

Course Announcements

Due Today midnight - D3

Due Friday midnight - A2

Keep eyes open for
next quiz
project proposal requirements
D4

Machine Learning

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Lectures : <https://github.com/COGS108/Lectures-Wi23>

Did they summarize the data? **Yes** → Did they report the summaries without interpretation? **No** → Did they quantify whether the discoveries are likely to hold in a new sample? **Yes** → Are they trying to figure out how changing the average of one measurement affects another?

Predictive: apply machine learning techniques to data you have currently to generate a model that will be able to make a prediction on future data

Classic Statistics
(parametric & nonparametric)

Text Analysis

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

Causal

No → Are the data a corpus of text?

Yes

No → Are the observations spatially related?

Yes

Inferential

Geospatial Statistics

Predictive

Supervised Machine Learning

Did the computer decide the features of your model?

Unsupervised Machine Learning

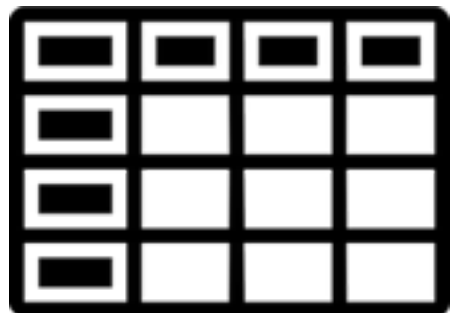
STOP!
Not a data analysis

- **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

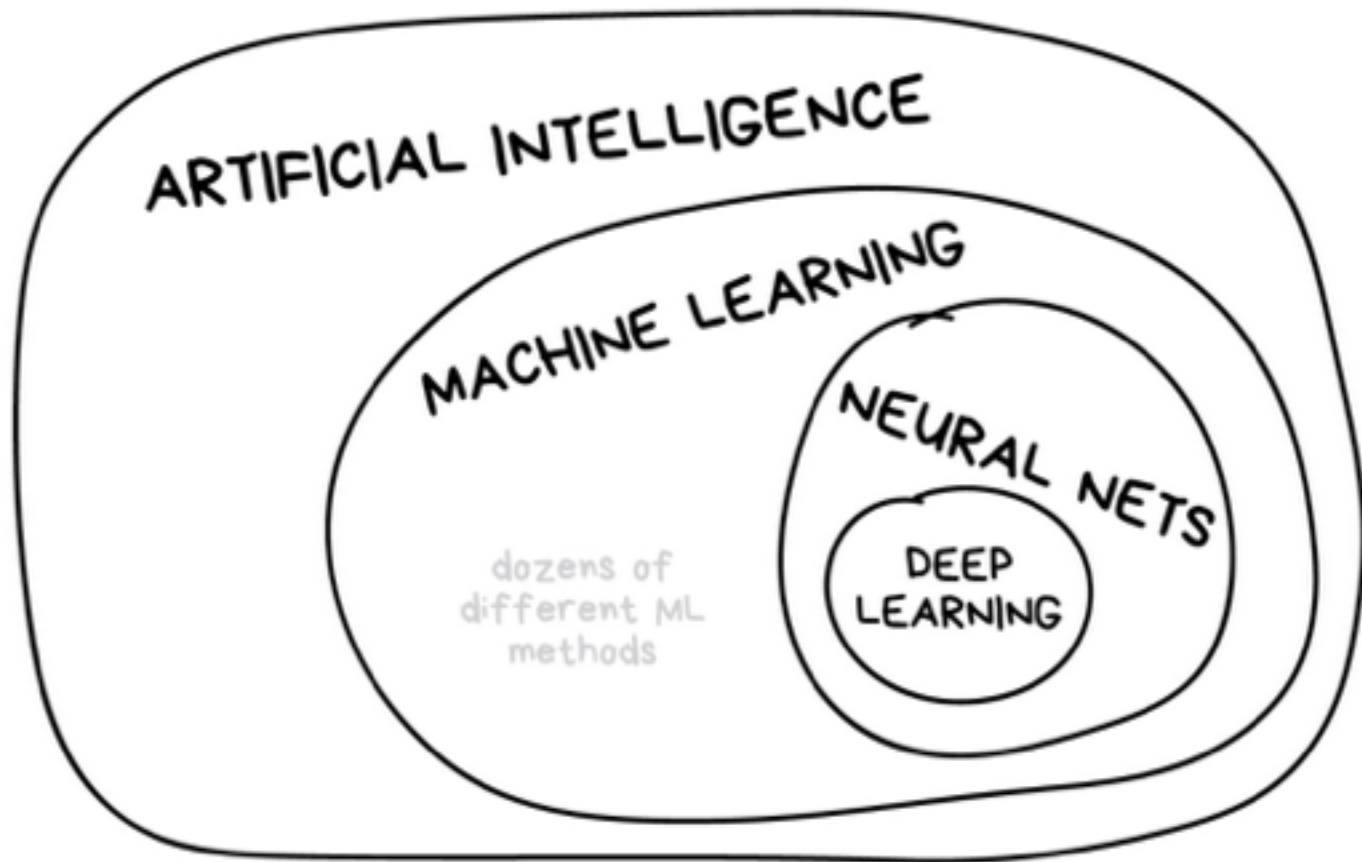
→ predict



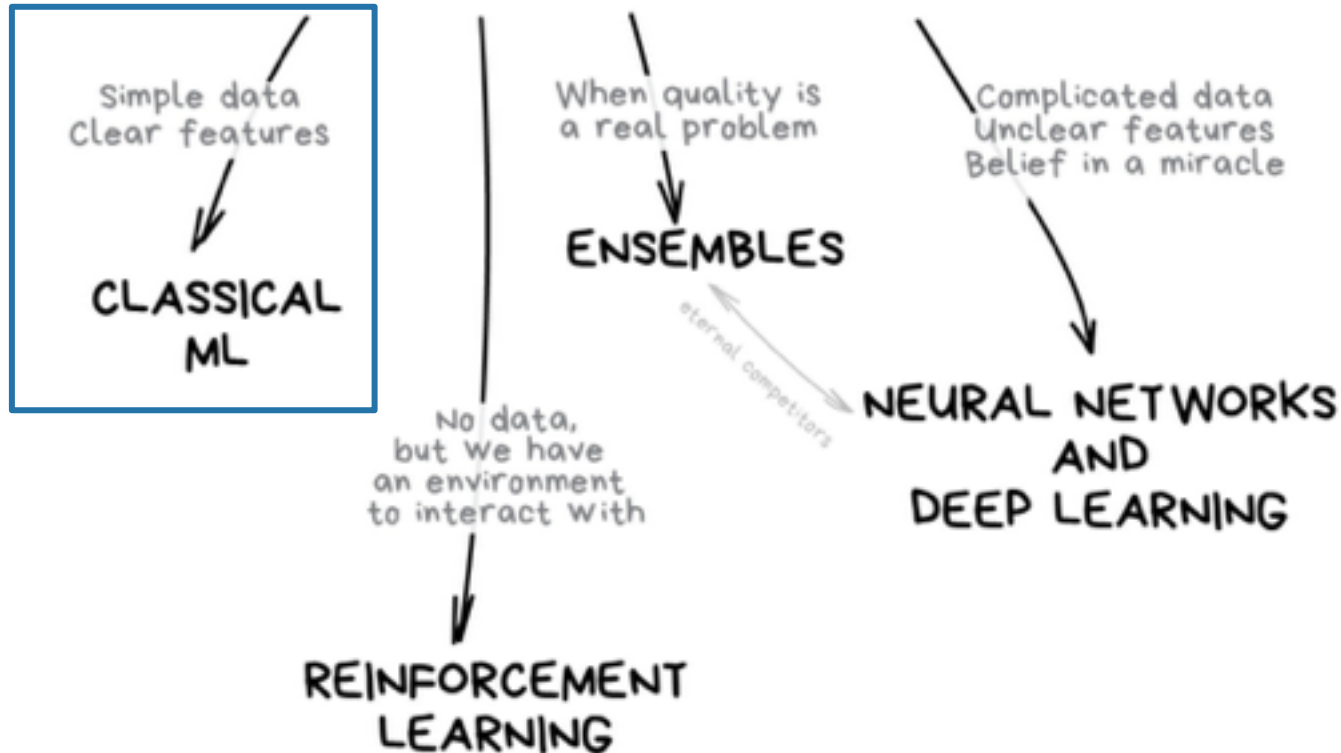
What is machine learning?

“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?

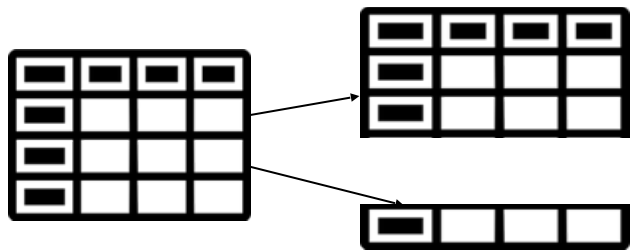
D Can we use information about one's early childhood to predict their success in elementary school?

E How does Sesame Street cause an increase in educational attainment?

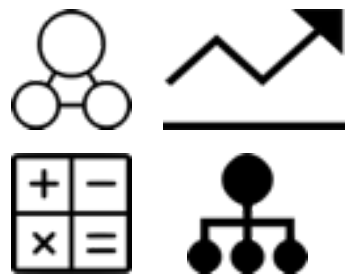


Machine Learning Generalizations

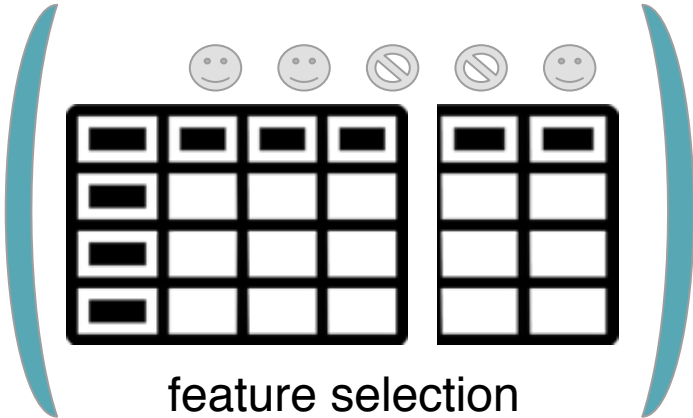
Basic Steps to Prediction



data
partitioning



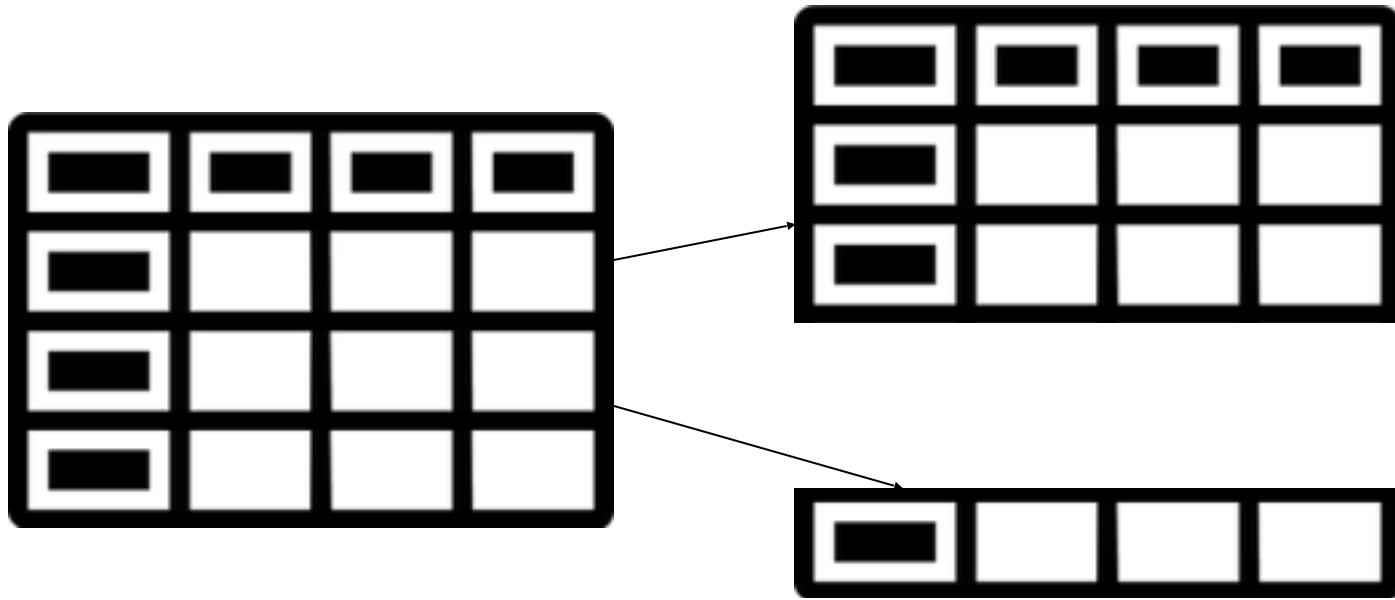
model selection



feature selection



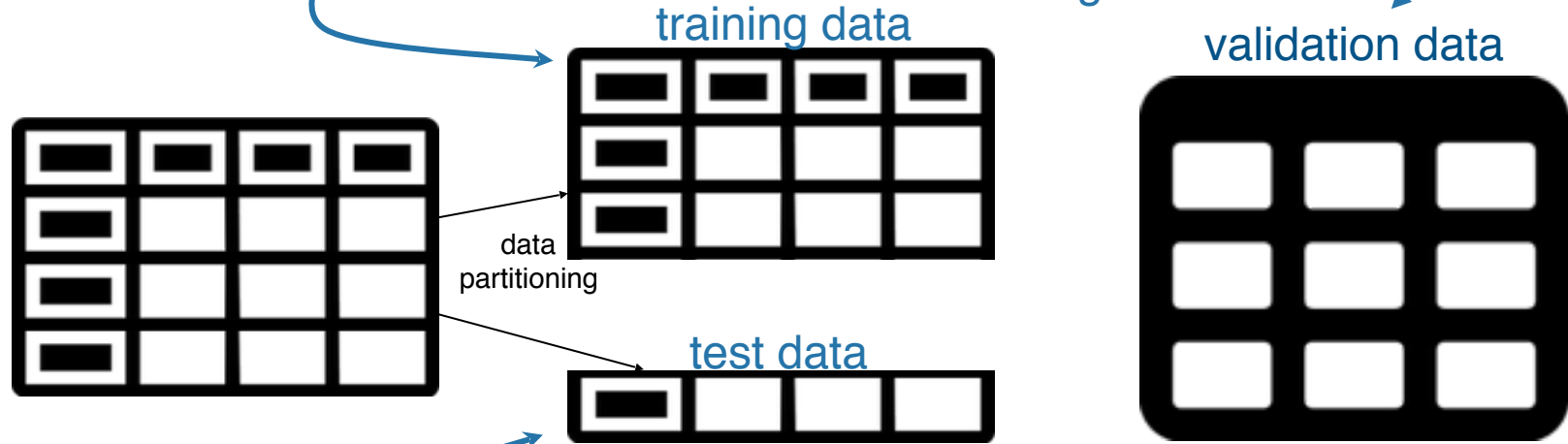
model assessment



data partitioning

the data used to
build your
predictive model

new and independent
data set used to assess if
prediction model is
generalizable



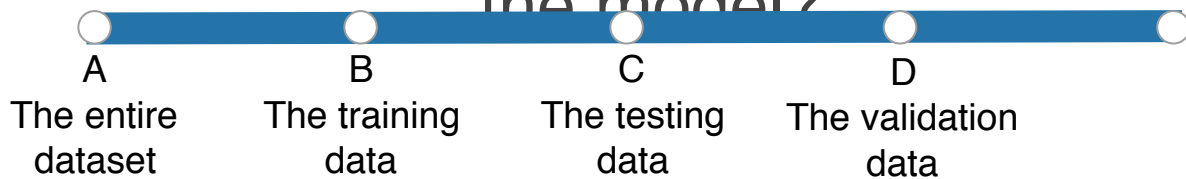
Data from original dataset that was
held out and not used in training the
model ; helpful in fine-tuning

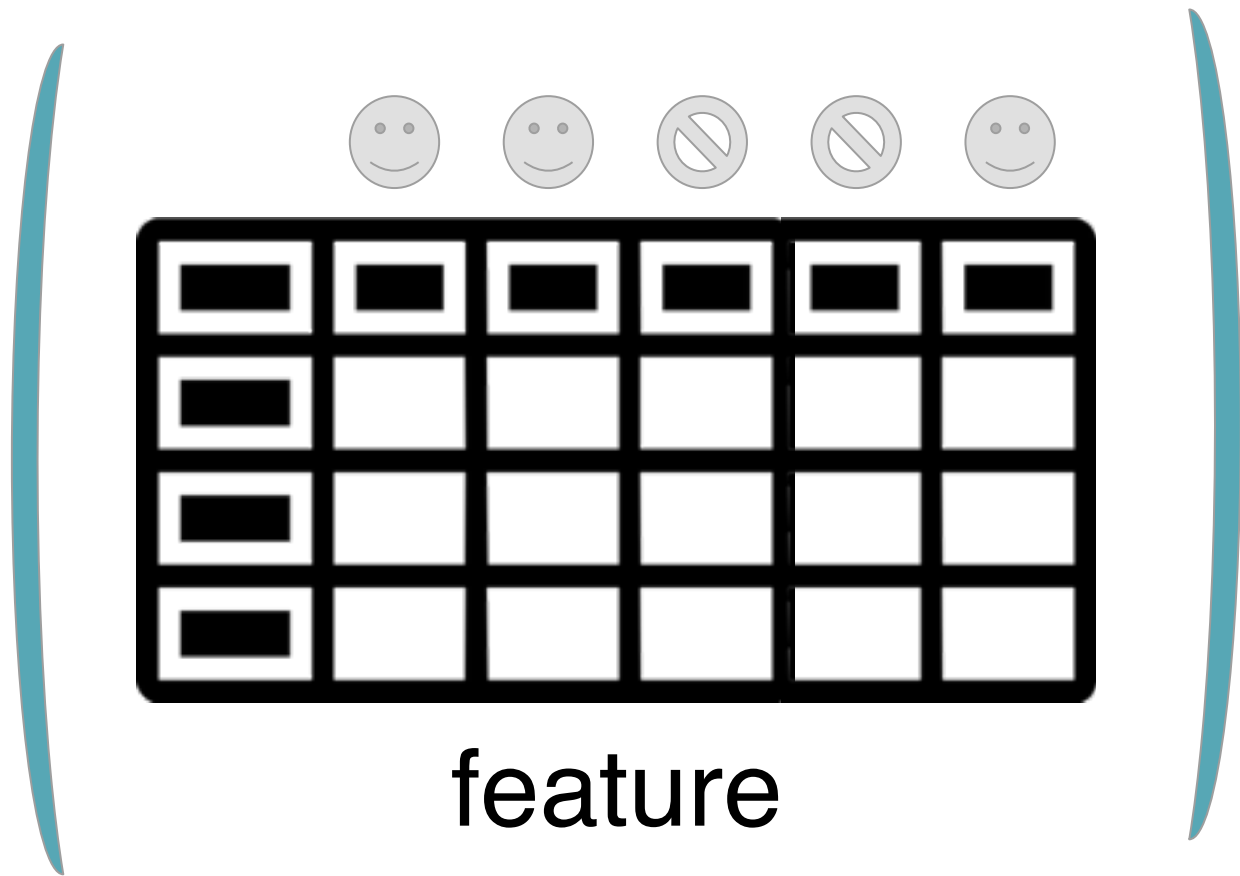
prediction accuracy

Data Partitioning



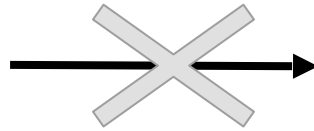
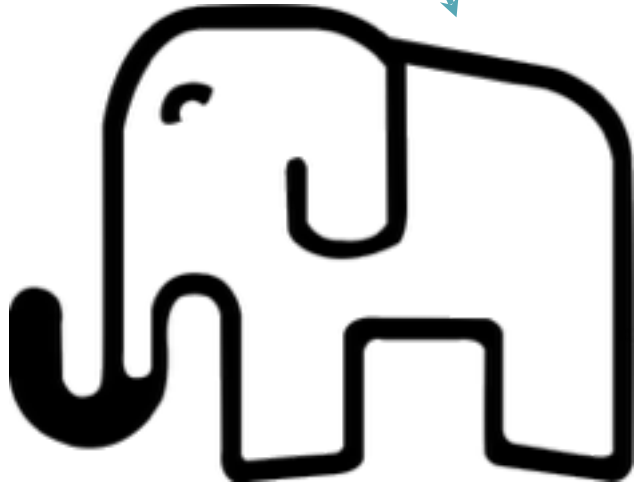
What portion of the data are typically used for generating
the model?

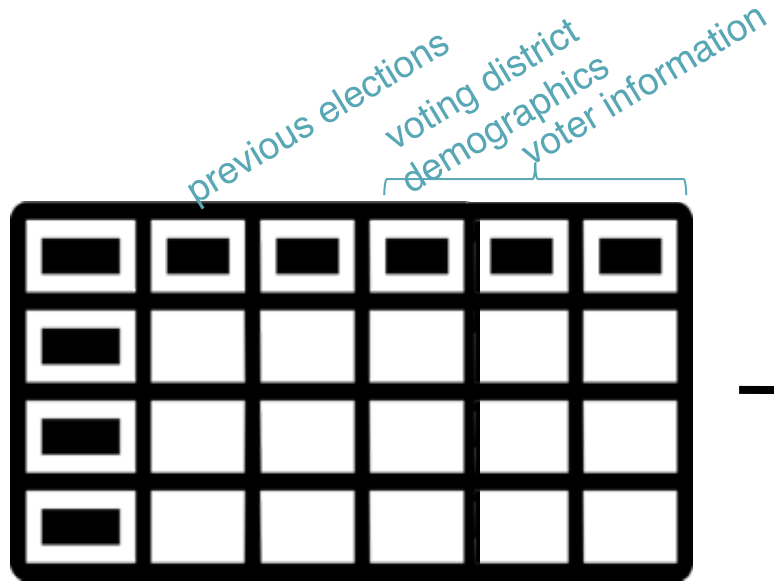




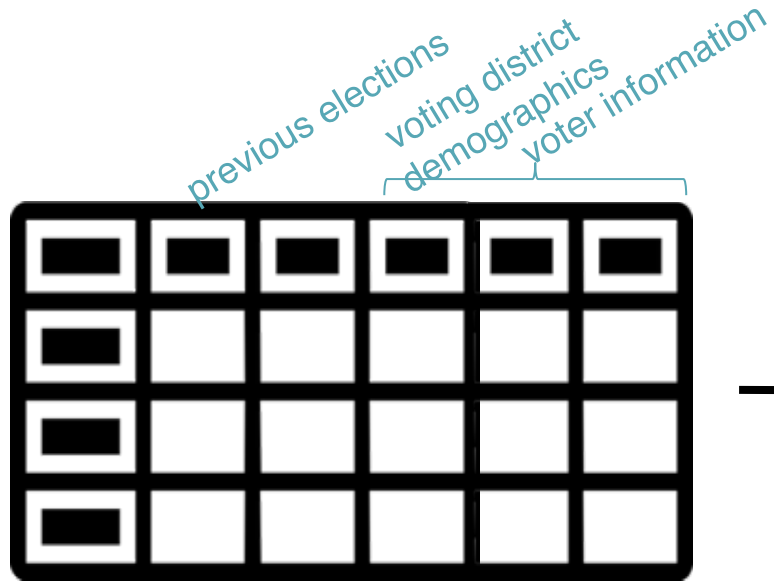
feature
selection

elephant height data
are likely not predictive
of US elections

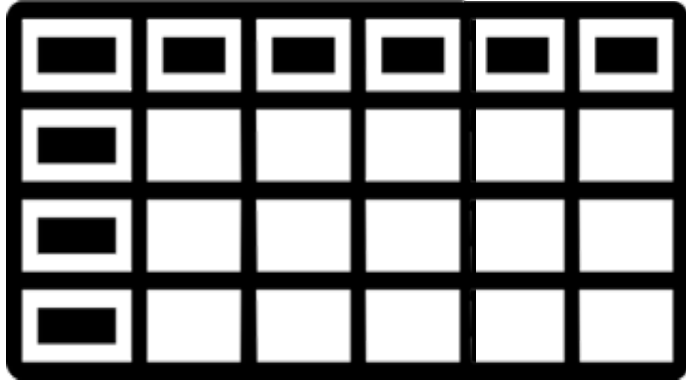




these data are likely
predictive of US
election outcomes



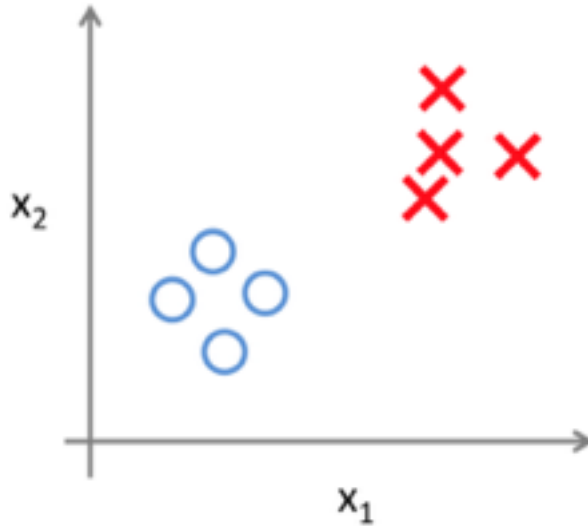
feature selection determines which variables are most predictive and includes them in the model



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

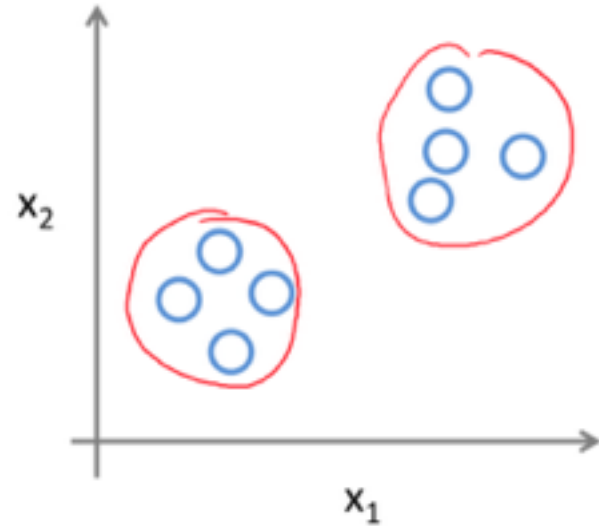
To modes of machine learning

Supervised Learning



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

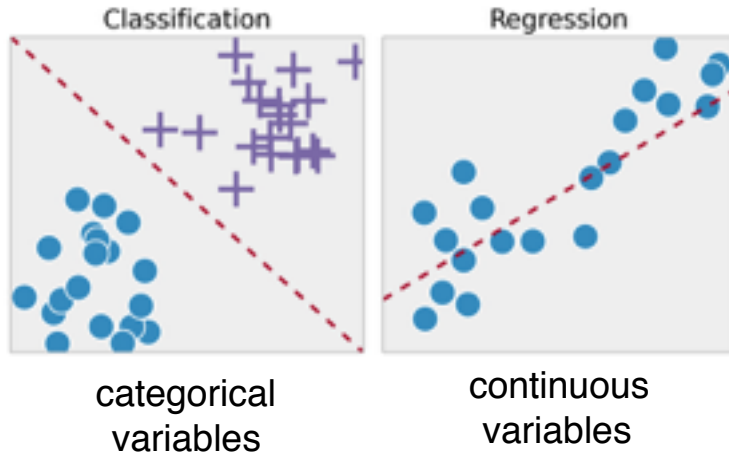
Unsupervised Learning



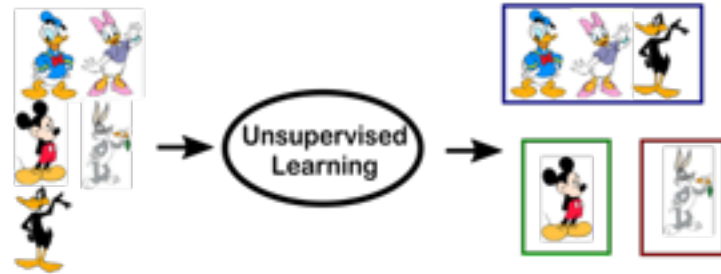
The computer determines how to classify based on properties within the data

Approaches to machine learning

Supervised Learning



Unsupervised Learning



Clustering (categorical)
& dimensionality reduction (continuous)

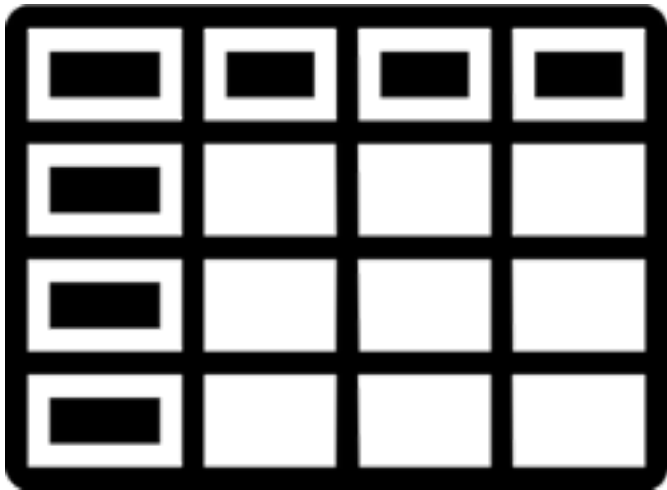
can automatically identify
structure in data



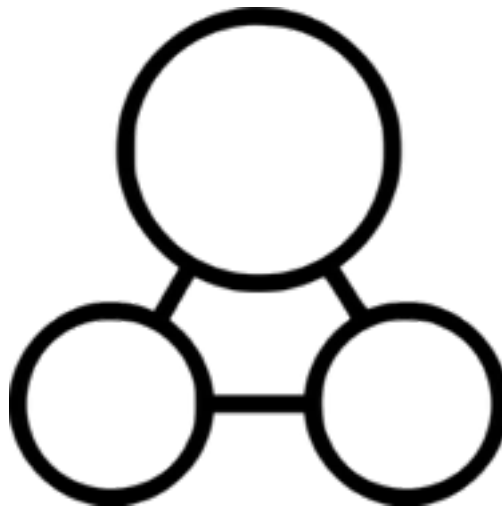
+	-
×	=



model selection

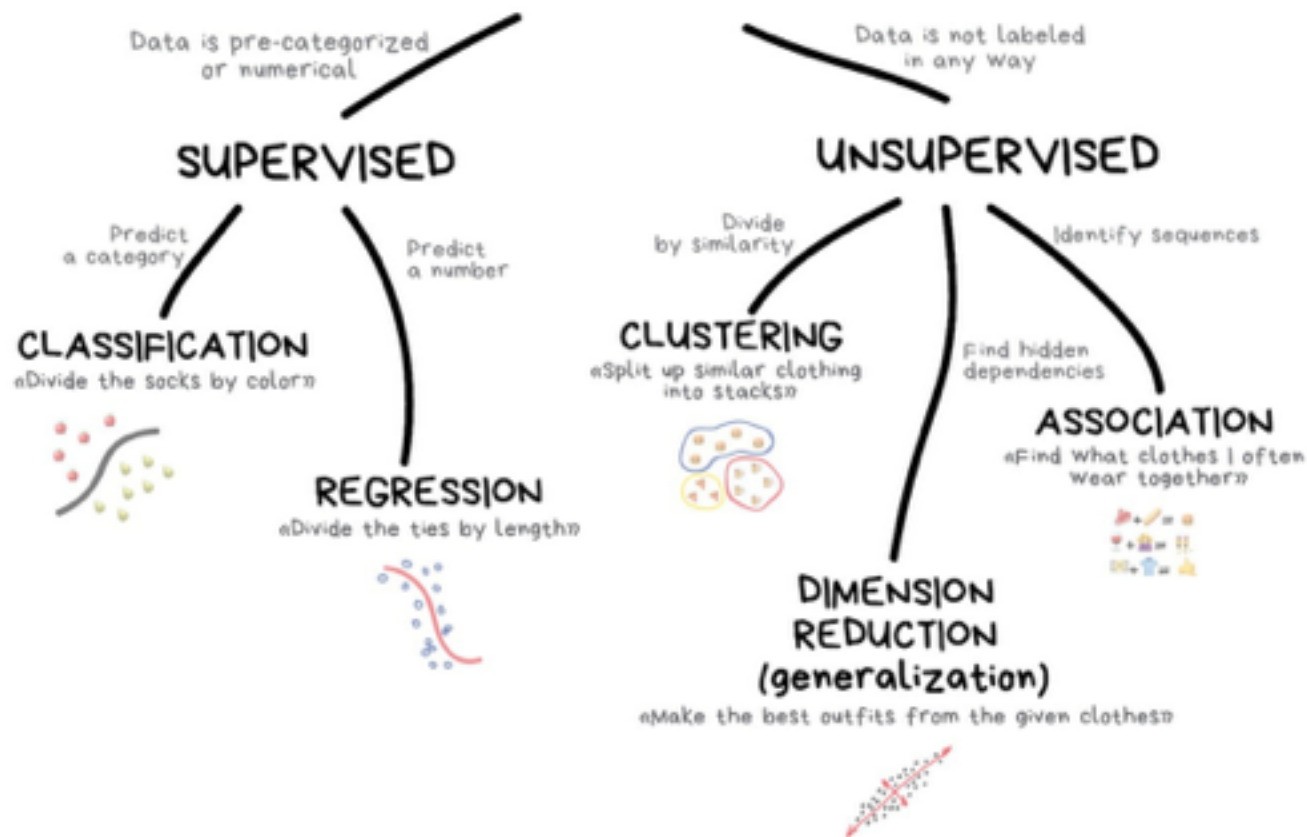


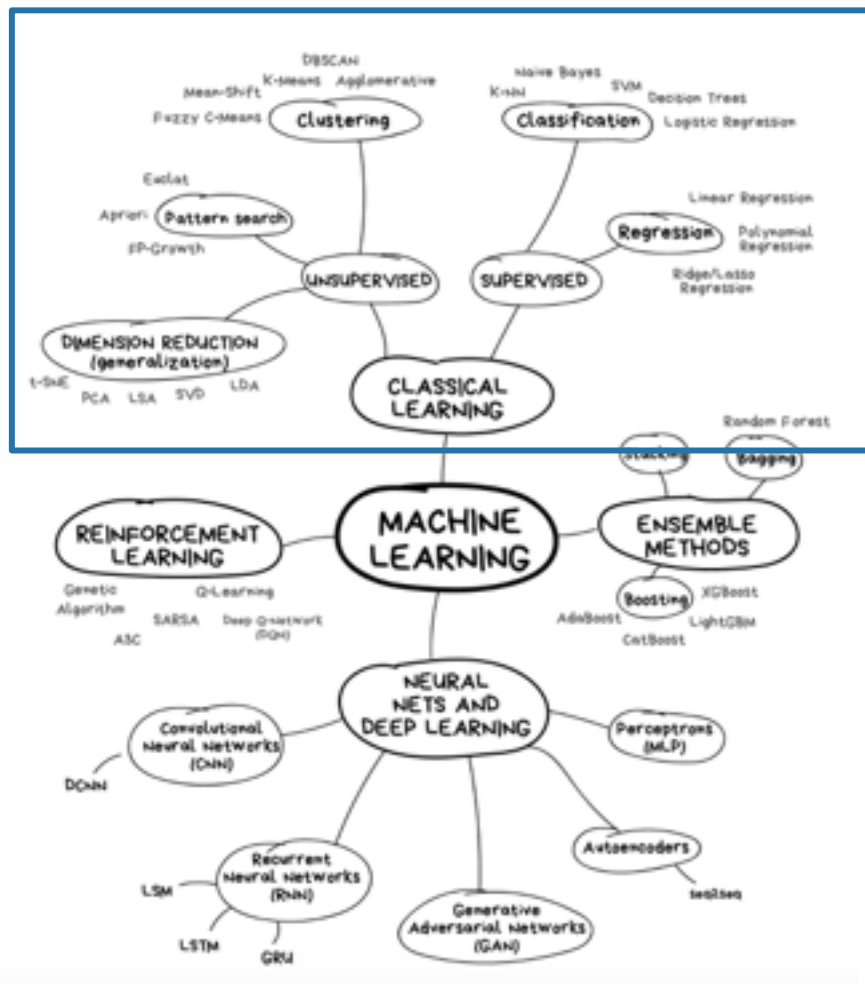
big
datasets



simple
models

CLASSICAL MACHINE LEARNING



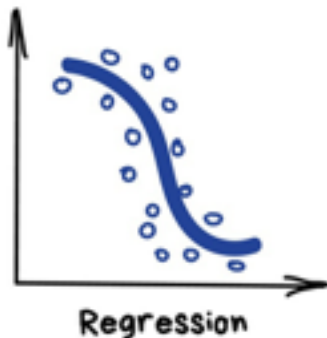


Regression

"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



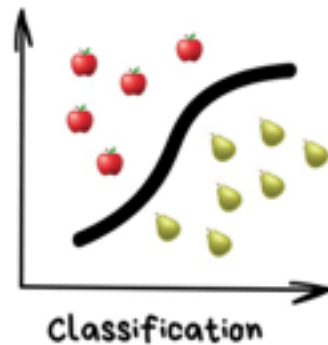
Popular algorithms are Linear and Polynomial regressions.

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



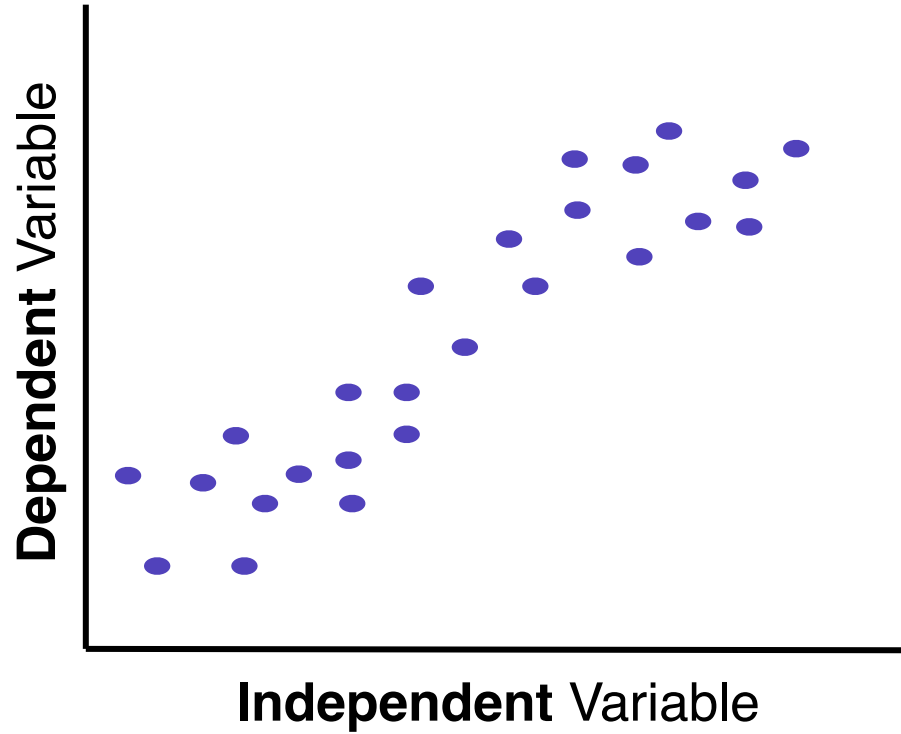
Regression:
predicting continuous
variables
(i.e. Age)

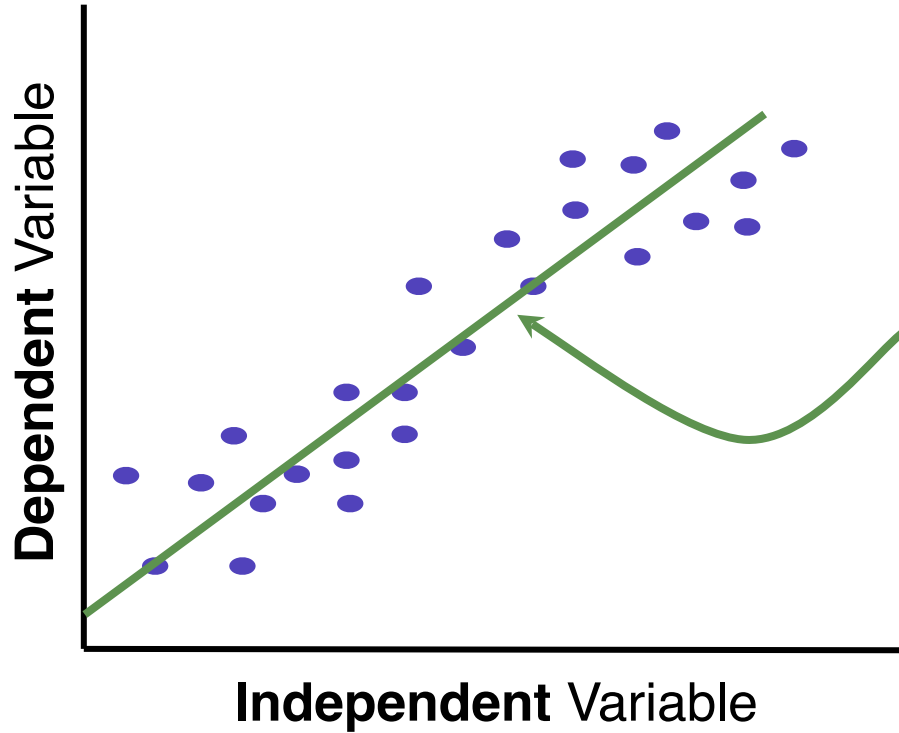
continuous variable
prediction



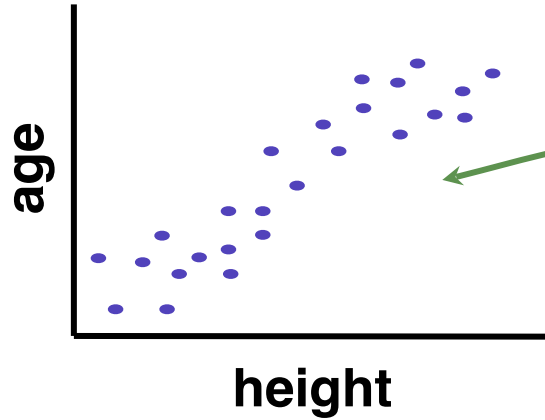
Classification:
predicting categorical
variables
(i.e. education level)

categorical variable
prediction

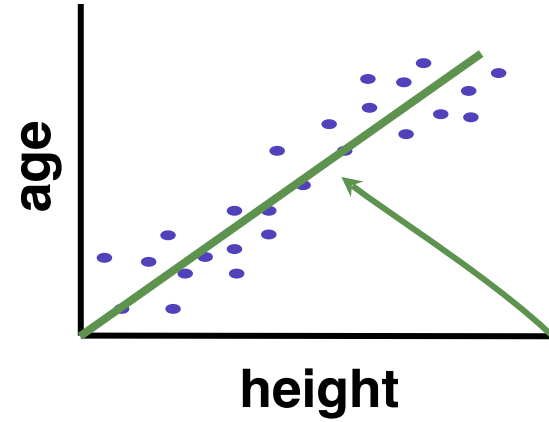
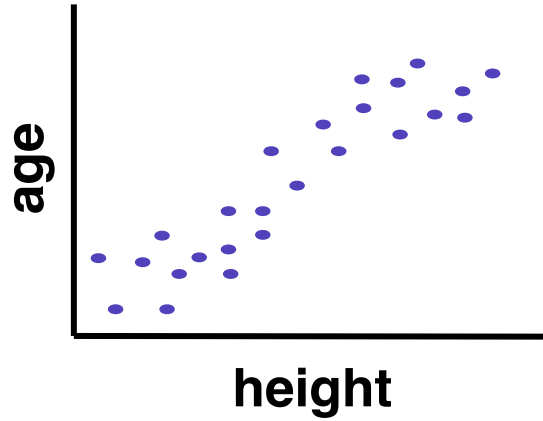




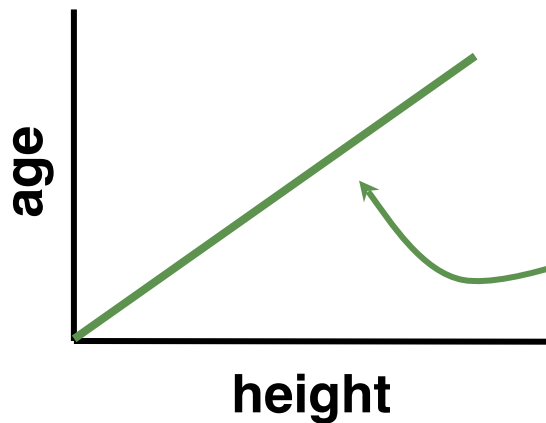
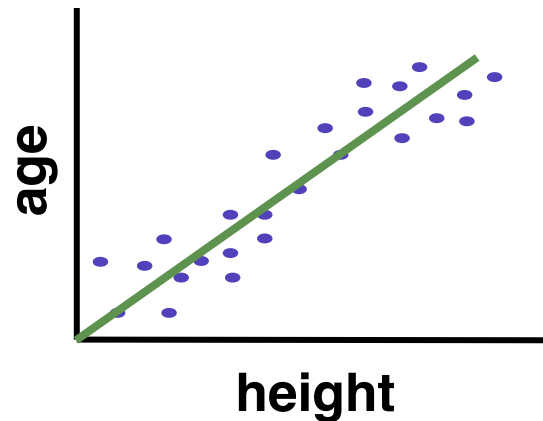
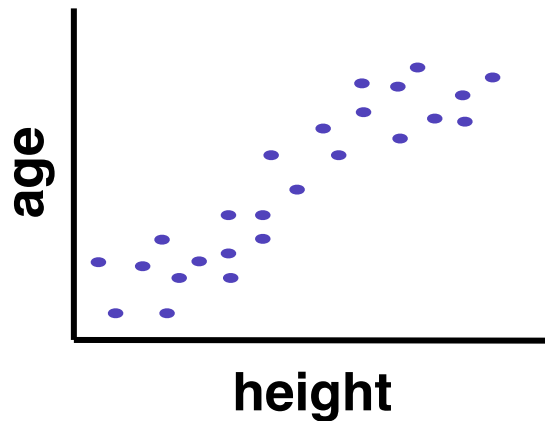
We'll use the
linear relationship
between variables
to generate a
predictive model



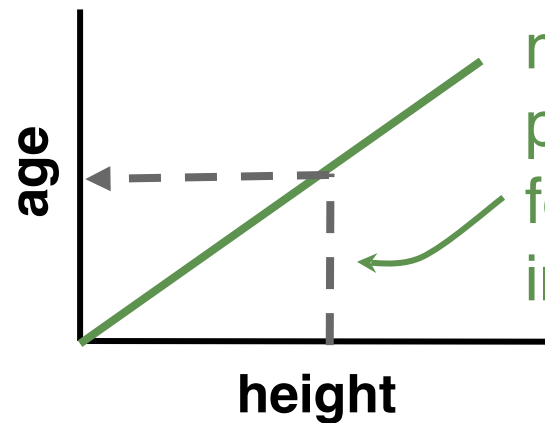
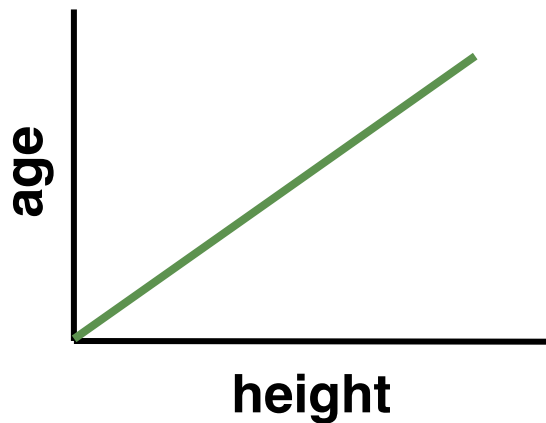
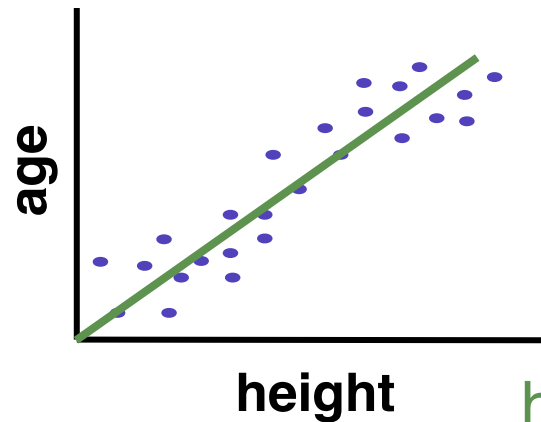
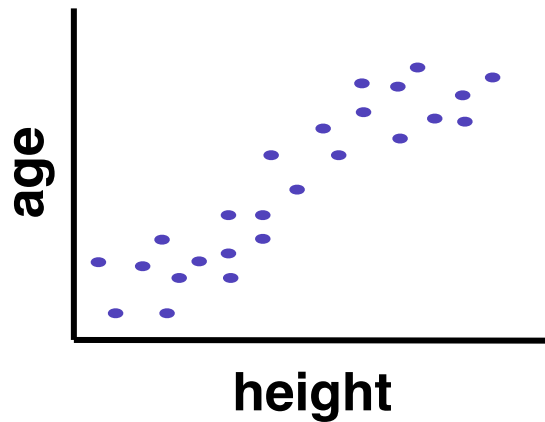
the training data
will be used to
build the
predictive
model



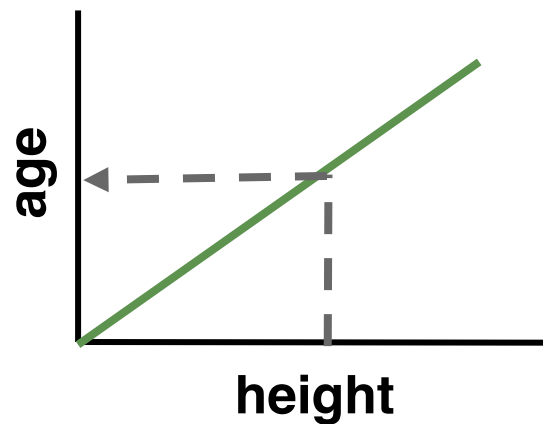
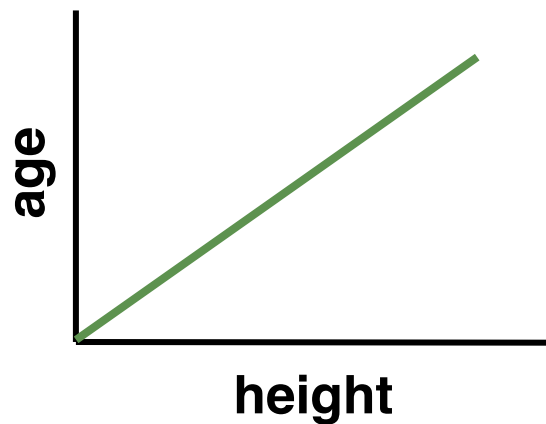
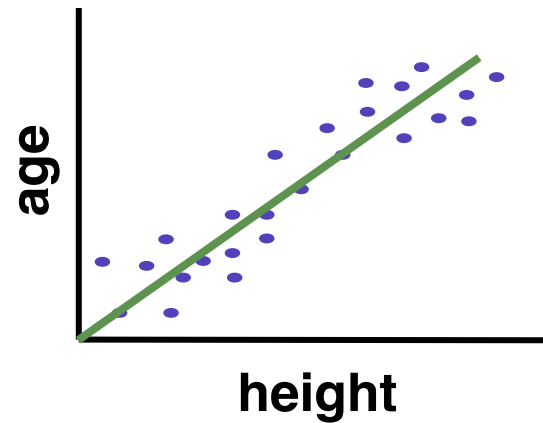
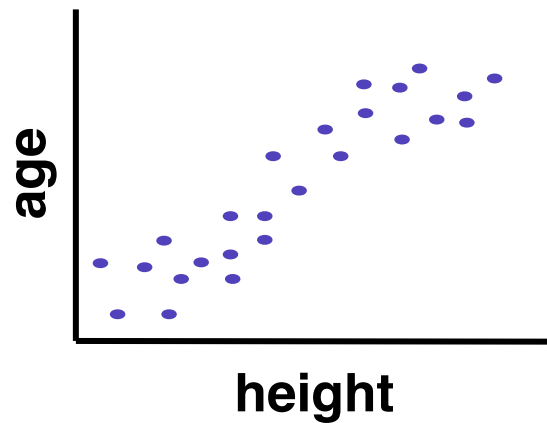
use linear
regression to
model the
relationship



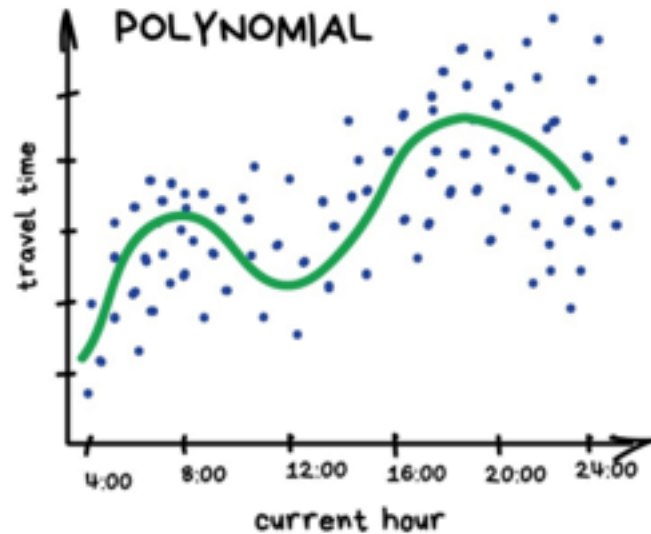
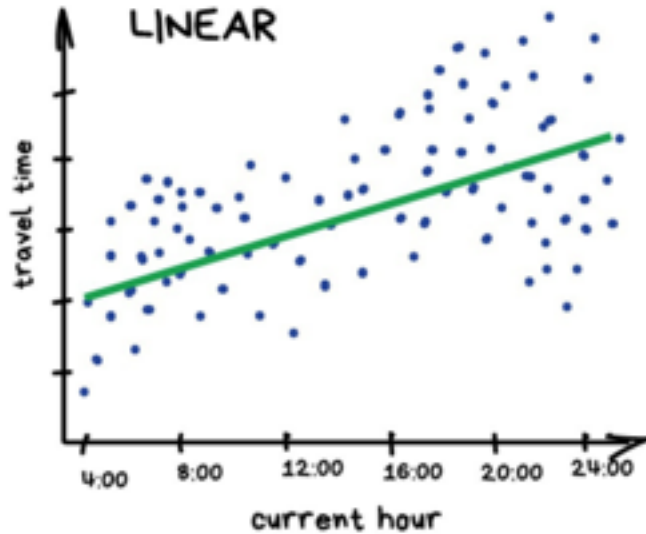
For prediction, the individual values in the training data are *not* important. We only need the model.



how we'll
make
predictions
for a future
individual



PREDICT TRAFFIC JAMS



REGRESSION



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



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

GIVE A LOAN?

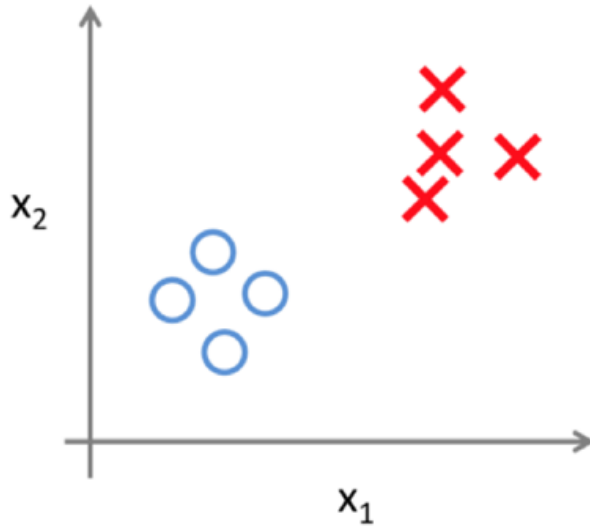


DECISION TREE

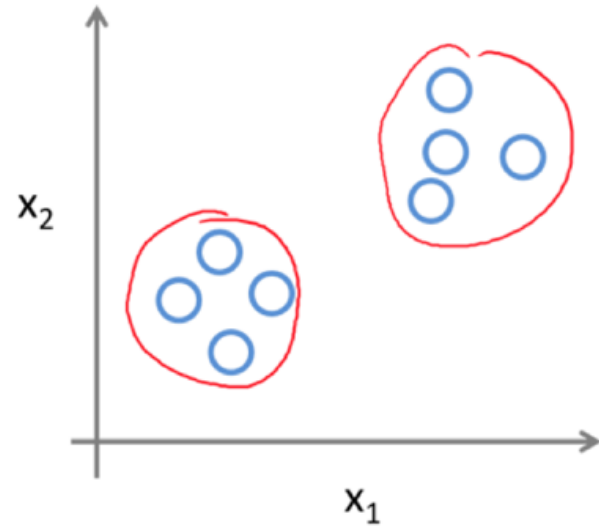
Unsupervised Learning

To modes of machine learning

Supervised Learning



Unsupervised Learning



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

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



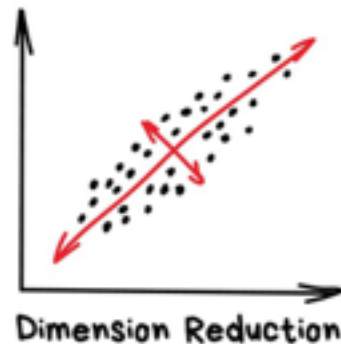
Popular algorithms: K-means clustering, Mean-Shift, DBSCAN

Dimensionality Reduction (Generalization)

"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



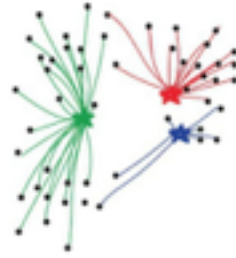
Popular algorithms: Principal Component Analysis (PCA), Singular Value Decomposition (SVD), Latent Dirichlet allocation (LDA), Latent Semantic Analysis (LSA, pLSA, GLSA), t-SNE (for visualization)

PUT KEBAB KIOSKS IN THE OPTIMAL WAY

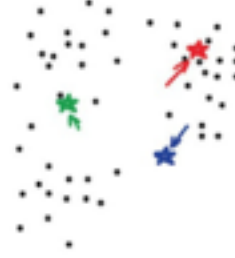
(also illustrating the K-means method)



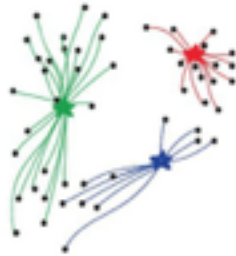
1. Put kebab kiosks in random places in city



2. Watch how buyers choose the nearest one



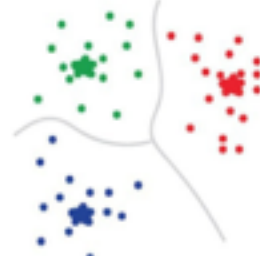
3. Move kiosks closer to the centers of their popularity



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?

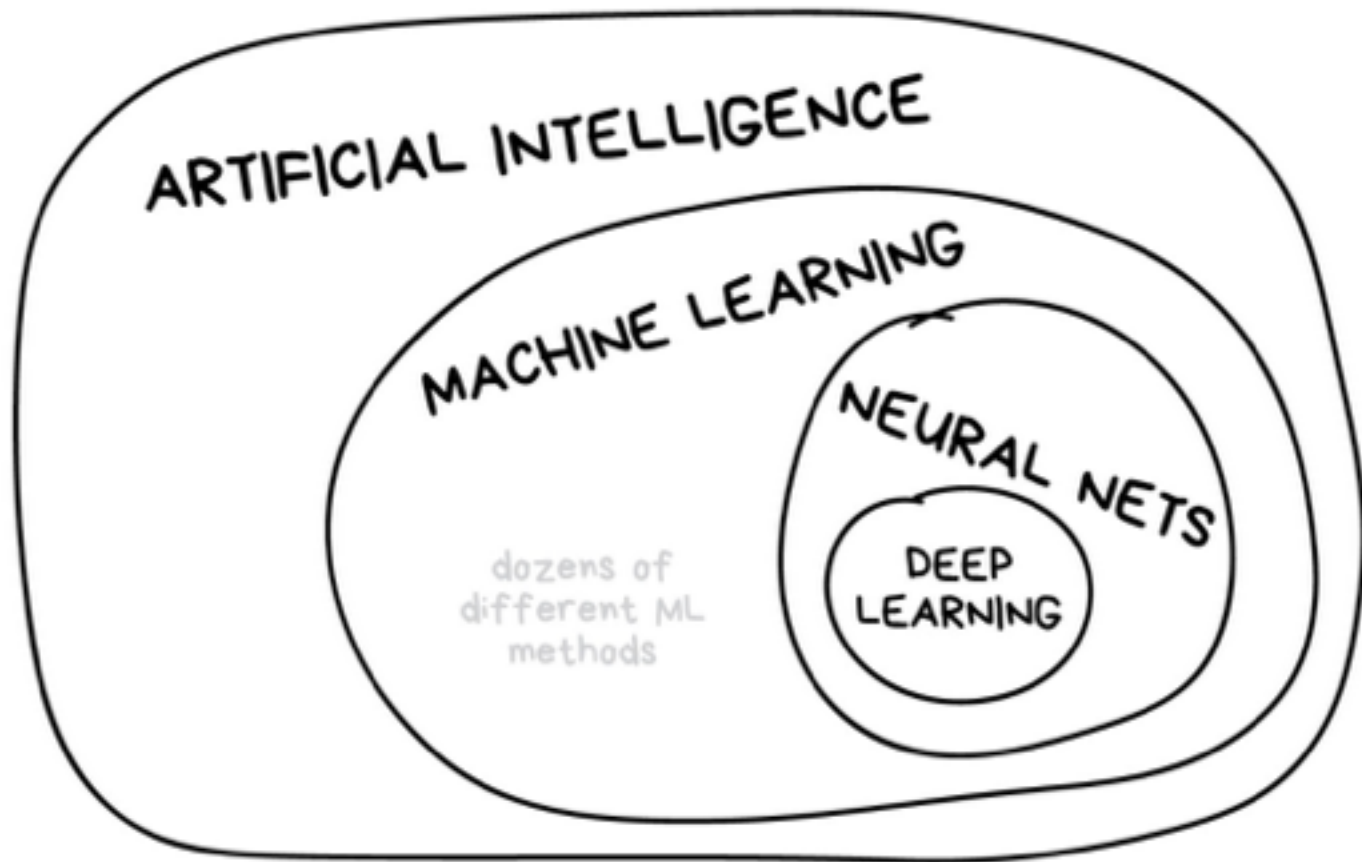
A
Supervised,
Regression

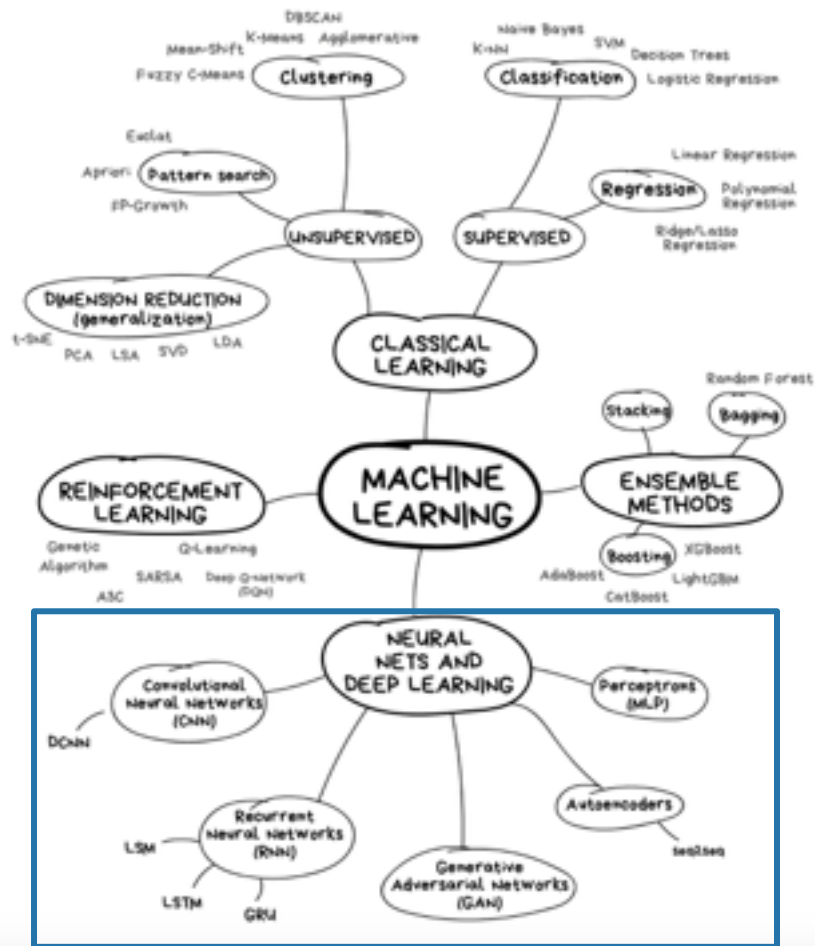
B
Supervised,
Classification

C
Unsupervised
,
dimensionality
reduction

D
Unsupervised,
clustering

E
Unsupervised,
Neural
Network

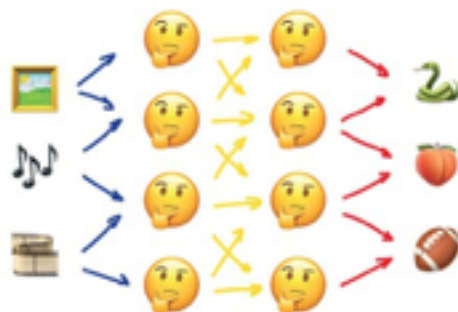




"We have a thousand-layer network, dozens of video cards, but still no idea where to use it. Let's generate cat pics!"

Used today for:

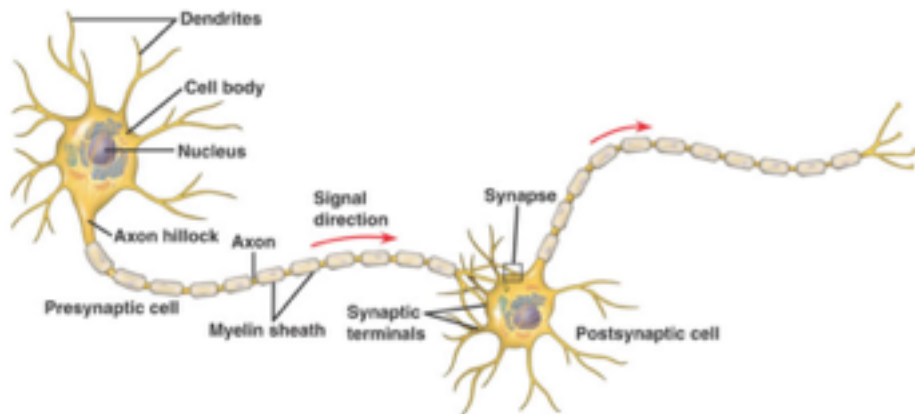
- Replacement of all algorithms above
- Object identification on photos and videos
- Speech recognition and synthesis
- Image processing, style transfer
- Machine translation



Neural Networks

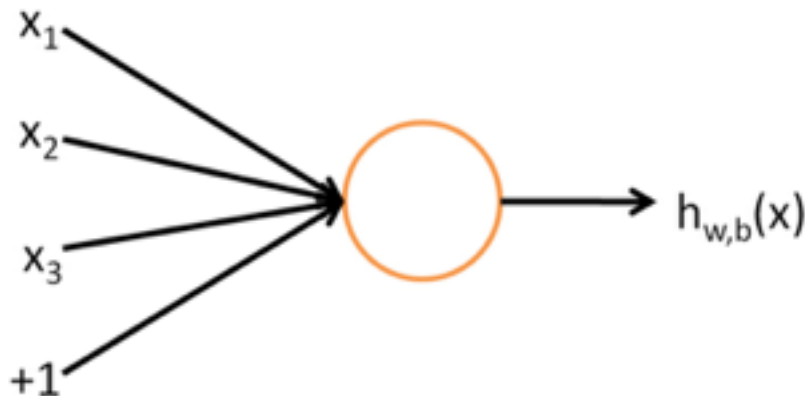
Popular architectures: Perceptron, Convolutional Network (CNN), Recurrent Networks (RNN), Autoencoders

WHAT IS A NEURON?



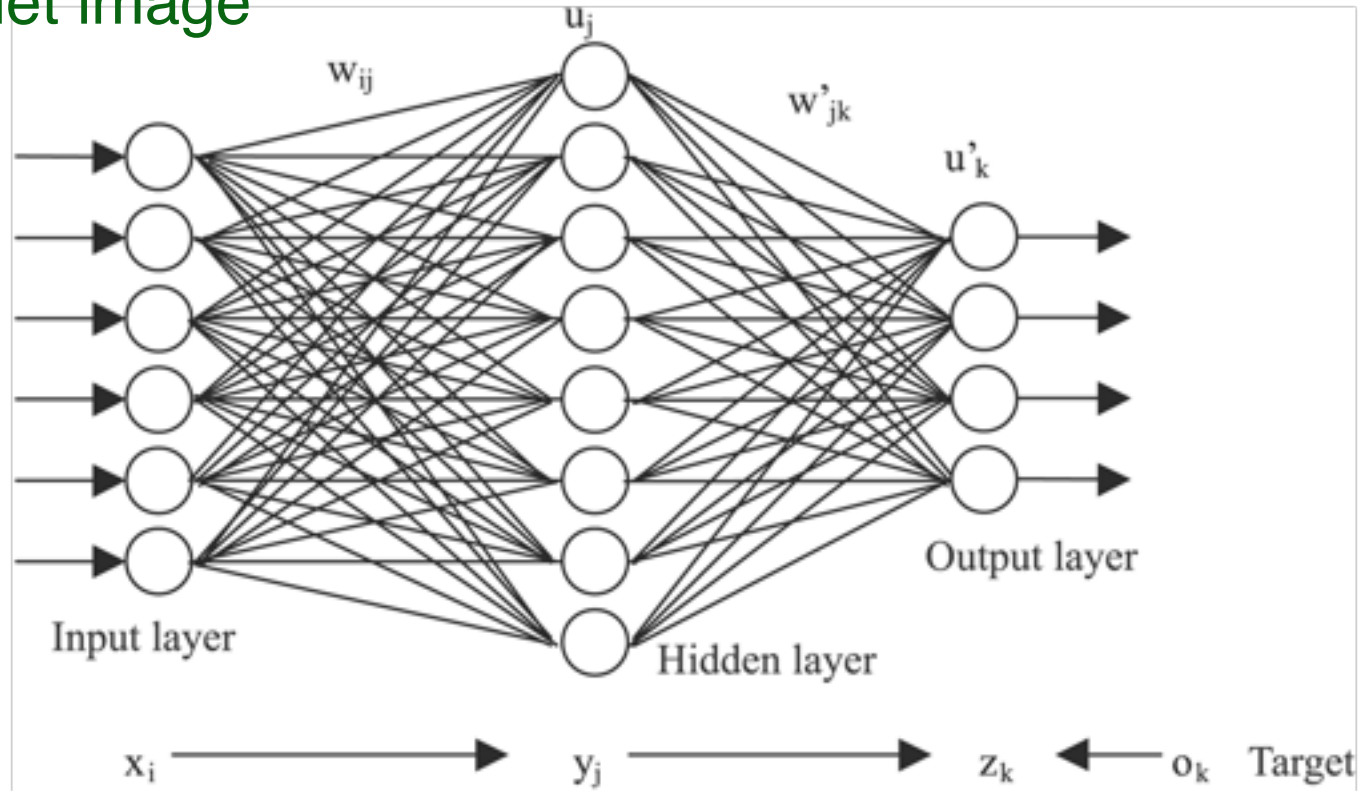
- Receives signal on synapse
- When trigger sends signal on axon

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

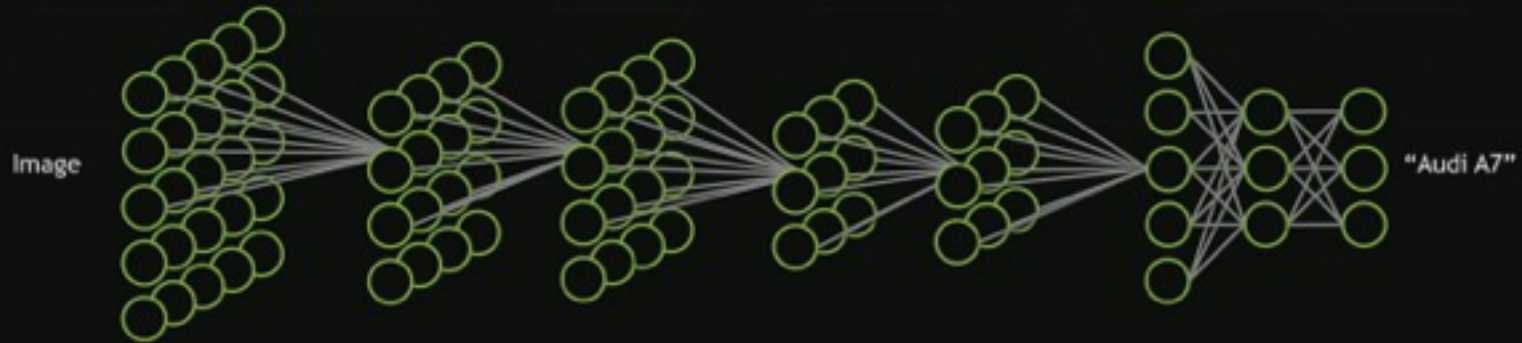
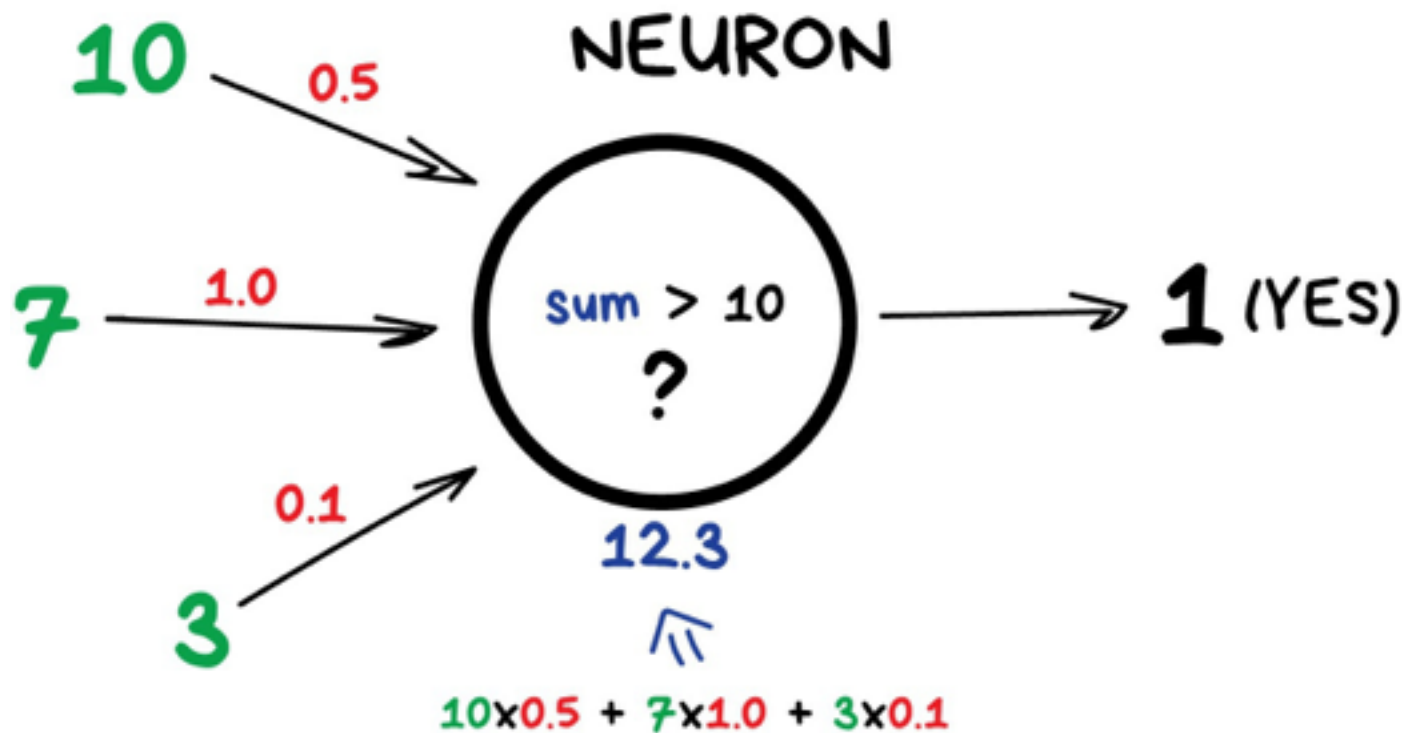
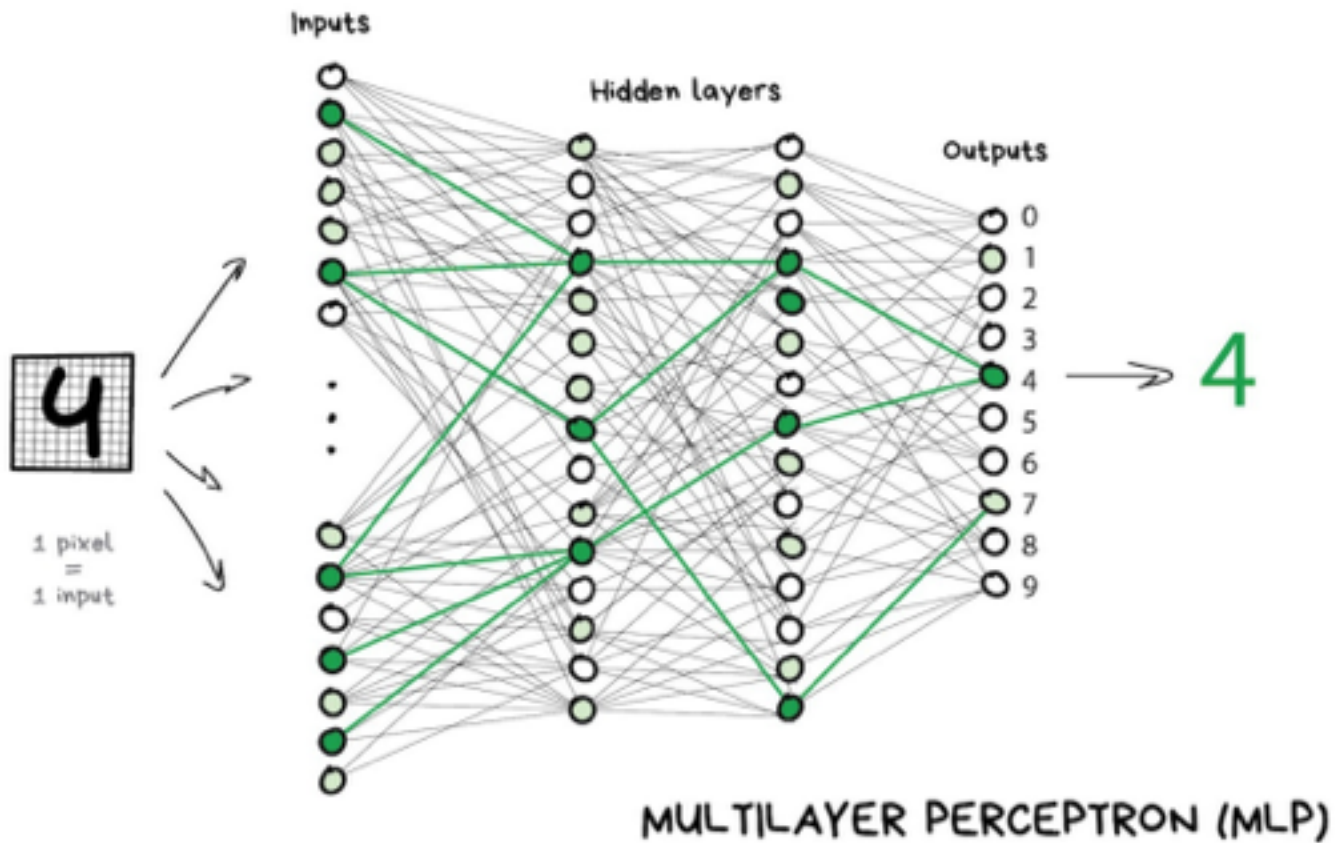


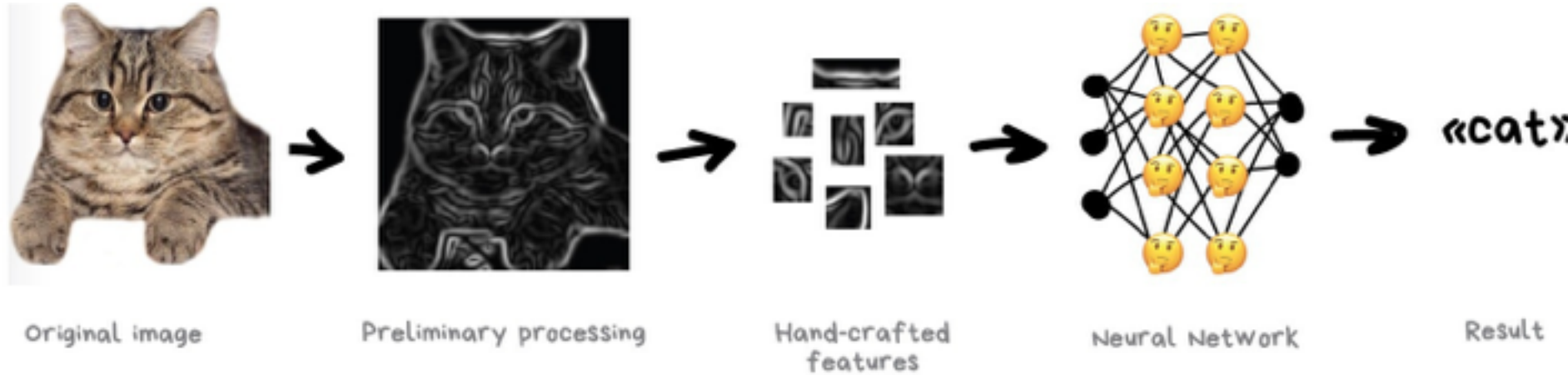
Image source: "Unsupervised Learning of Hierarchical Representations with Convolutional Deep Belief Networks" (ICML 2009 & Comm. ACM 2011, Honglak Lee, Roger Grasse, Rajesh Ranganath, and Andrew Ng.

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.

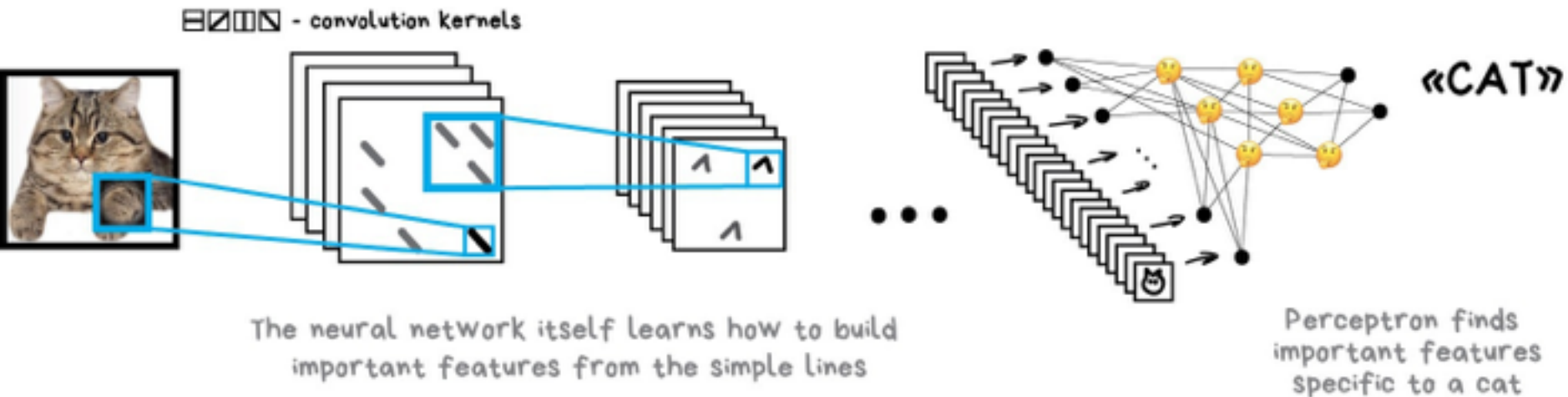




Manually labeling used to be the way...



CNNs avoid manual labeling



“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)

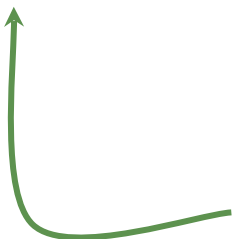




model assessment

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

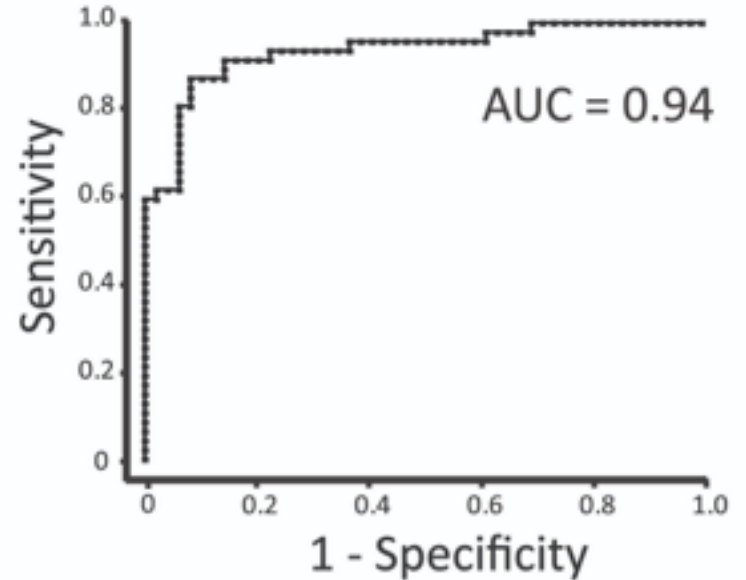
prediction

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

		Actual	
		Positive	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

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



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

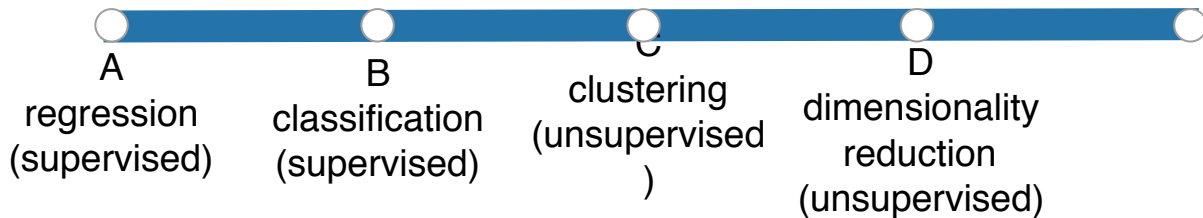
Accuracy	What % were predicted correctly?
Sensitivity	Of those that <i>were</i> positives , what % were predicted to be positive?
Specificity	Of those that were <i>negatives</i> , what % were predicted to be negative?

Prediction Approach



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

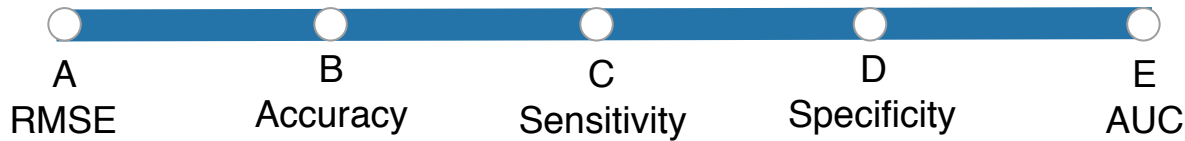
What prediction approach would you use?



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?

