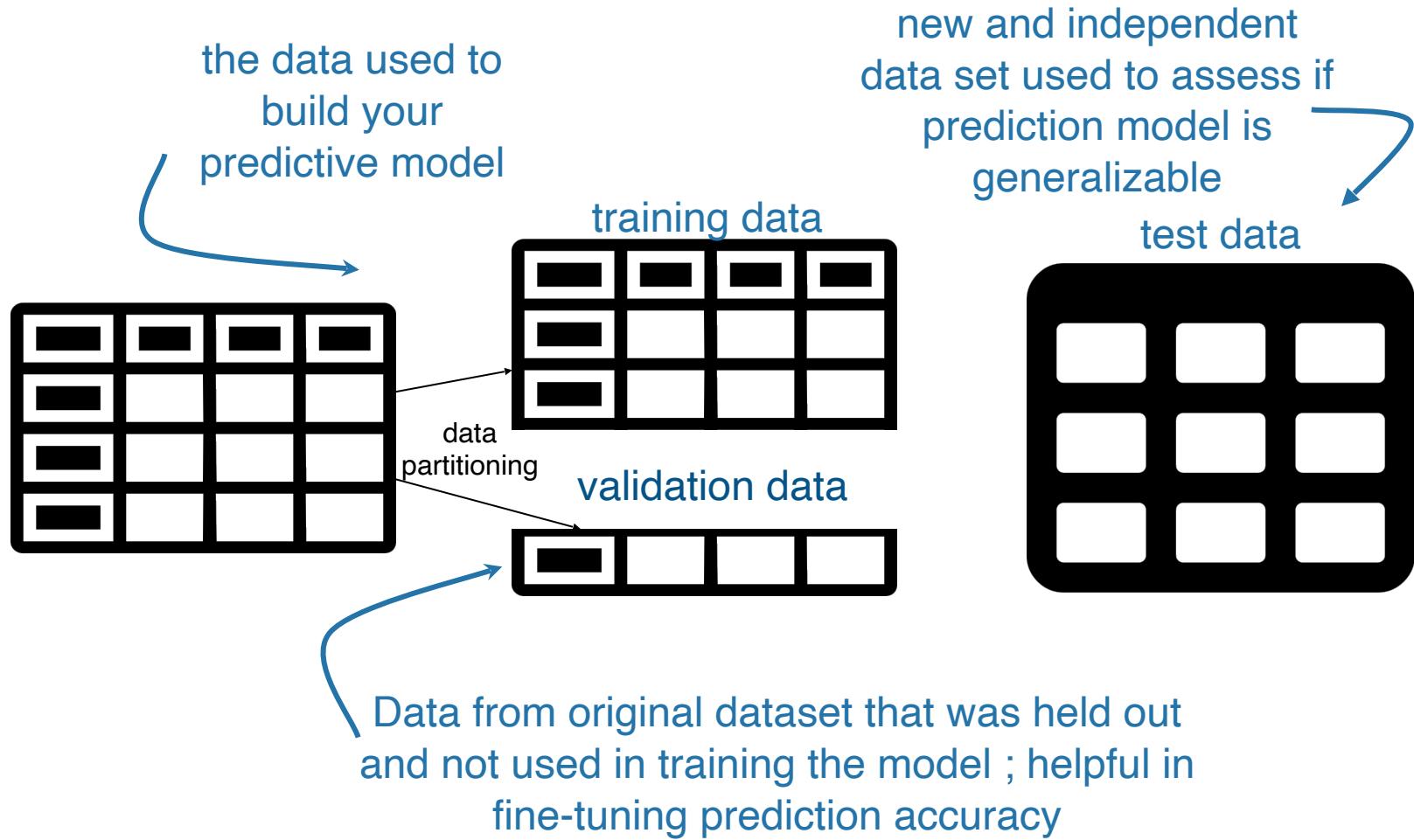
feature selection



model assessment



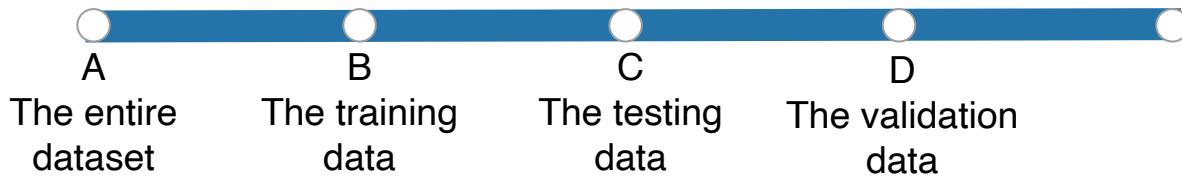
data partitioning

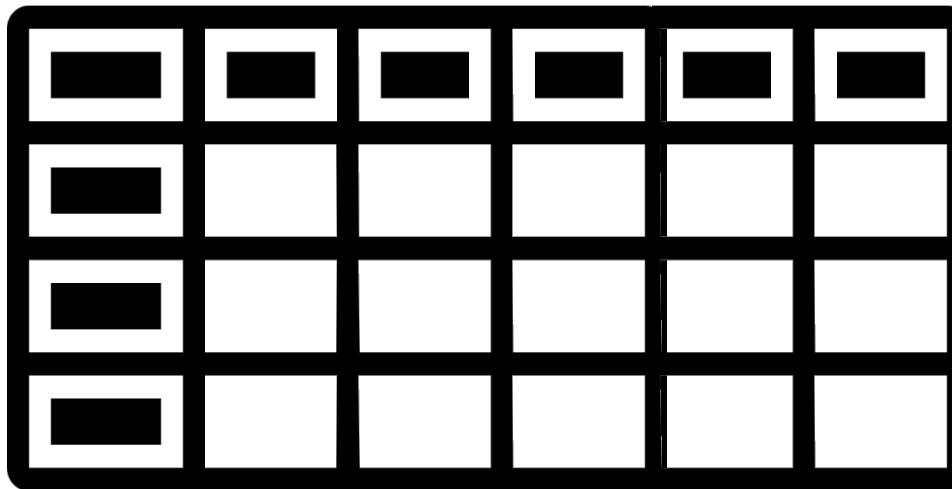




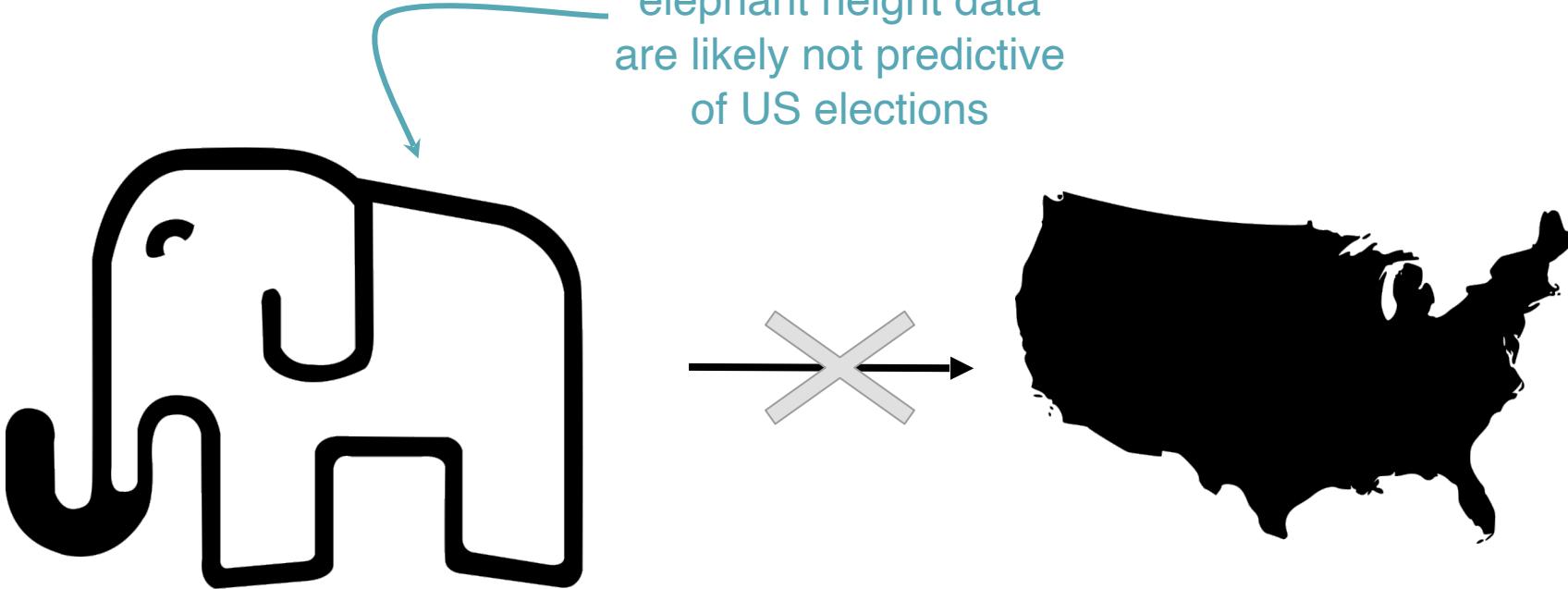
Data Partitioning

What portion of the data are typically used for generating the model?

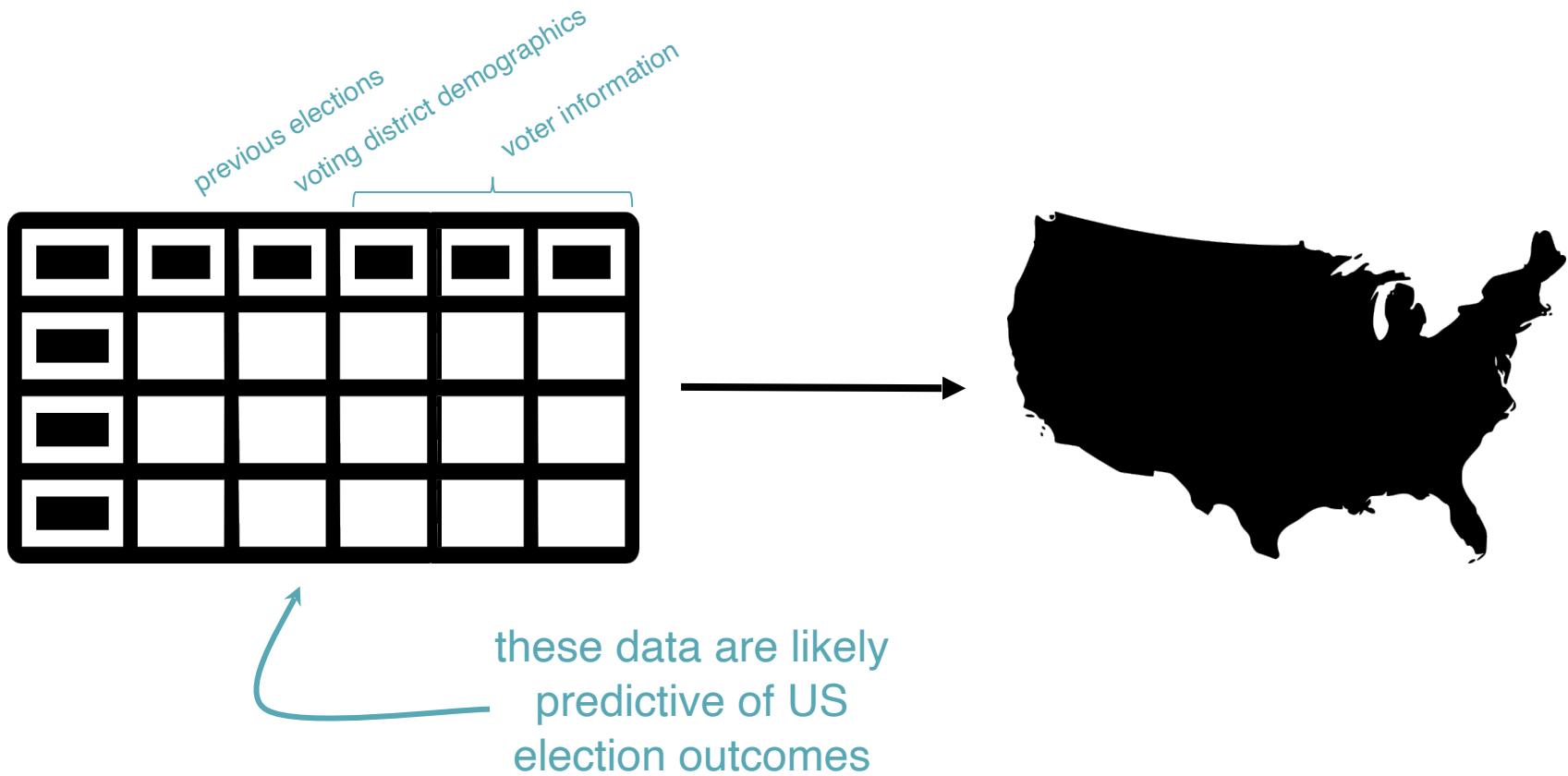


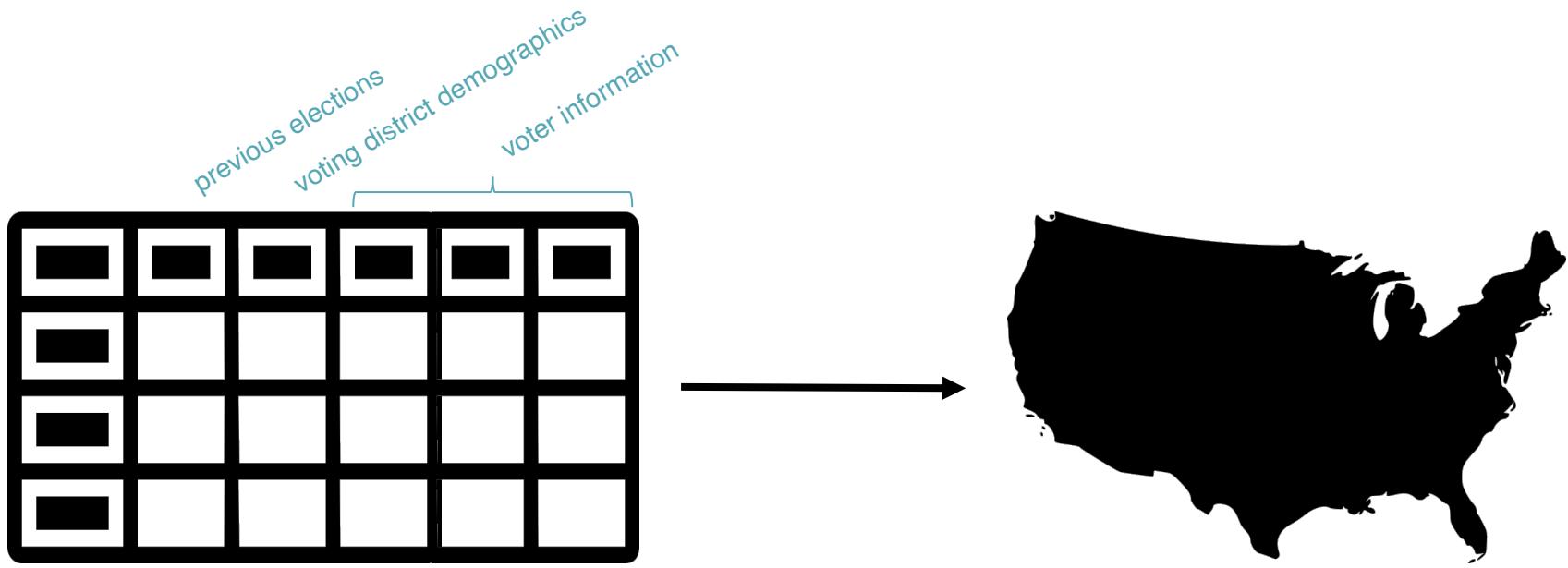


**feature
selection**

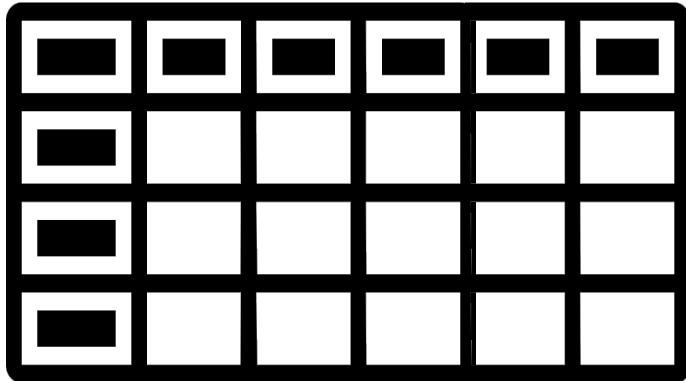


elephant height data
are likely not predictive
of US elections





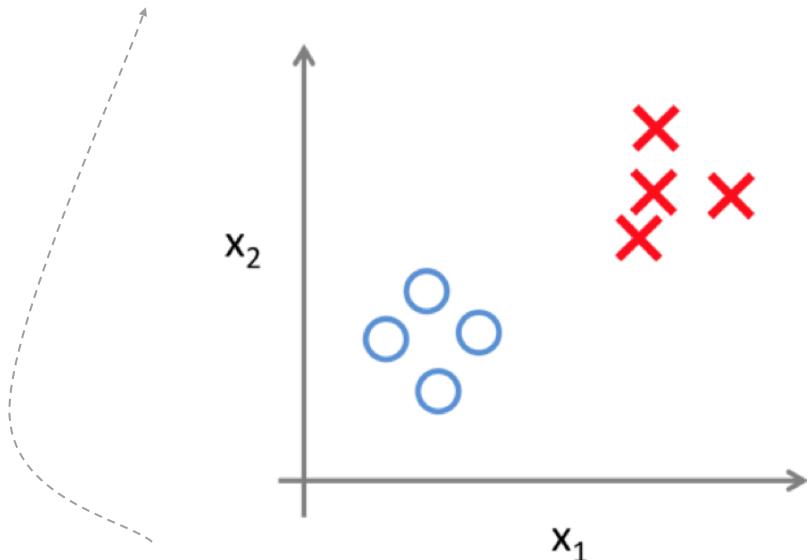
feature selection determines which variables are most predictive and includes them in the model



variables that can be used for accurate prediction exploit the relationship between the variables but do NOT mean that one causes the other

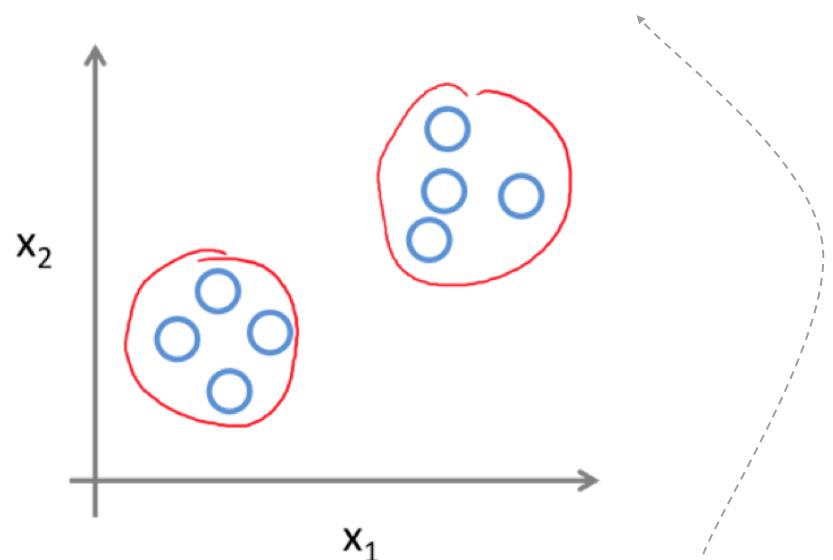
Two modes of machine learning

Supervised Learning



You tell the computer what features to use to classify the observations

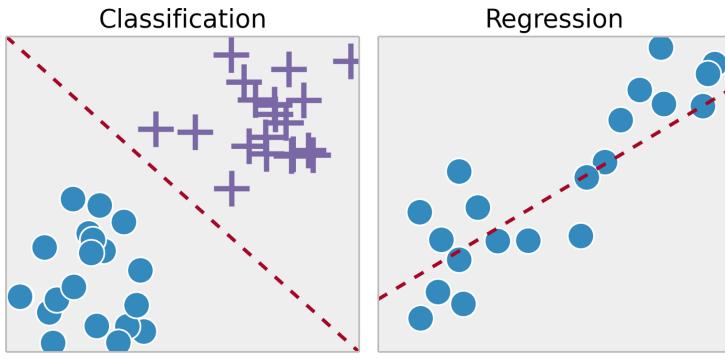
Unsupervised Learning



The computer determines how to classify based on properties within the data

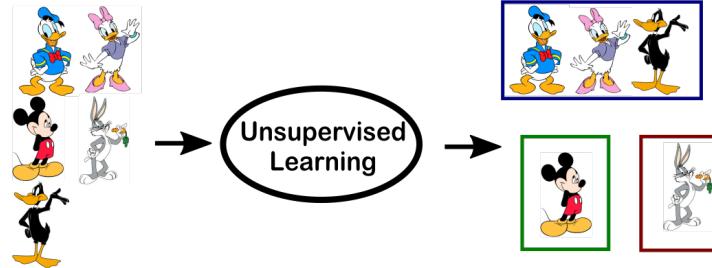
Approaches to machine learning

Supervised Learning



categorical variables

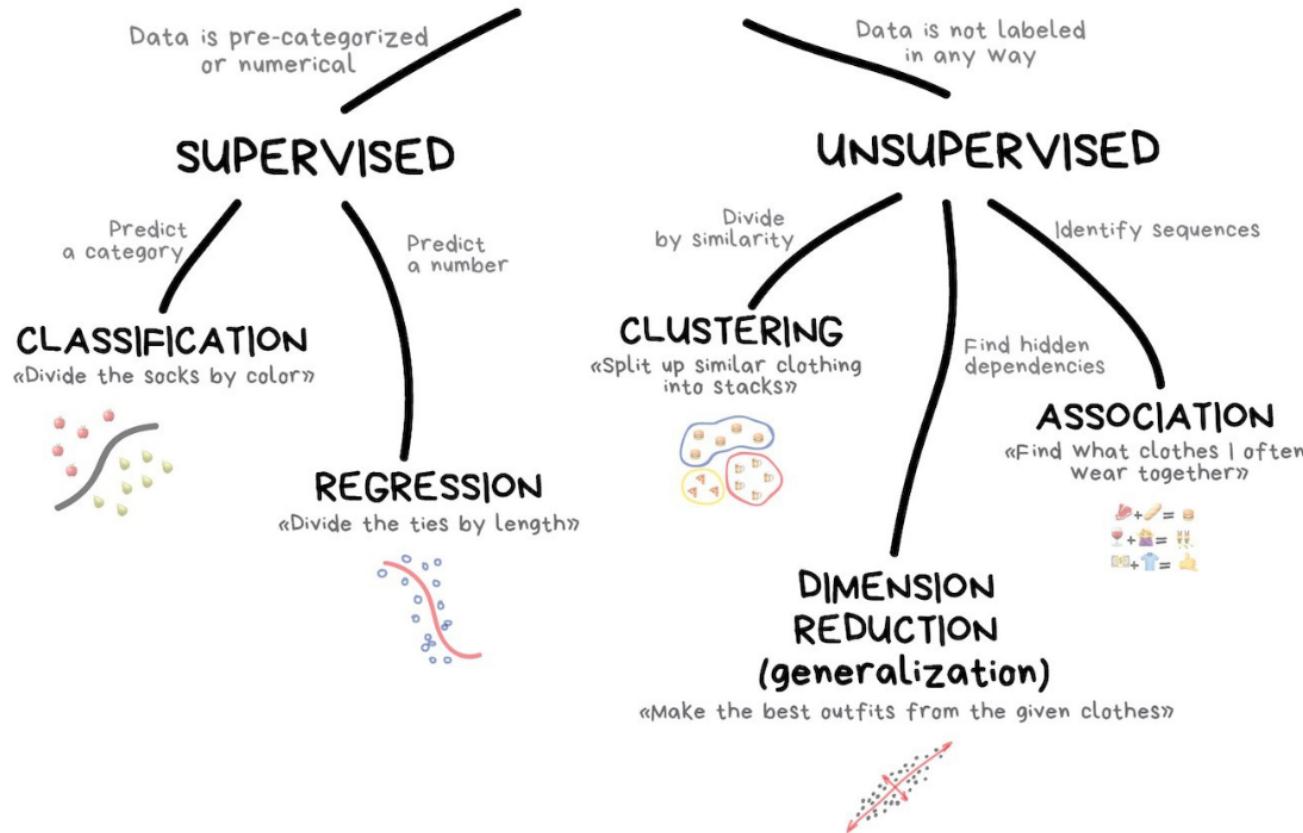
Unsupervised Learning

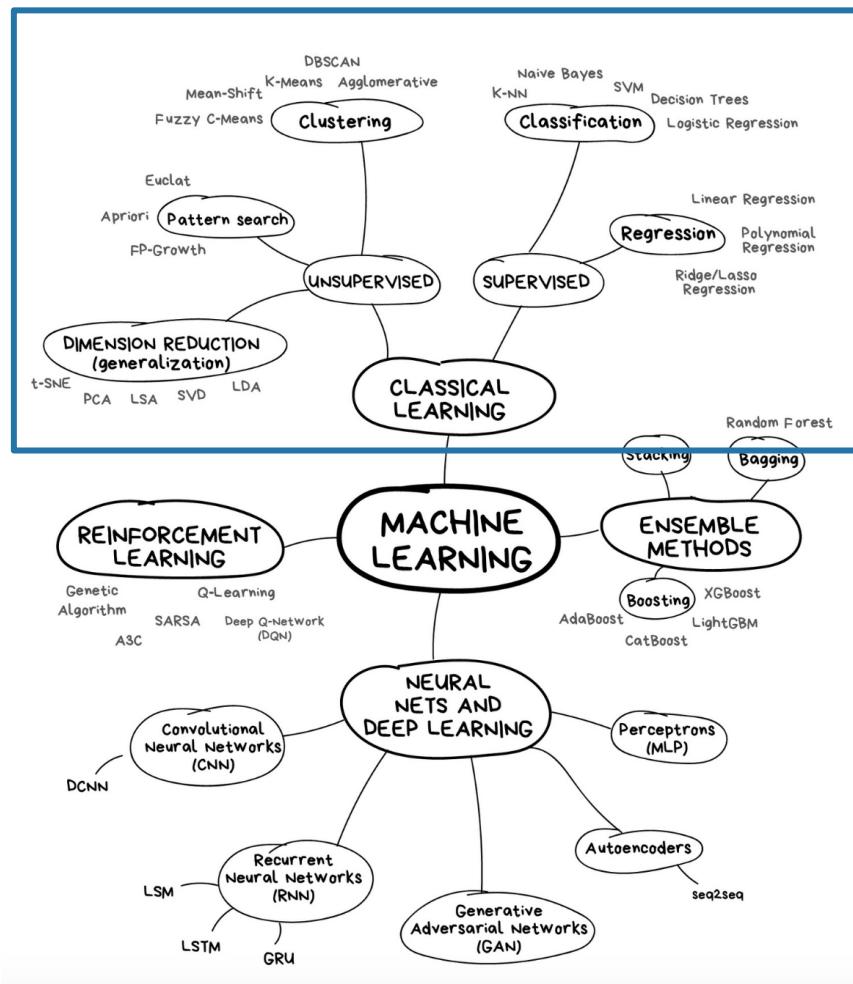


Clustering (categorical)
& dimensionality reduction (continuous)

can automatically identify
structure in data

CLASSICAL MACHINE LEARNING



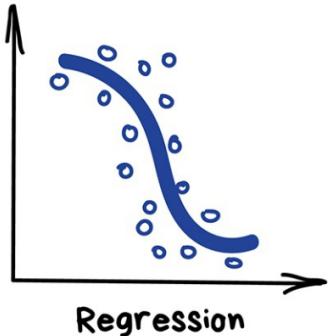


Regression

"Draw a line through these dots. Yep, that's the machine learning"

Today this is used for:

- Stock price forecasts
- Demand and sales volume analysis
- Medical diagnosis
- Any number-time correlations

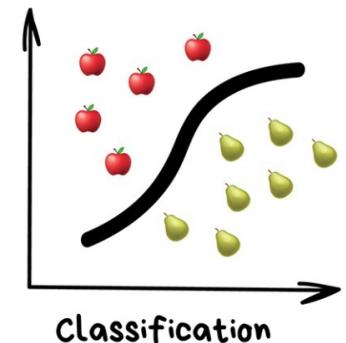


Classification

"Splits objects based at one of the attributes known beforehand. Separate socks by based on color, documents based on language, music by genre"

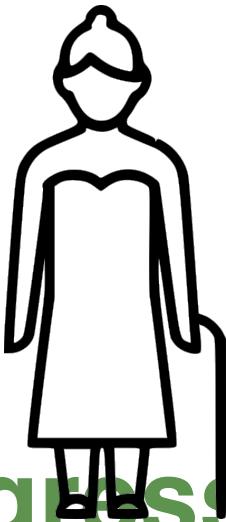
Today used for:

- Spam filtering
- Language detection
- A search of similar documents
- Sentiment analysis
- Recognition of handwritten characters and numbers
- Fraud detection



Popular algorithms are Linear and Polynomial regressions.

Popular algorithms: Naive Bayes, Decision Tree, Logistic Regression, K-Nearest Neighbours, Support Vector Machine



Regression

:

predicting
continuous variables

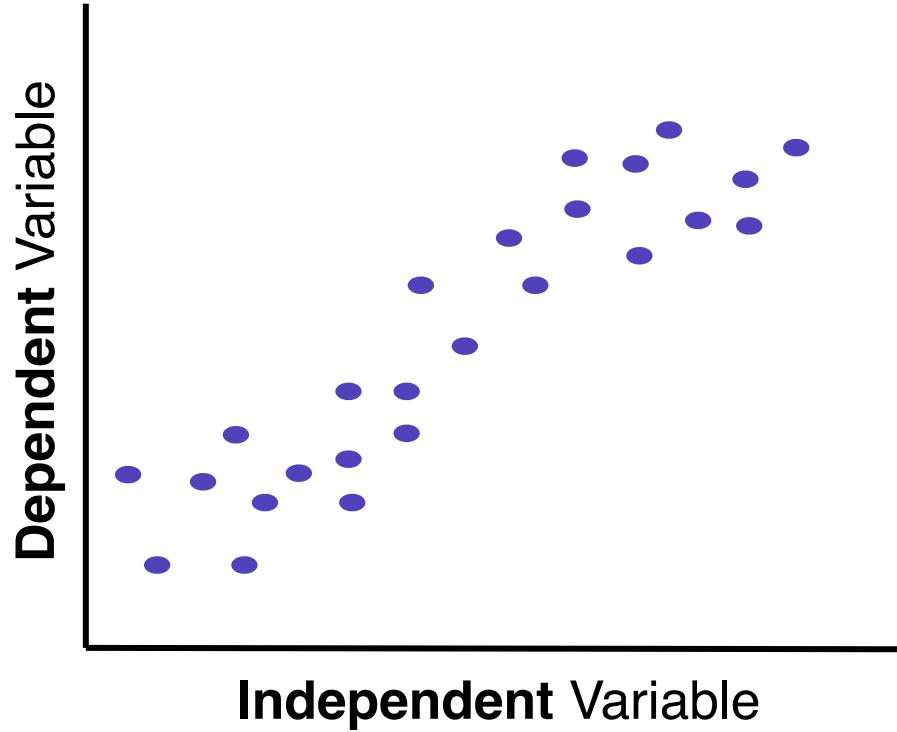
continuous variable
prediction

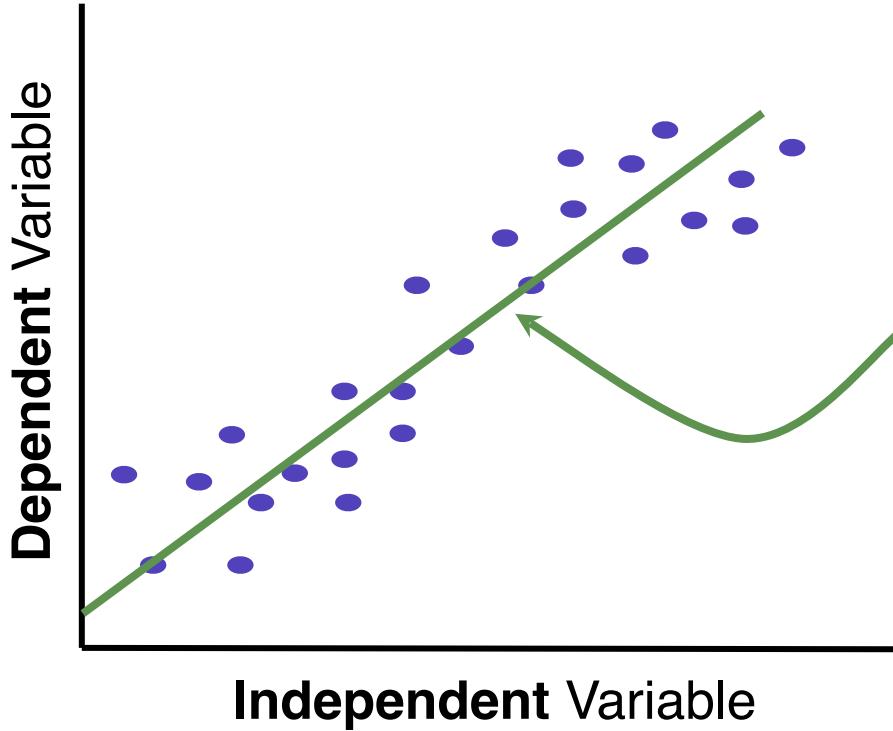


Classification:

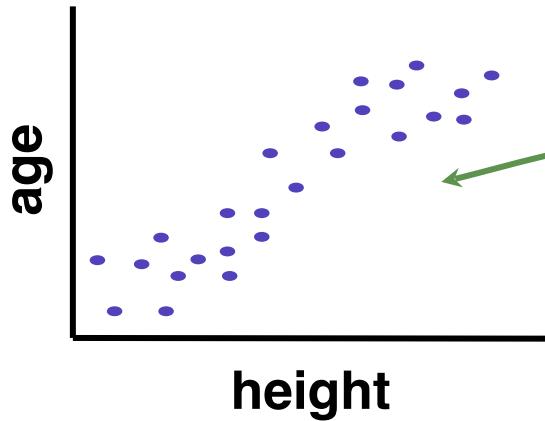
predicting categorical
variables

(i. categorical variable
prediction)

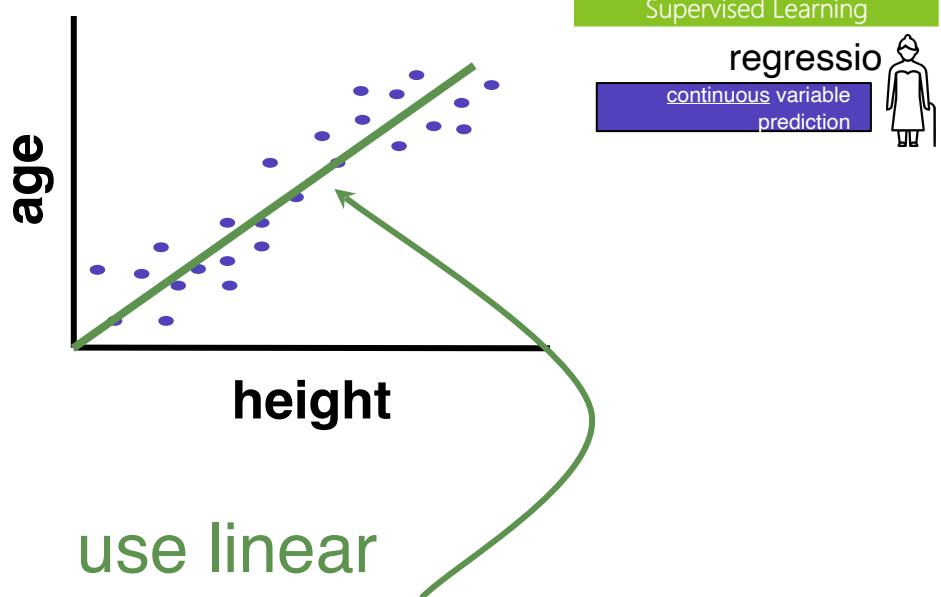
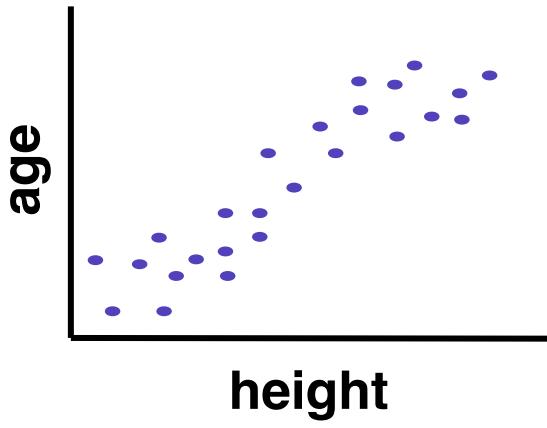




We'll use the linear relationship between variables to generate a **predictive model**



the training data
will be used to
build the
predictive
model



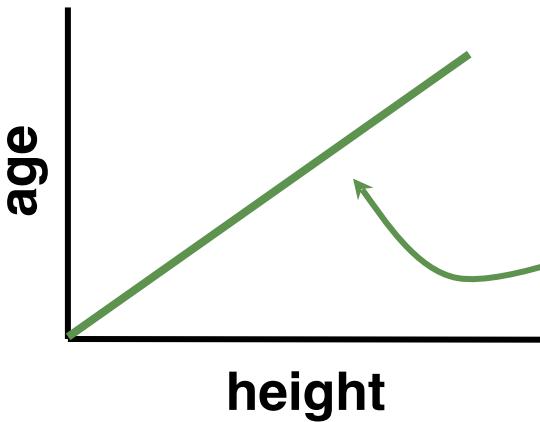
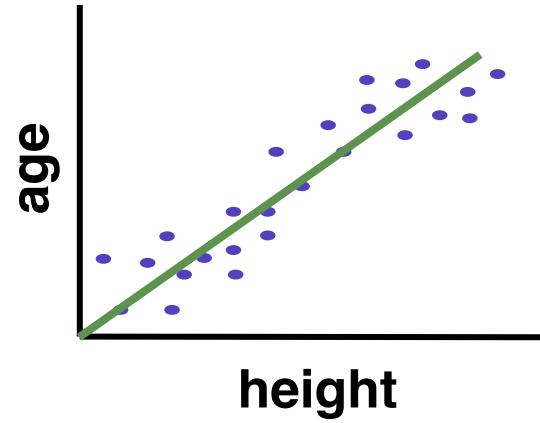
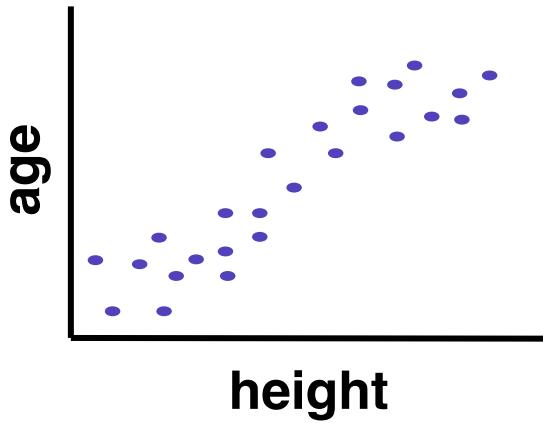
use linear
regression to
model the
relationship

Supervised Learning

regressio

continuous variable
prediction



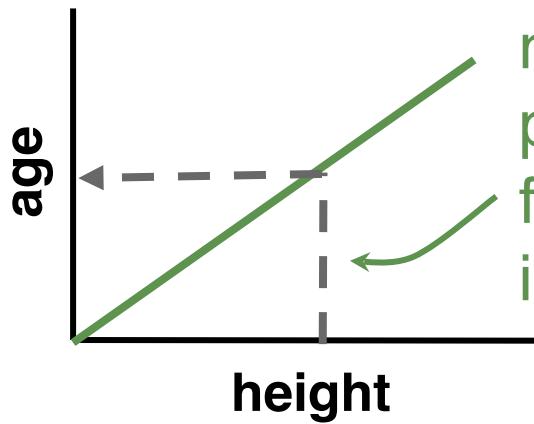
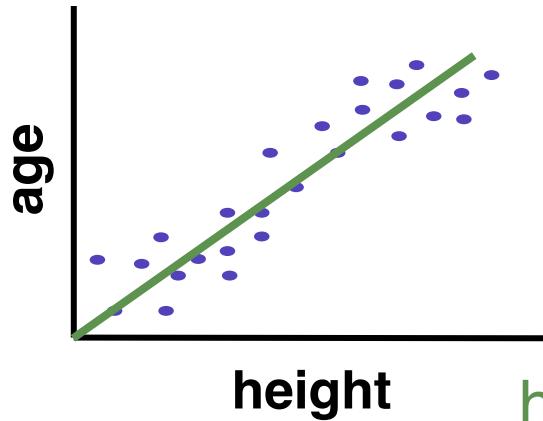
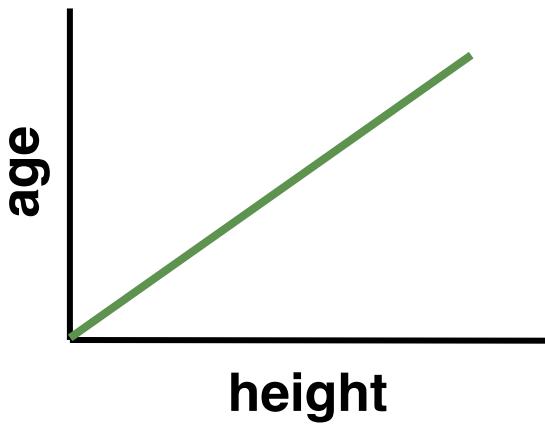
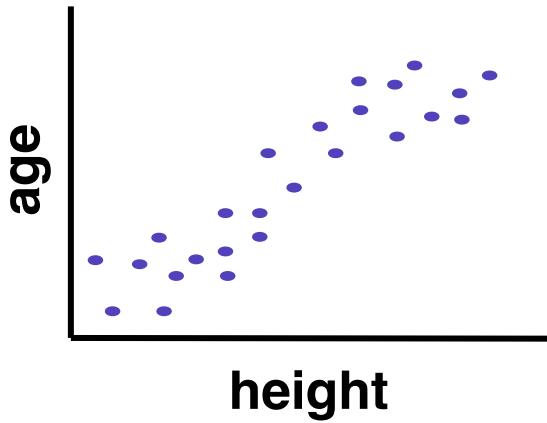


For prediction, the individual values in the training data are *not* important. We only need the model.

Supervised Learning

regressio

continuous variable
prediction

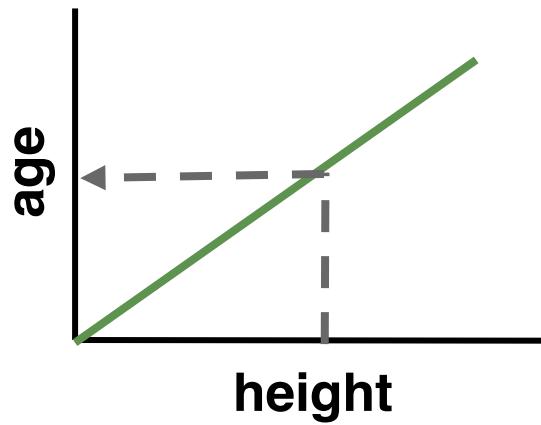
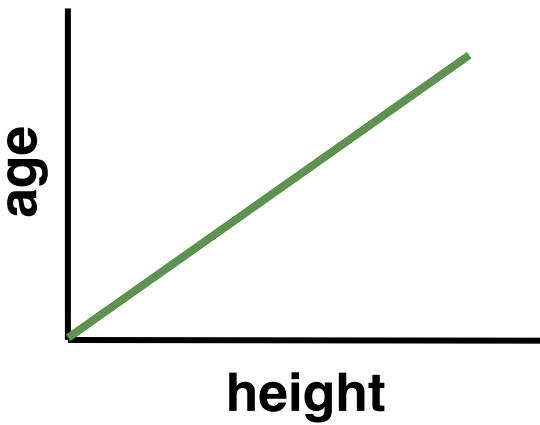
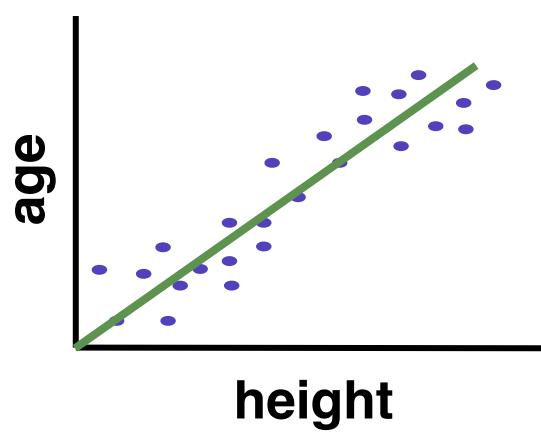
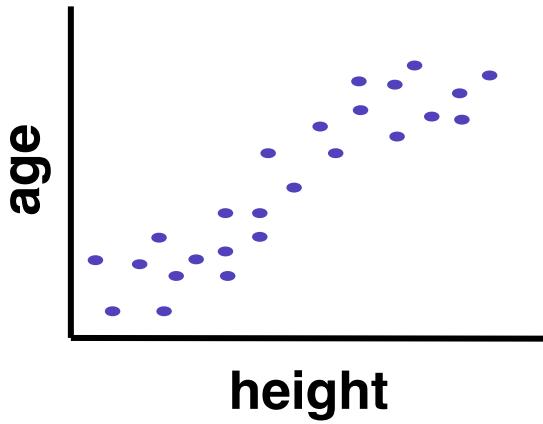


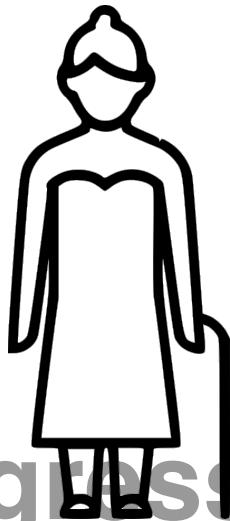
how we'll
make
predictions
for a future
individual

Supervised Learning

regressio

continuous variable
prediction





Regression
:
predicting
continuous variables
(i.e. Age)

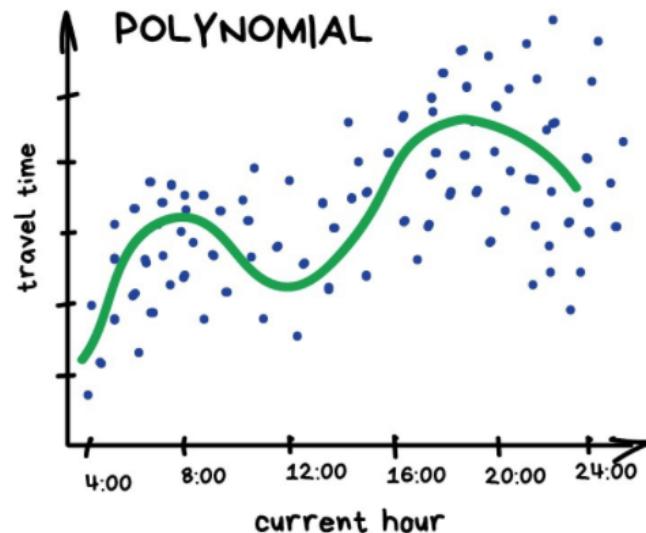
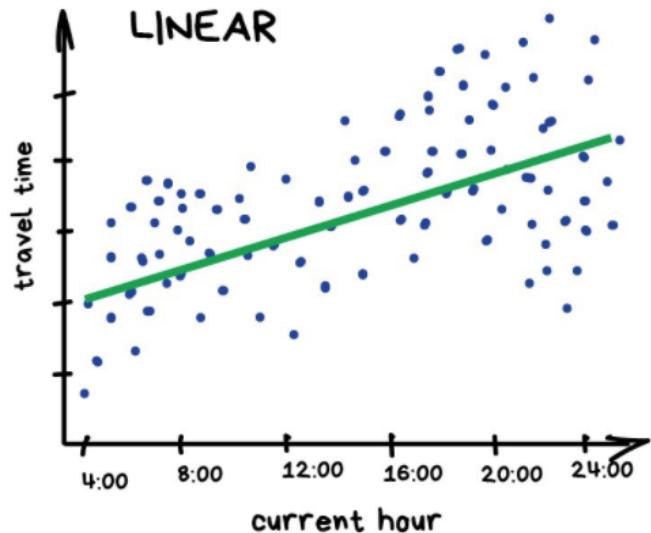


Classification

on:

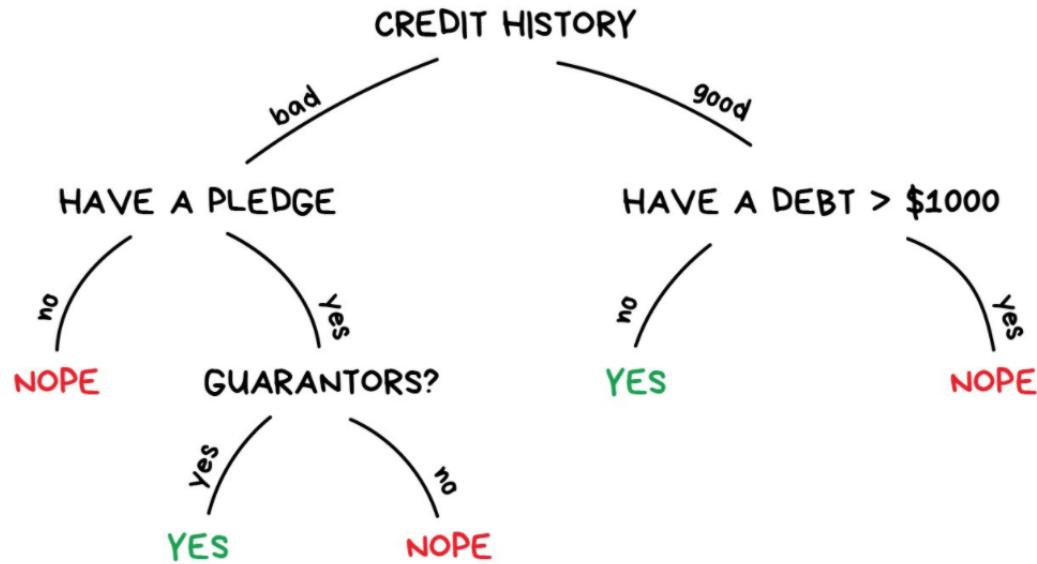
predicting categorical
variables
(i.e. give a loan?)

PREDICT TRAFFIC JAMS



REGRESSION

GIVE A LOAN?

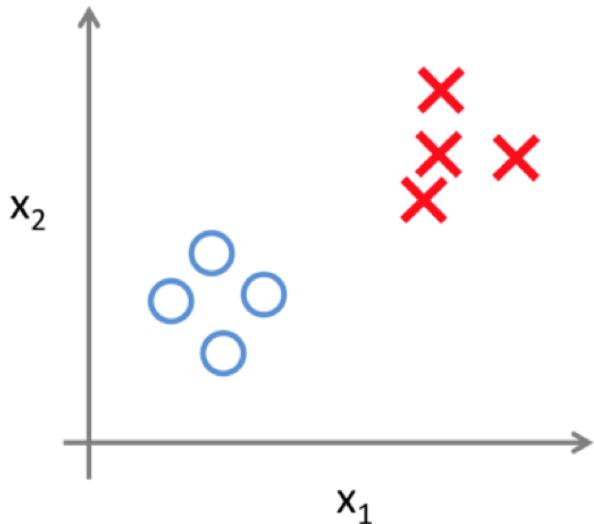


DECISION TREE

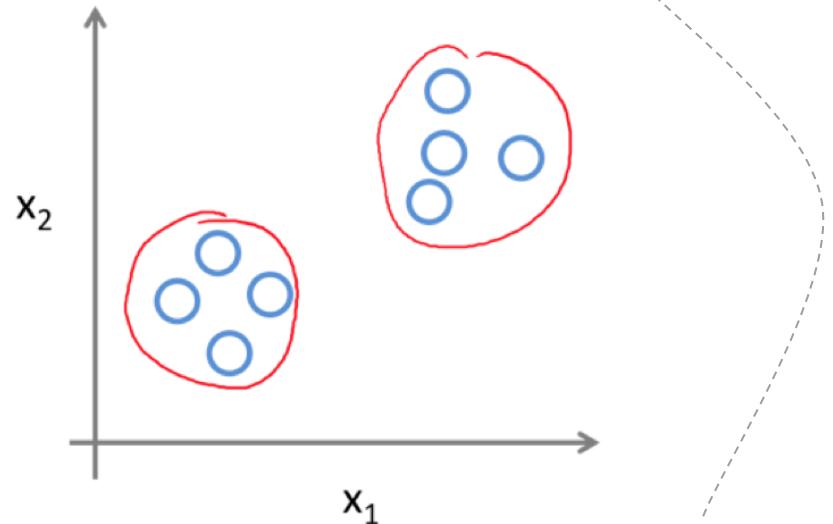
Unsupervised Learning

To modes of machine learning

Supervised Learning



Unsupervised Learning



The computer determines how to classify based on properties within the data

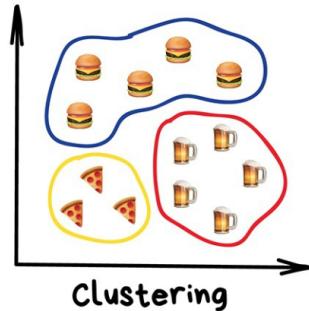
Dimensionality Reduction (Generalization)

Clustering

*"Divides objects based on unknown features.
Machine chooses the best way"*

Nowadays used:

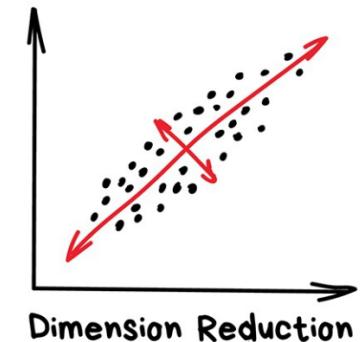
- For market segmentation (types of customers, loyalty)
- To merge close points on a map
- For image compression
- To analyze and label new data
- To detect abnormal behavior



"Assembles specific features into more high-level ones"

Nowadays is used for:

- Recommender systems (★)
- Beautiful visualizations
- Topic modeling and similar document search
- Fake image analysis
- Risk management



Popular algorithms: [K-means clustering](#), [Mean-Shift](#), [DBSCAN](#)

Popular algorithms: [Principal Component Analysis \(PCA\)](#), [Singular Value Decomposition \(SVD\)](#), [Latent Dirichlet allocation \(LDA\)](#), [Latent Semantic Analysis \(LSA, pLSA, GLSA\)](#), [t-SNE](#) (for visualization)

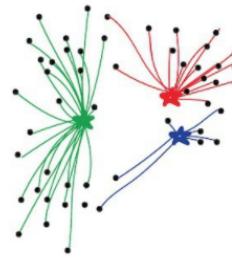
PUT KEBAB KIOSKS IN THE OPTIMAL WAY

(also illustrating the K-means method)

Unsupervised Learning



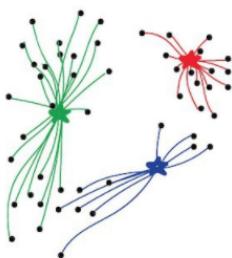
1. Put kebab kiosks in random places in city



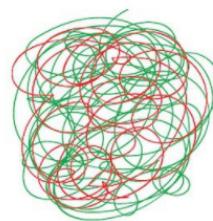
2. Watch how buyers choose the nearest one



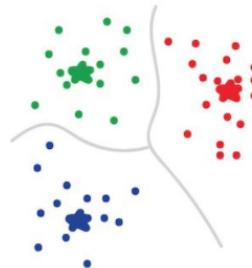
3. Move kiosks closer to the centers of their popularity



4. Watch and move again



5. Repeat a million times



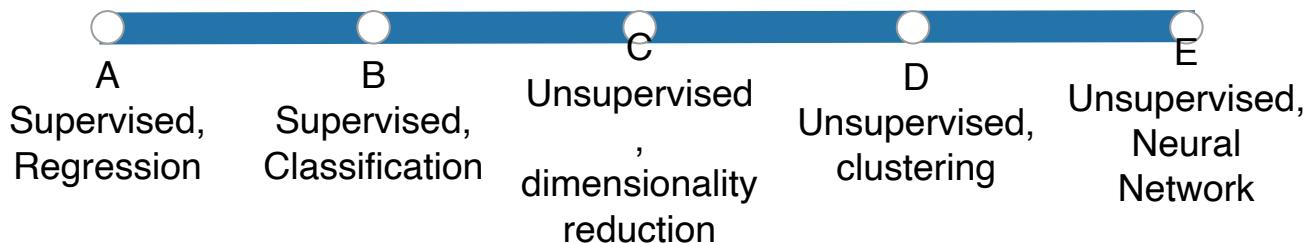
6. Done!
You're god of kebabs!

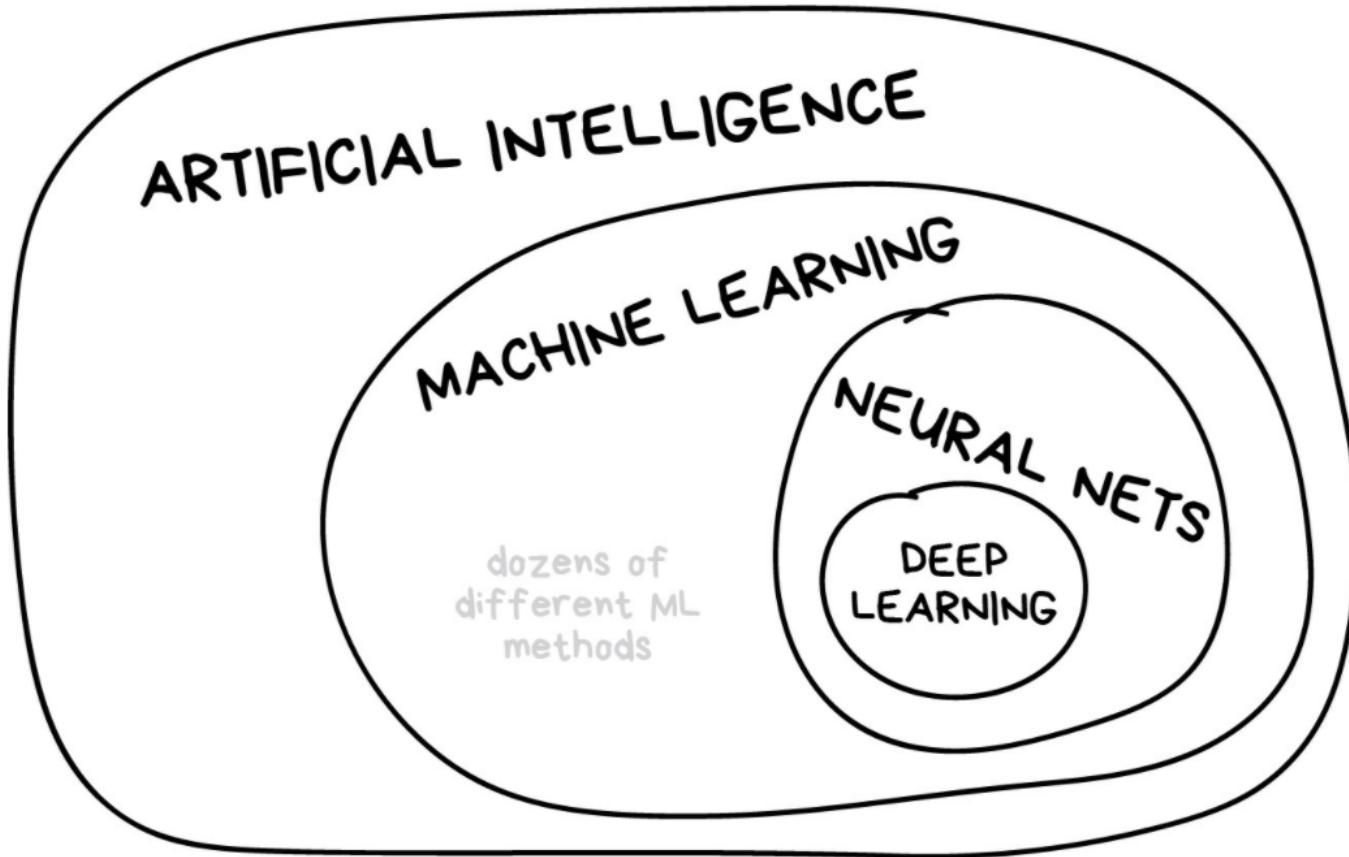


Prediction Approach

You want to predict someone's emotion based on an image.

How would you approach this with machine learning?





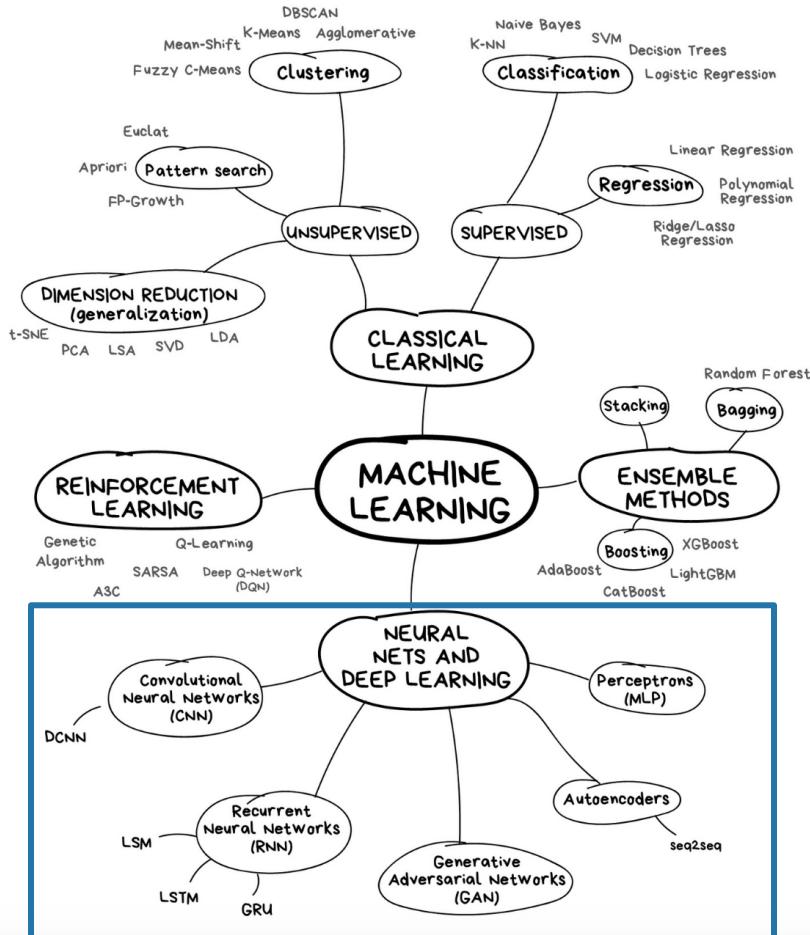
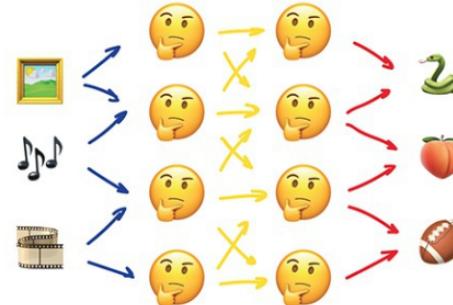


Image source: https://vas3k.com/blog/machine_learning/

"We have a thousand-layer network, dozens of video cards, but still no idea where to use it. Let's generate cat pics!"

Used today for:

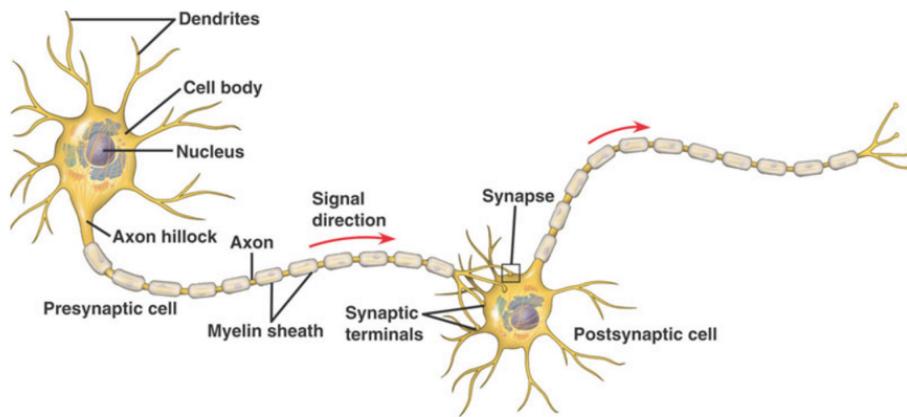
- Replacement of all algorithms above
- Object identification on photos and videos
- Speech recognition and synthesis
- Image processing, style transfer
- Machine translation



Neural Networks

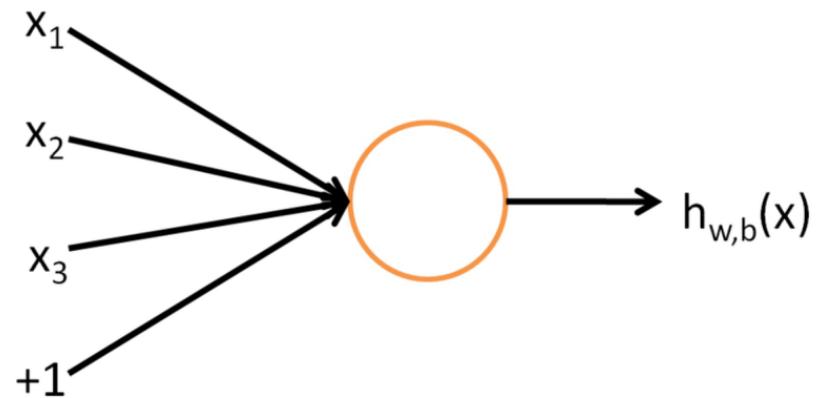
Popular architectures: Perceptron, Convolutional Network (CNN), Recurrent Networks (RNN), Autoencoders

WHAT IS A NEURON?



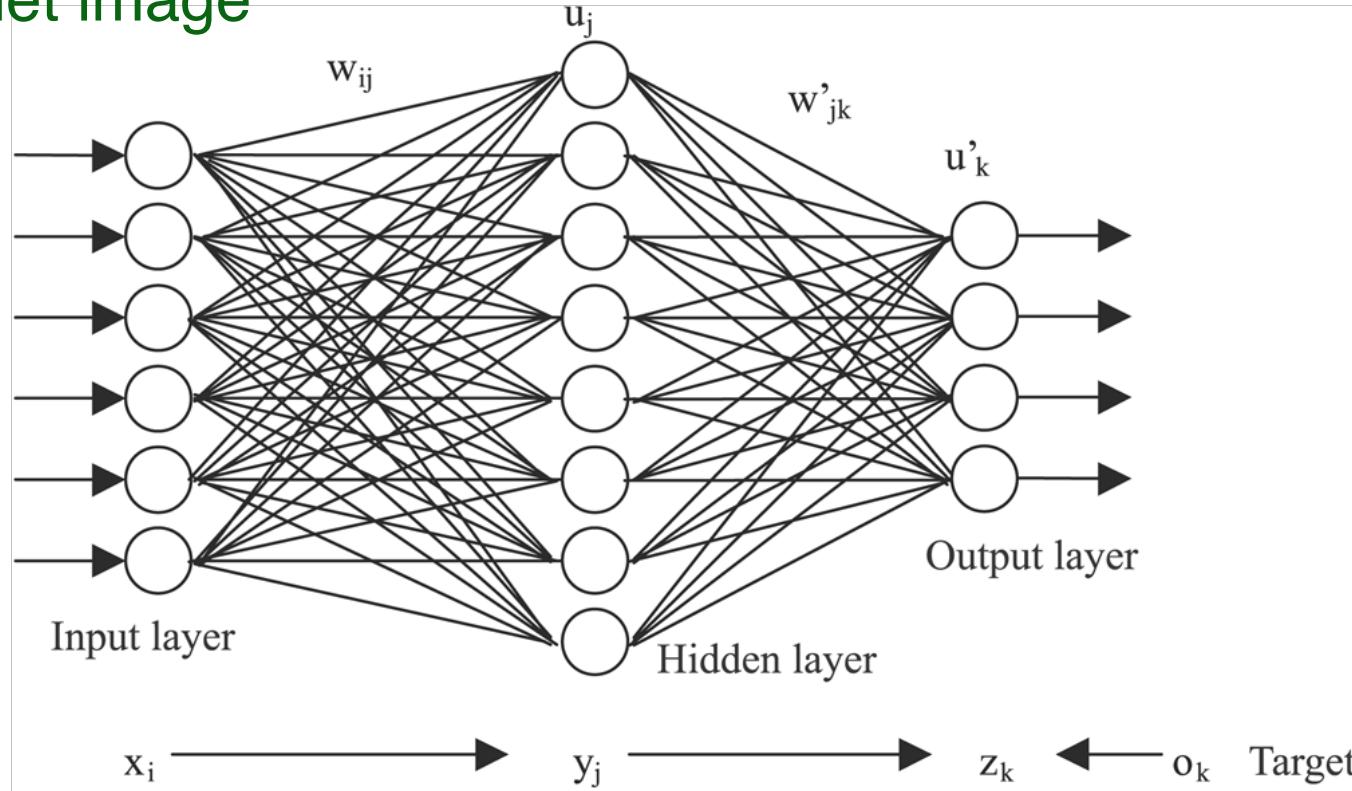
- Receives signal on synapse
- When trigger sends signal on axon

MATHEMATICAL NEURON



- Mathematical abstraction, inspired by biological neuron
- Either on or off based on sum of input

This will likely not be the last time you see this (mostly unhelpful) neural net image



HOW A DEEP NEURAL NETWORK SEES

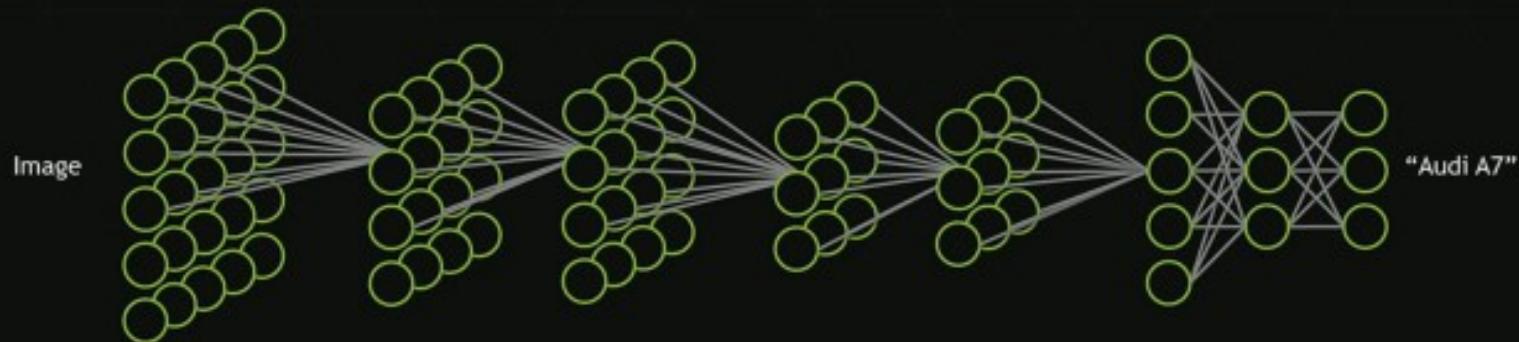
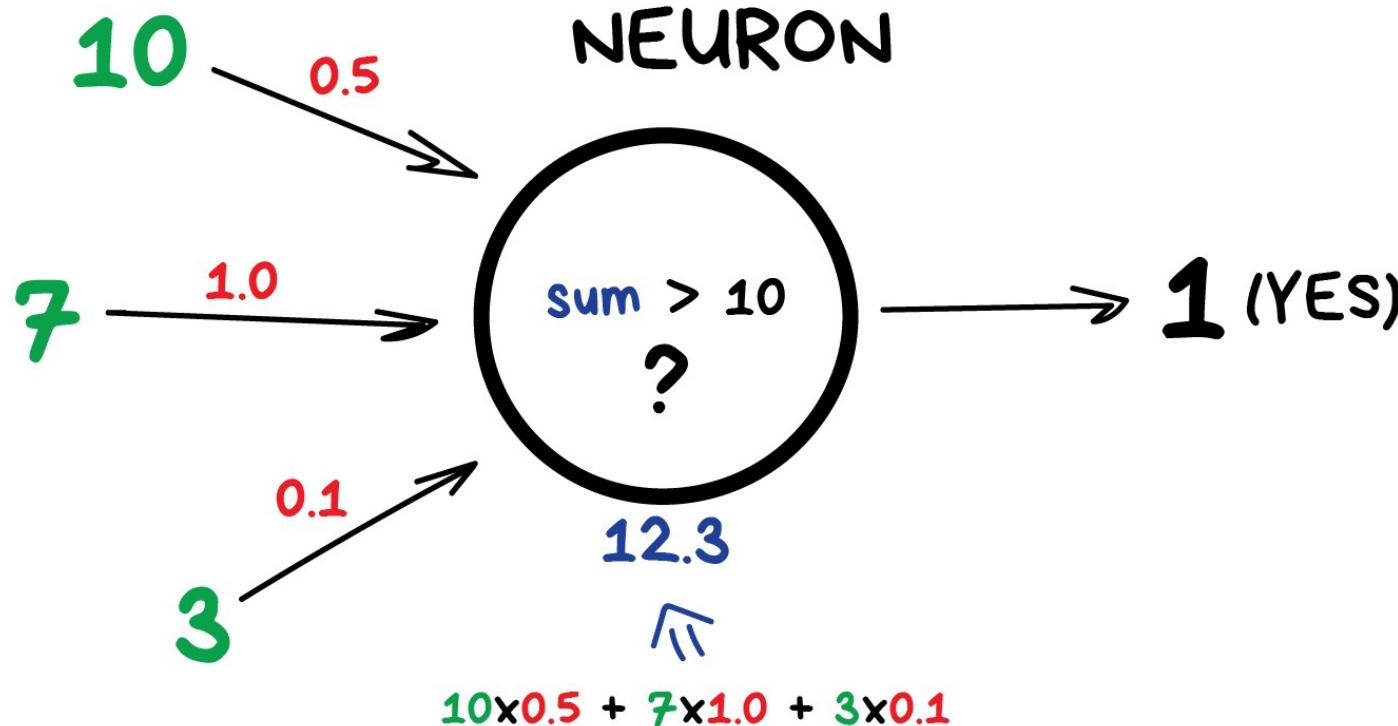
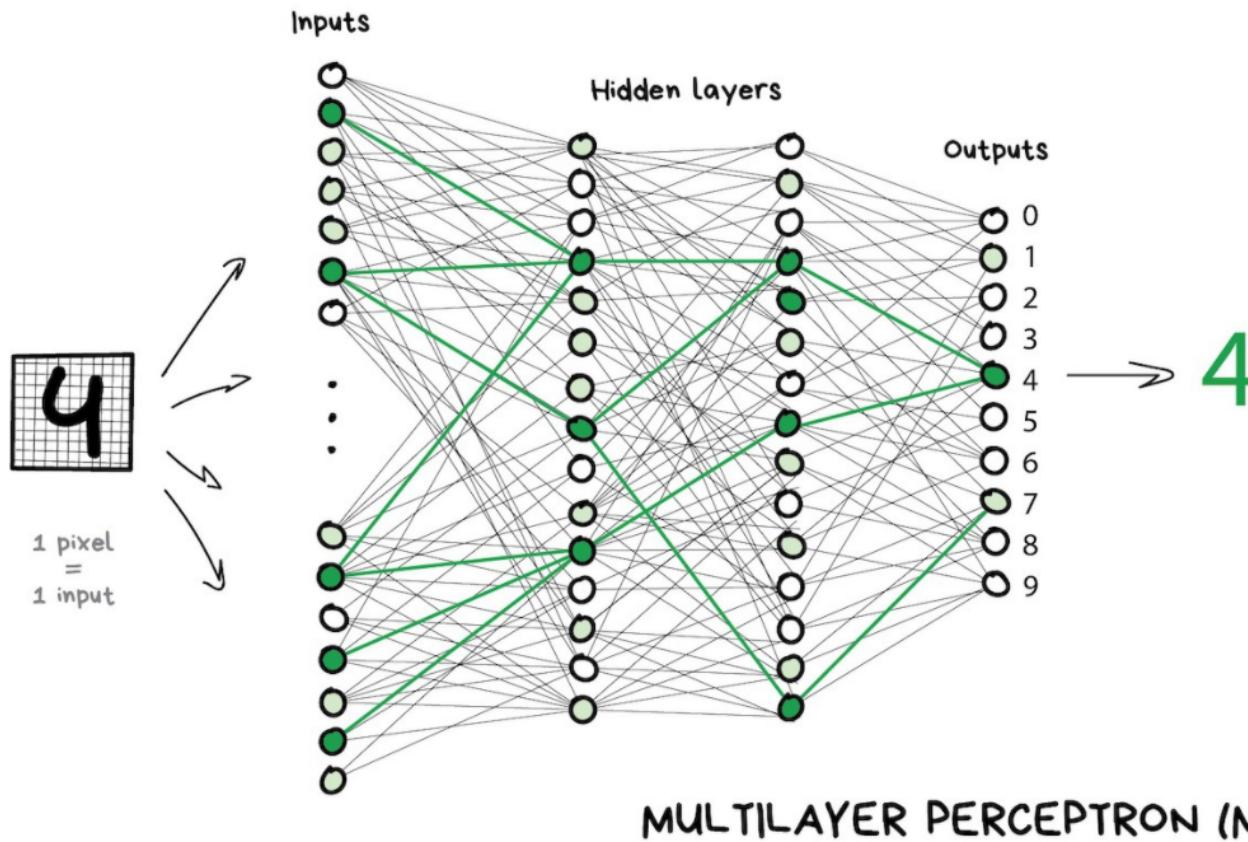


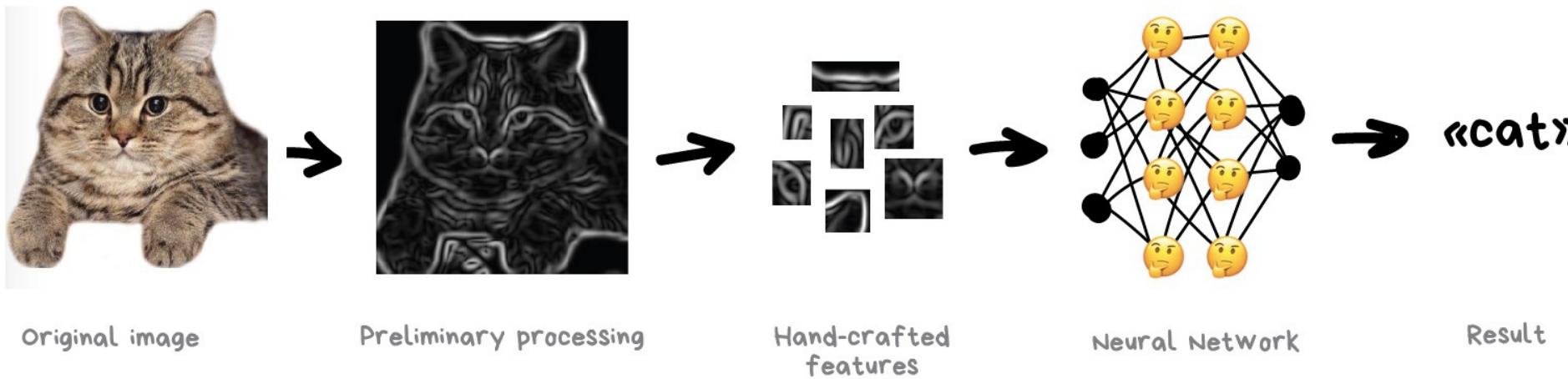
Image source: "Unsupervised Learning of Hierarchical Representations with Convolutional Deep Belief Networks" (ICML 2009 & Comm. ACM 2011).
Honglak Lee, Roger Grosse, Rajesh Ranganath, and Andrew Ng.

These weights tell the neuron to respond more to one input and less to another. Weights are adjusted when training — that's how the network learns. Basically, that's all there is to it.

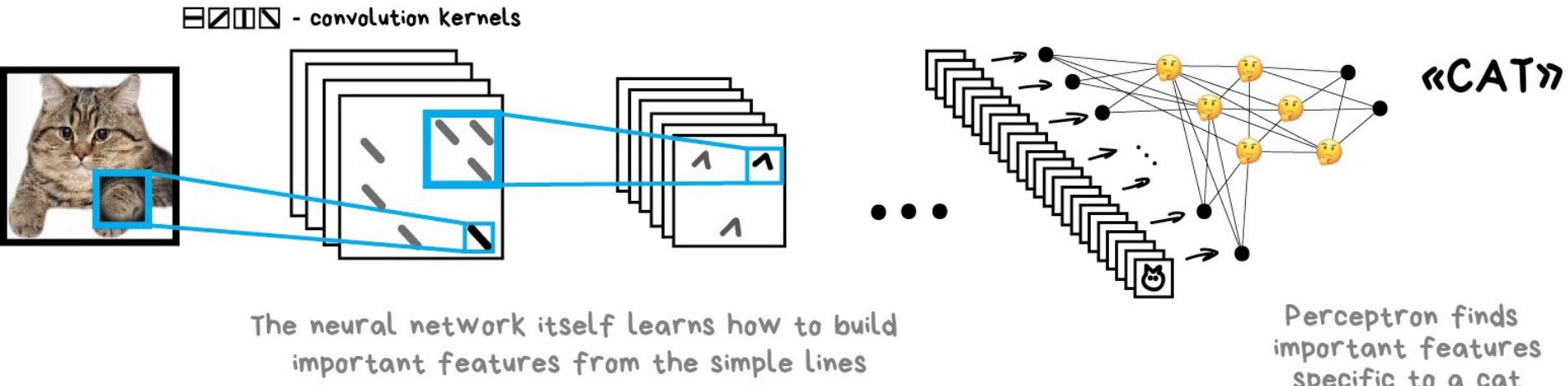




Manually labeling used to be the way...



CNNs avoid manual features



“CNNs are all the rage right now. They are used to search for objects on photos and in videos, face recognition, style transfer, generating and enhancing images, creating effects like slow-mo and improving image quality. Nowadays CNNs are used in all the cases that involve pictures and videos.”

CONVOLUTIONAL NEURAL NETWORK (CNN)



Much of DL success comes from semi-supervised tricks to avoid large hand labelled datasets

Masked LM

- **Solution:** Mask out $k\%$ of the input words, and then predict the masked words
 - We always use $k = 15\%$

the man went to the [MASK] to buy a [MASK] of milk

↑ ↑
store gallon

- Too little masking: Too expensive to train
- Too much masking: Not enough context



all your base are belong to
@jasongfleischer

...

I used [@OpenAI](#) GPT-3 to make some #programming themed candy hearts for you this Valentine's Day. Hope you feel the 100% computer generated ❤️! Here's a selection of the one's I liked best (thread 1/3)





all your base are belong to @jasongfleischer · 1h

...

It's too long to fit on the candy heart generator but 🍬🍬💀 (bonus post, now the thread is over really)

candy theme: normal

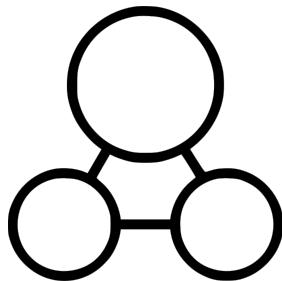
messages: BESTIE, CUTIE PIE, SOUL MATE, SWEET PEA,
UR CUTE, YOU + ME, BE MINE, PICK ME, KISS ME, LOVE
BIRDS, MARRY ME, OOO LA LA, TRUE LOVE, WINK WINK,
XOXO

candy theme: programming

messages:

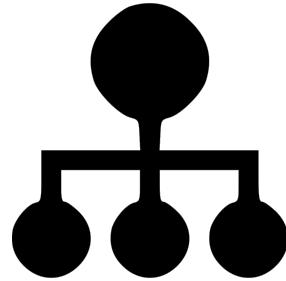


(1) I <3 programmers, (2) Programmers do it better, (3) If you can't code it, you can't date it, (4) Code is poetry, (5) Programmers make the world go round, (6) If you can't code, you can't love, (7) Love is the language of the future, (8) I heart code, (9) Code is life, (10) If you don't code, you

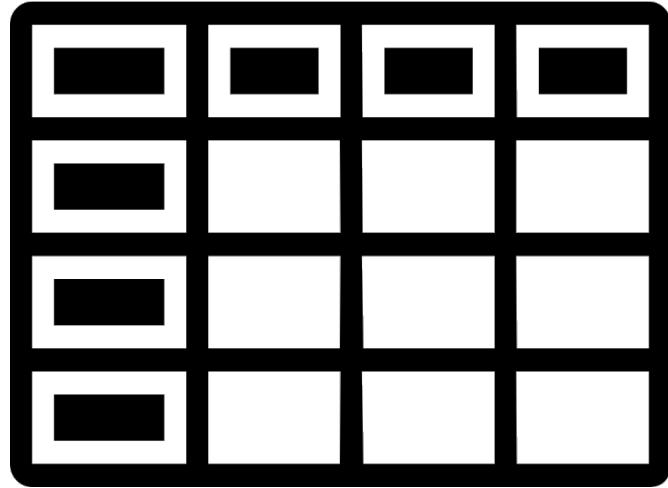


A 2x2 grid table with the following symbols:
Top-left: +
Top-right: -
Bottom-left: x
Bottom-right: =

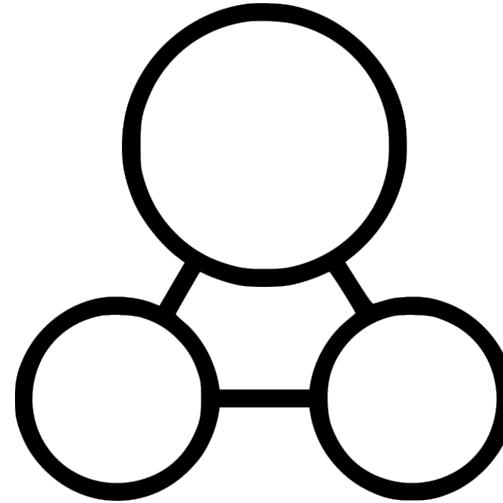
+	-
x	=



model selection



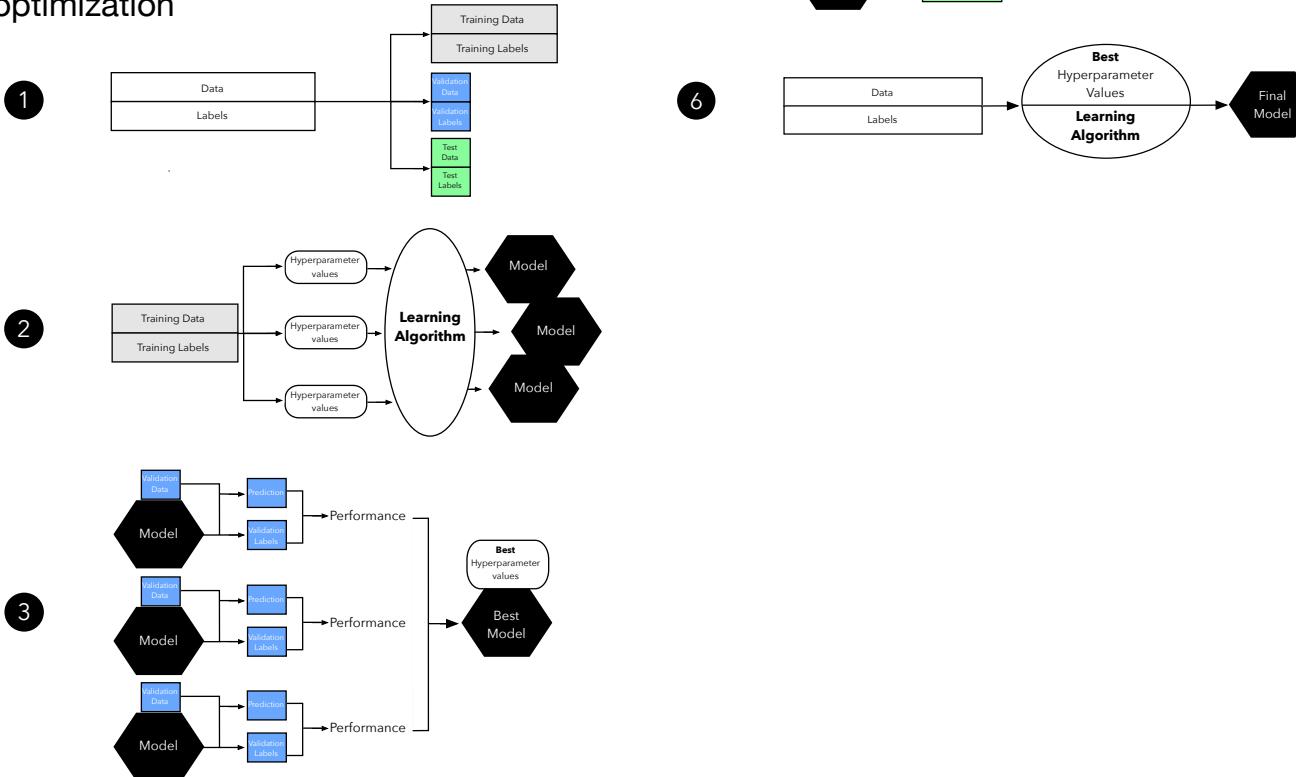
big
datasets



simple
models

3-Way Holdout

instead of "regular" holdout to avoid "data leakage" during hyperparameter optimization

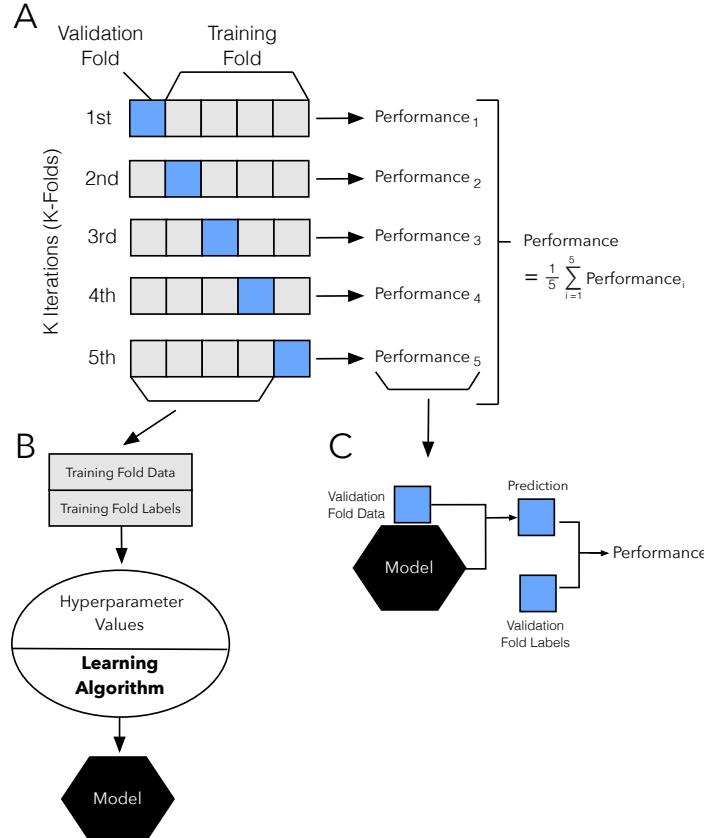


Model, feature, and hyperparameter selection

- Needs: Don't leak information from training/selection process into the test set!
- Trade-offs: Usually not enough data to have completely separate train, validation, test sets. Which one do we prioritize?
 - Low training data -> bad fit
 - Low validation data -> bad selection of model/feature/hparam
 - Low test data -> poor estimate of generalization performance

- Non-overlapping test data, overlapping training data among folds
- Small $k \rightarrow$ biased pessimistic from small training data
- Variance increases with k ; LOOCV is unbiased but hi variability

k -Fold Cross-Validation





model assessment

Root Mean Squared Error (RMSE)

$$RMSE = \sqrt{\frac{\sum_{i=1}^N (Predicted_i - Actual_i)^2}{N}}$$



A few outliers can lead to a big increase in RMSE, even if all the other predictions are pretty good

categorical variable
prediction
continuous variable
prediction

$$\text{Accuracy} = \frac{\# \text{ of samples predicted correctly}}{\# \text{ of samples predicted}} * 100$$

		Actual	
		Positive	
Predicted	Positive	True Positive (TP)	False Positive
	Negative	False Negative	True Negative (TN)

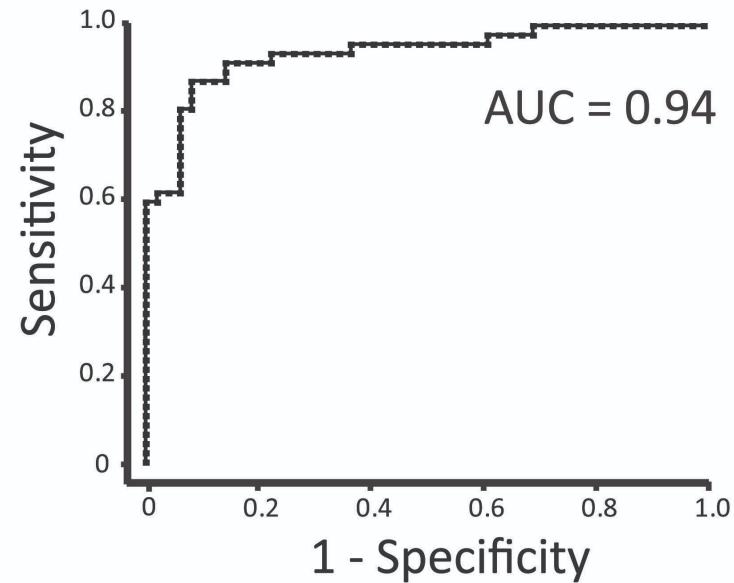
The diagram illustrates a 2x2 matrix for medical test results, comparing Actual status (Positive or Negative) against Predicted status (Positive or Negative). The matrix is divided into four quadrants:

- True Positive (TP):** Actual Positive, Predicted Positive. Image: A digital pregnancy test strip showing a single line for "not pregnant".
- False Positive:** Actual Positive, Predicted Negative. Image: A doctor examining a pregnant woman's belly.
- False Negative:** Actual Negative, Predicted Negative. Image: A doctor telling a patient they are pregnant.
- True Negative (TN):** Actual Negative, Predicted Positive. Image: A digital pregnancy test strip showing two lines for "pregnant".

		Actual	
		Positive	Negative
Predicted	Positive	True Positive (TP)	False Positive (FP)
	Negative	False Negative (FN)	True Negative (TN)

A 2x2 table is a type of confusion matrix

$$\text{Sensitivity} = \frac{\text{TP}}{\text{TP} + \text{FN}}$$



$$\text{Specificity} = \frac{\text{TN}}{\text{TN} + \text{FP}}$$

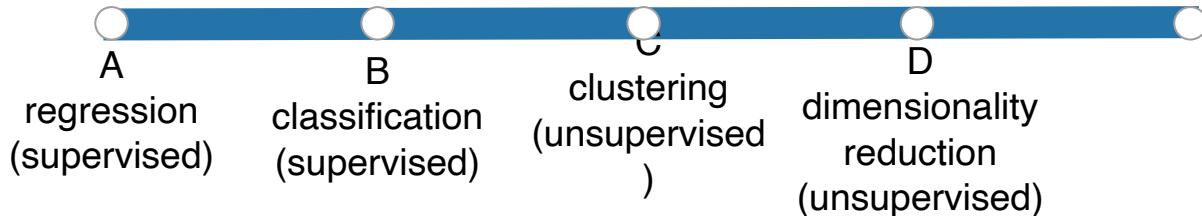
Accuracy	What % were predicted correctly?
Sensitivity	Of those that <i>were positives</i> , what % were predicted to be positive?
Specificity	Of those that were <i>negatives</i> , what % were predicted to be negative?



Prediction Approach

You've been given a dataset with a number of features and have been asked to predict each individual's age.

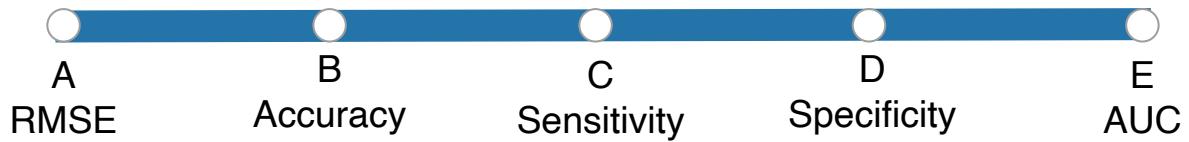
What prediction approach would you use?





Prediction Approach

After predicting each person's age, how would you assess your model?





Prediction Approach

Which would be the error value you'd want from your model?

