

# Course Announcements

Due today (Friday; 11:59 PM):

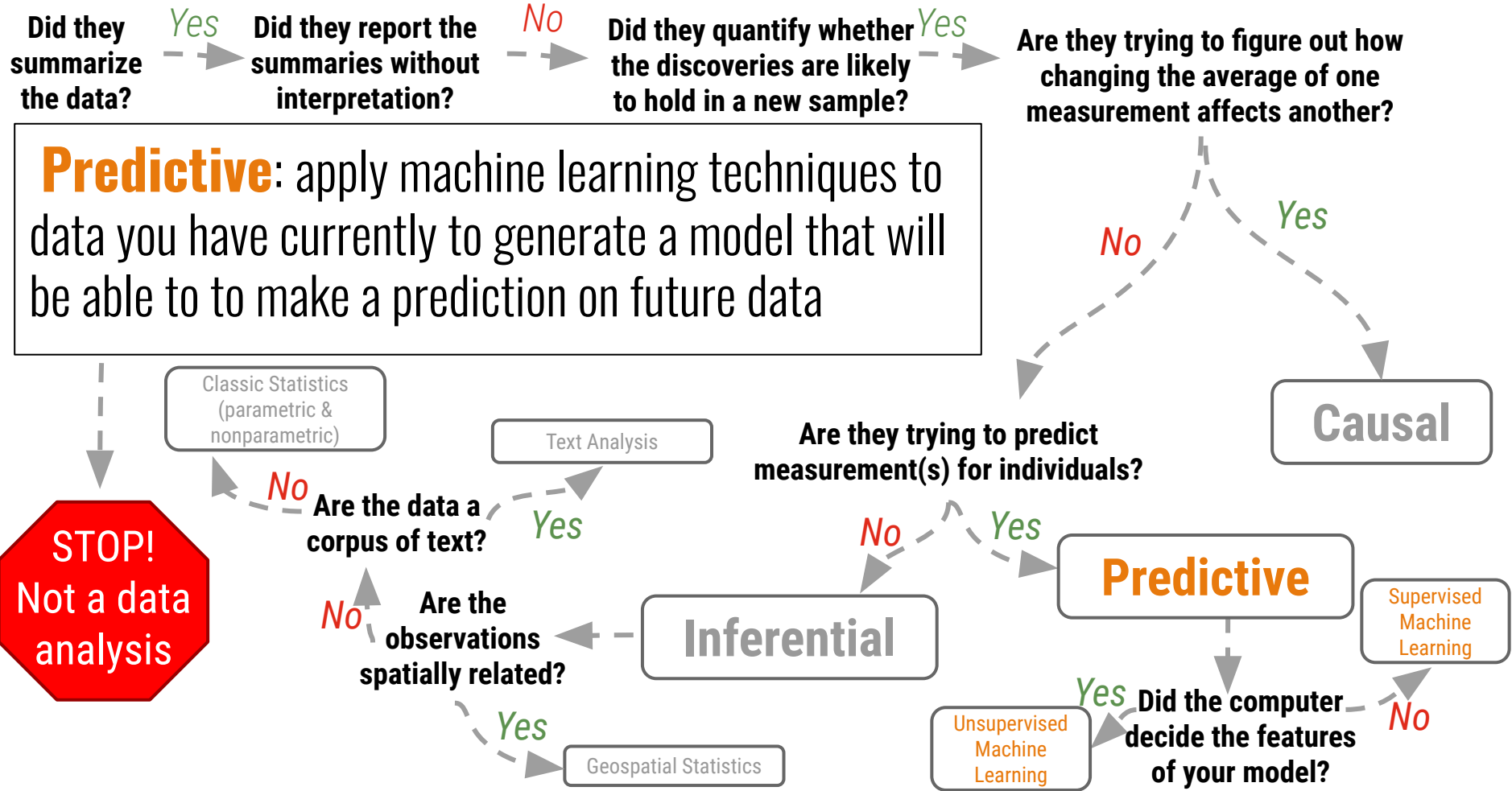
- D6
- Q6
- A3
- Weekly Project Survey

# Machine Learning

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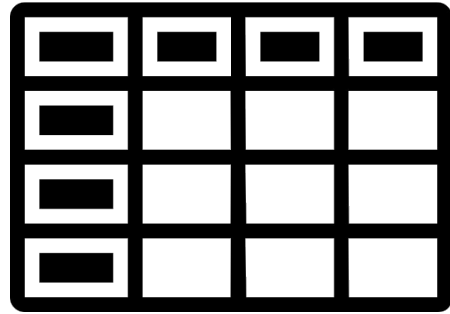


- **Problem:** Detecting whether credit card charges are fraudulent.
- **Data science question:** Can we use the time of the charge, the location of the charge, and the price of the charge to predict whether that charge is fraudulent or not?
- **Type of analysis:** Predictive analysis



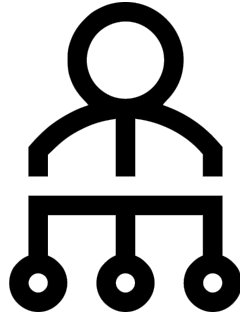
**predictive analysis** uses data  
you have now to make  
predictions in the future

**machine learning**  
approaches are used for  
predictive analysis!



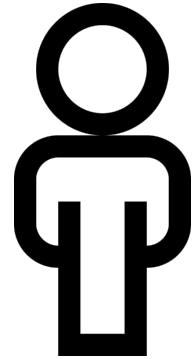
data

train →



model

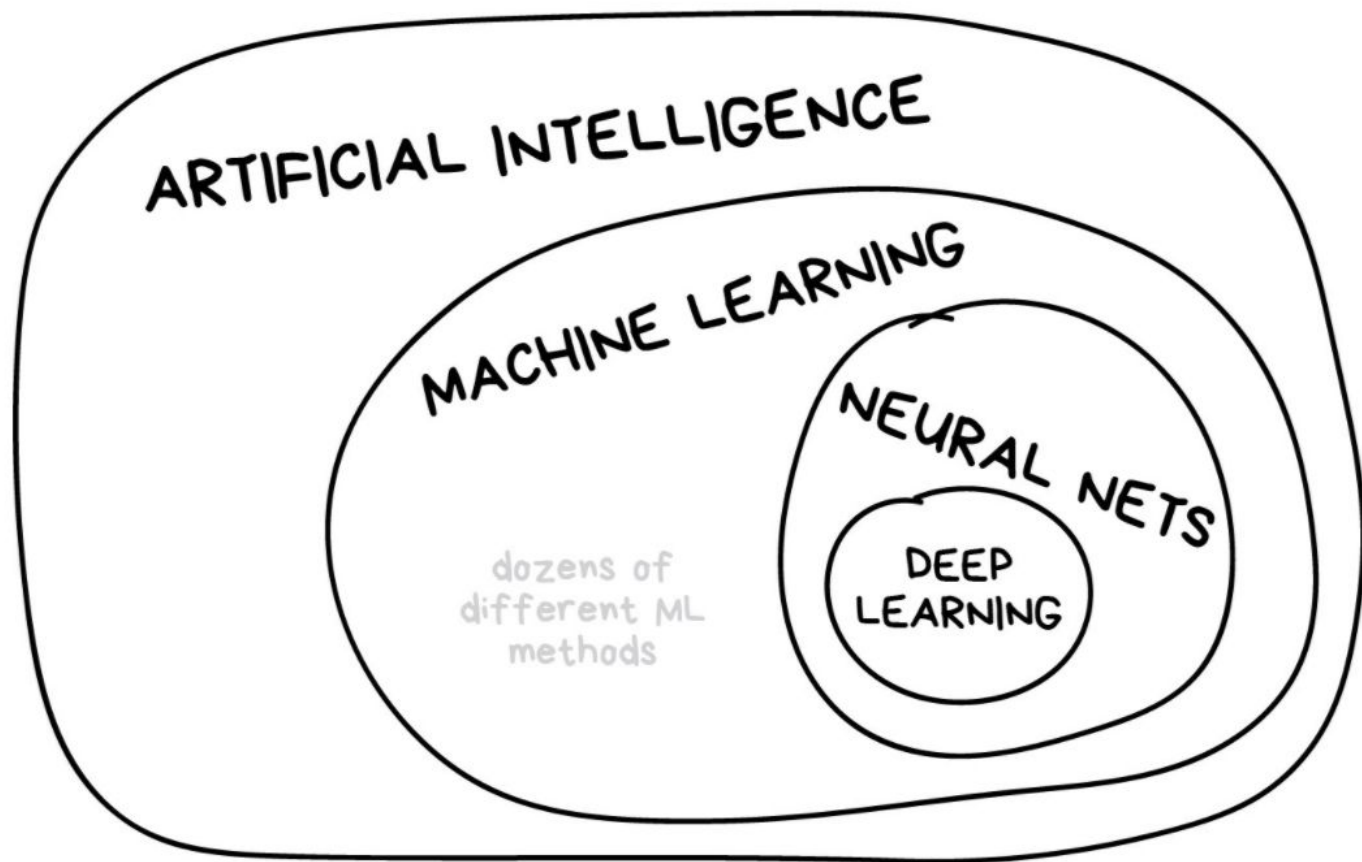
→ predict



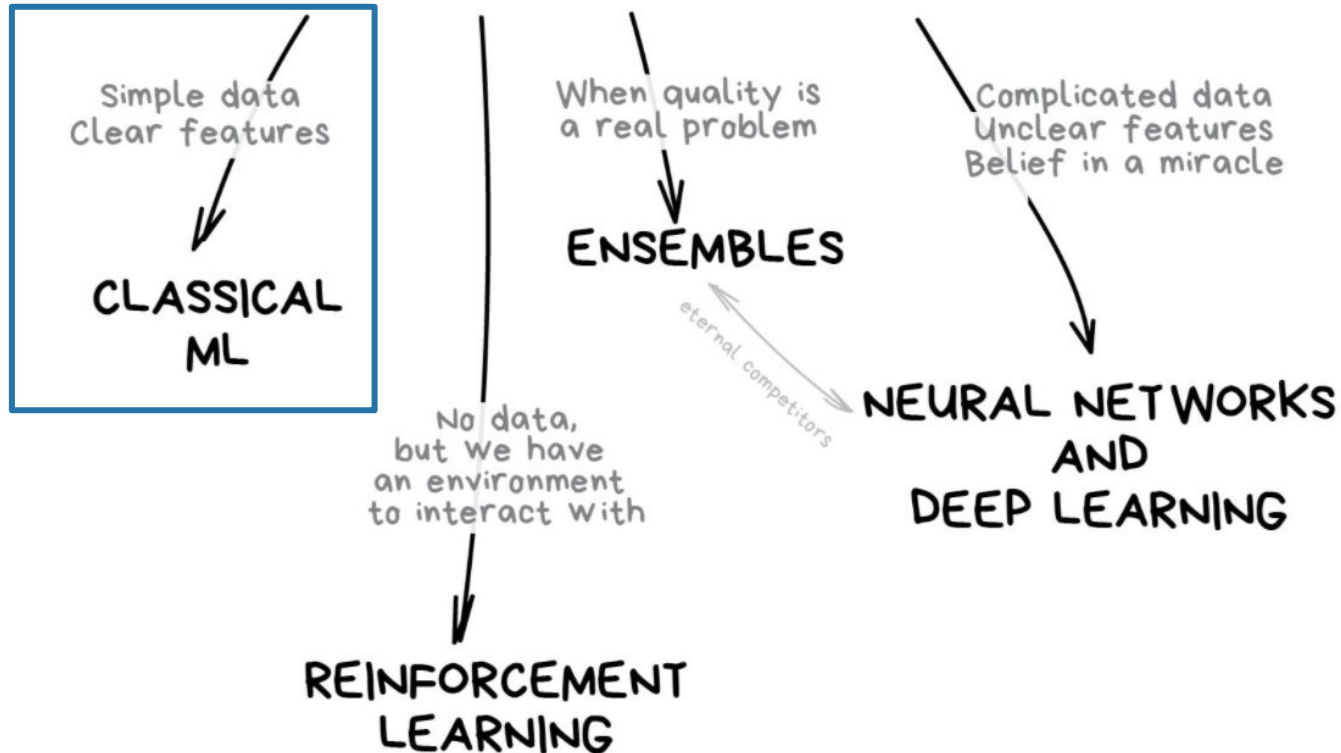
# What is machine learning?

“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?

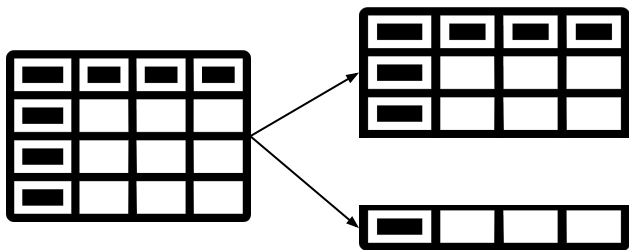
**E** How does Sesame Street cause an increase in educational attainment?



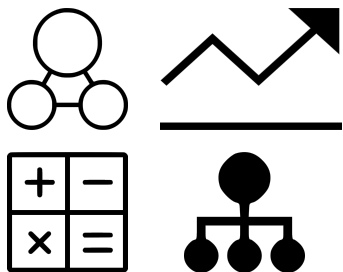
# Machine Learning Generalizations

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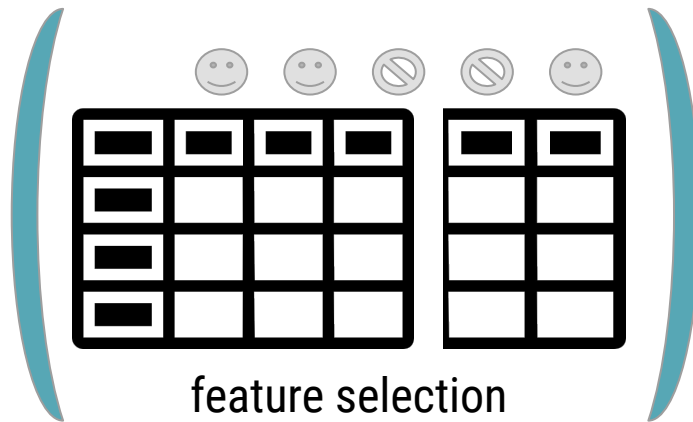
# Basic Steps to Prediction



data  
partitioning



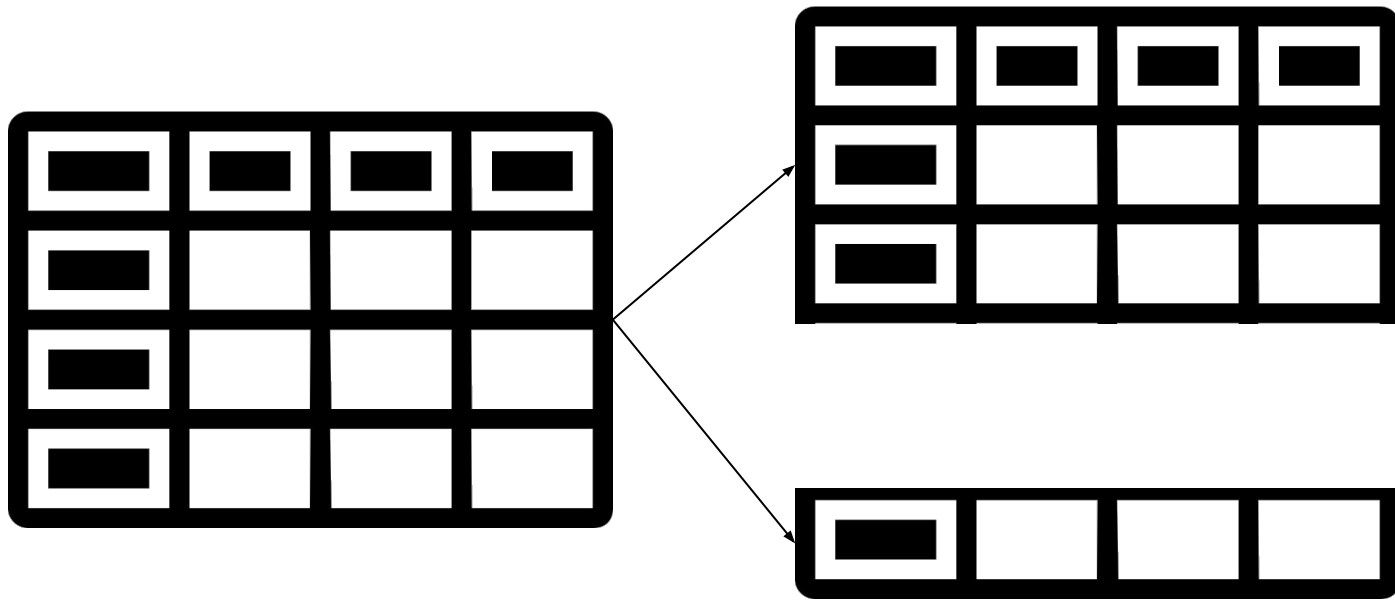
model selection



feature selection



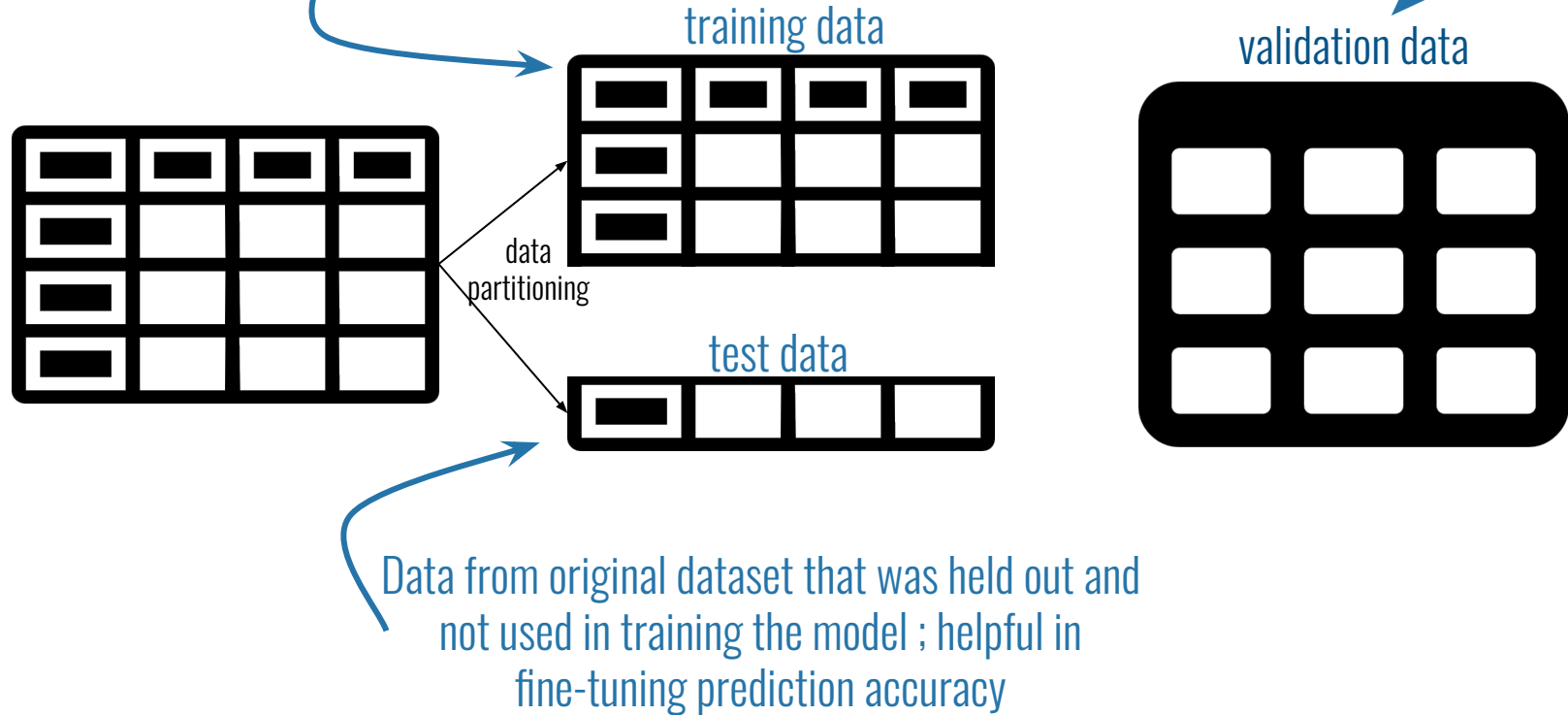
model assessment



data partitioning

the data used to build  
your predictive model

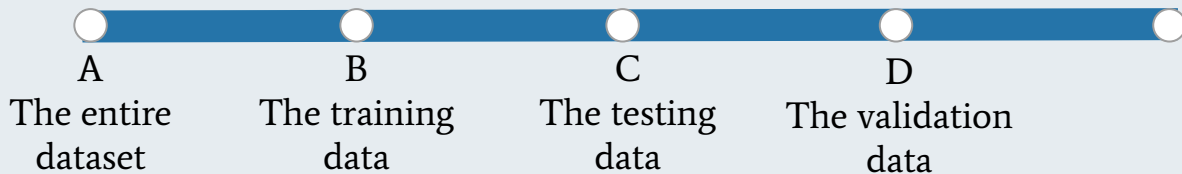
new and independent data set  
used to assess if prediction model  
is generalizable

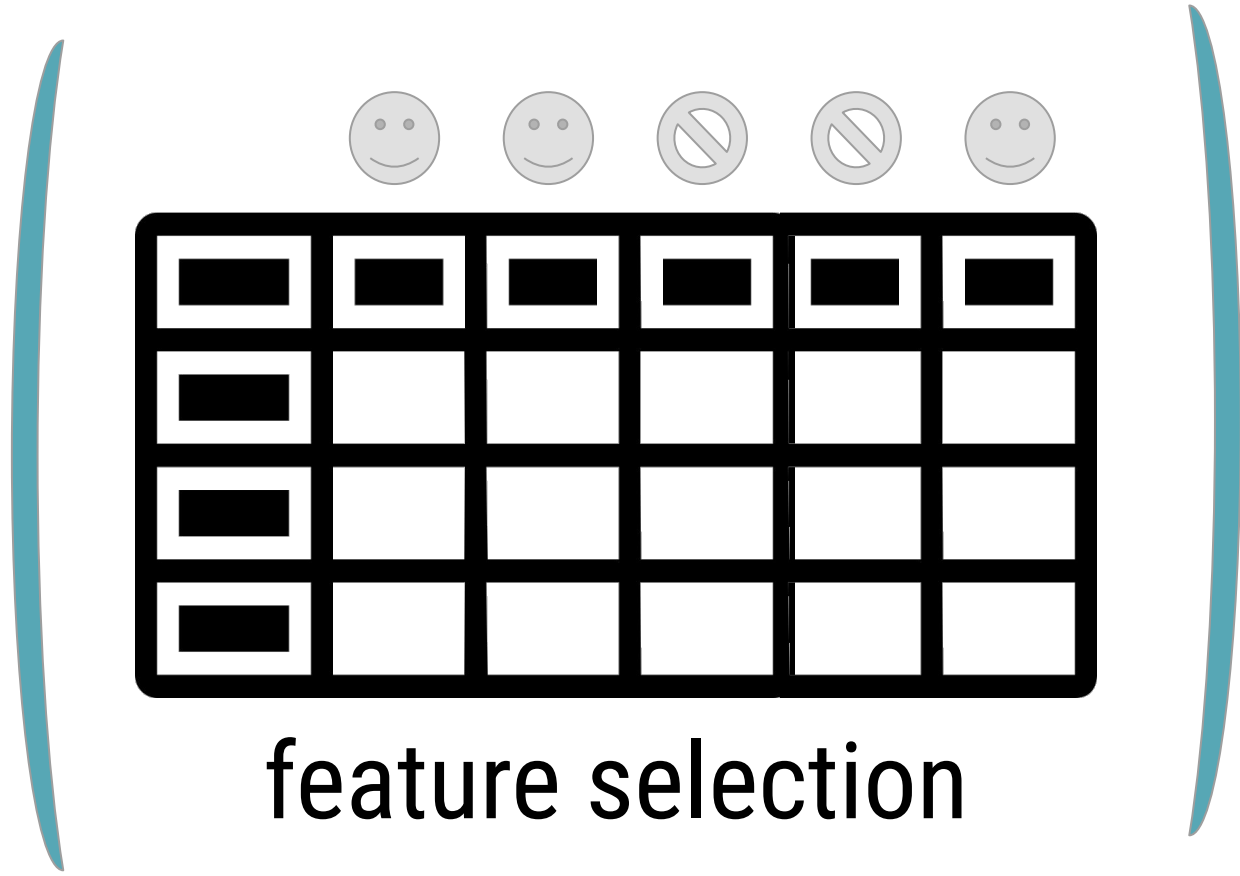


# Data Partitioning



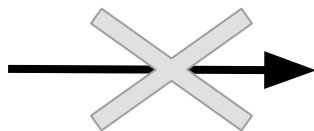
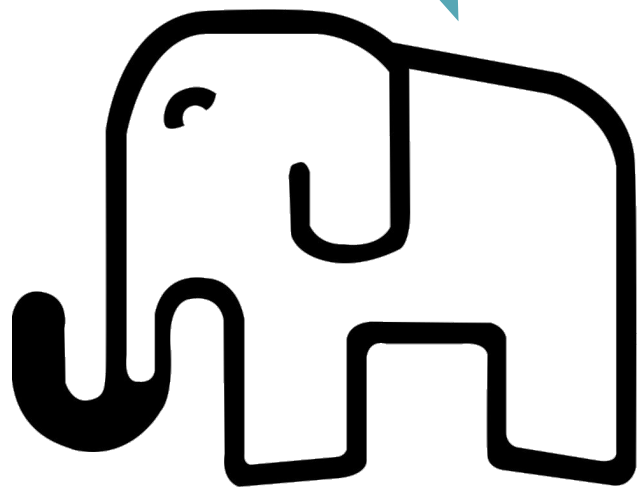
What portion of the data are typically used for generating the model?



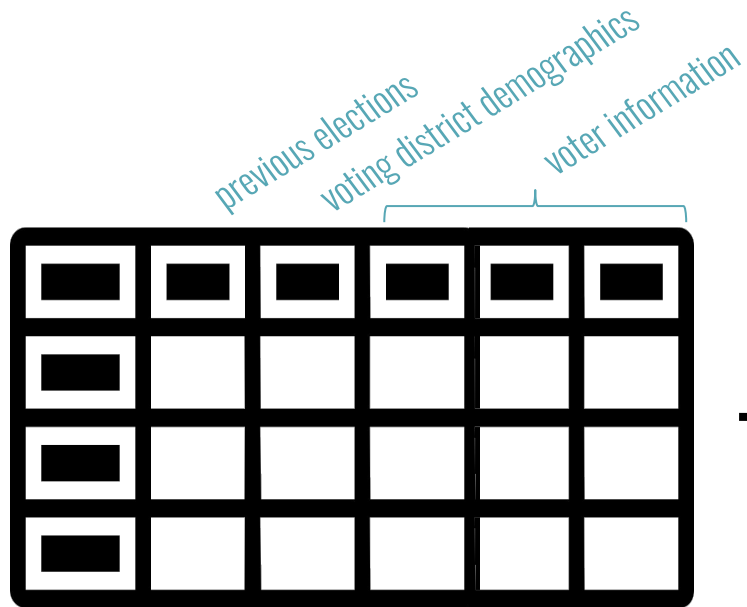


feature selection

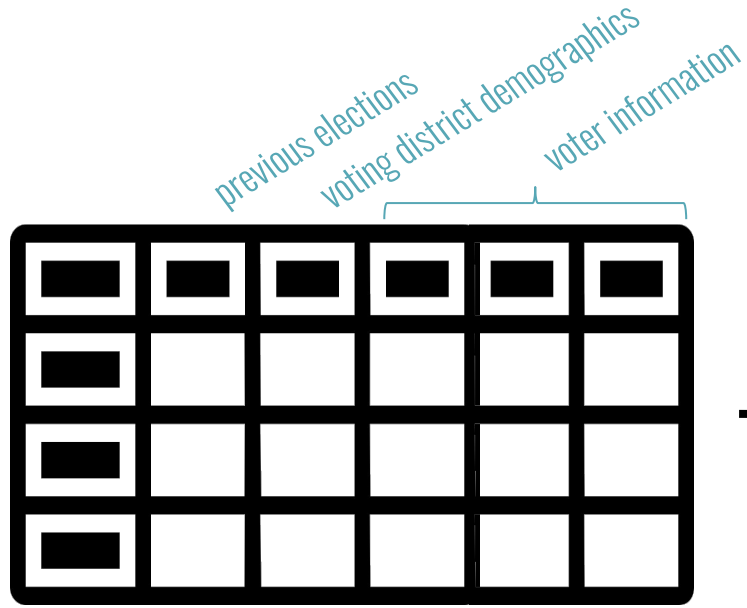
elephant height data are likely  
not predictive of US elections







these data are likely  
predictive of US election  
outcomes



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

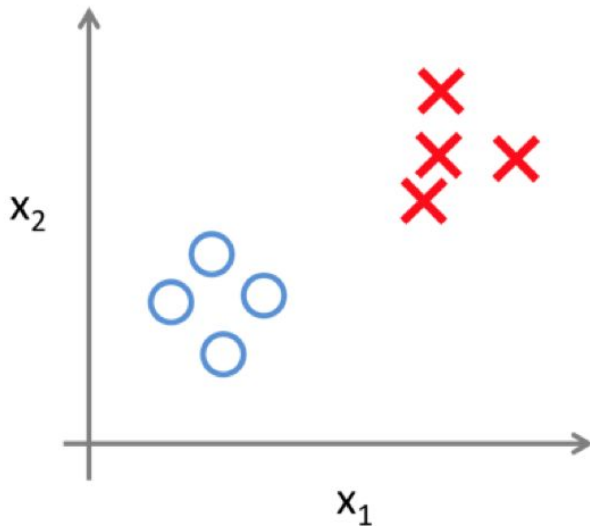
■	■	■	■	■	■
■					
■					
■					



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

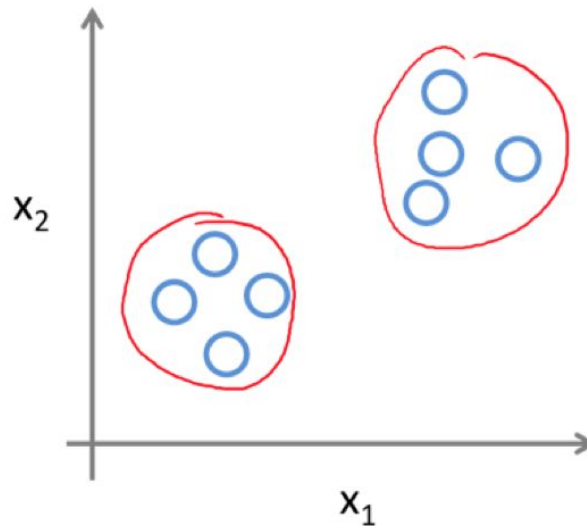
# To modes of machine learning

## Supervised Learning



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

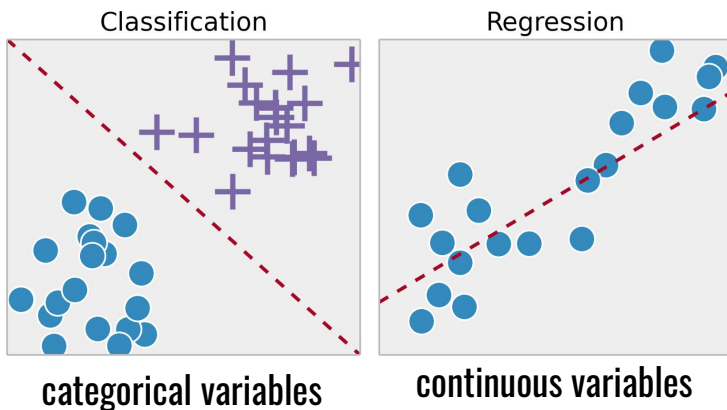
## Unsupervised Learning



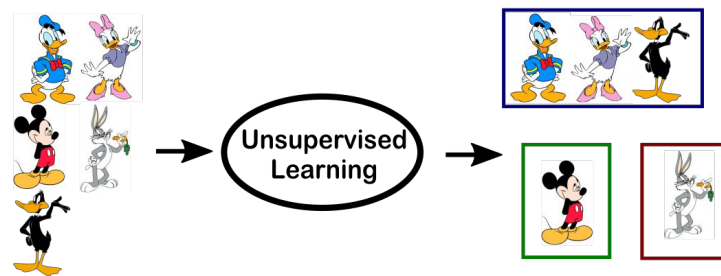
The computer determines how to classify based on properties within the data

# Approaches to machine learning

## Supervised Learning

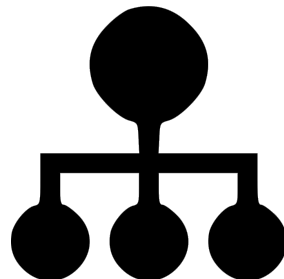
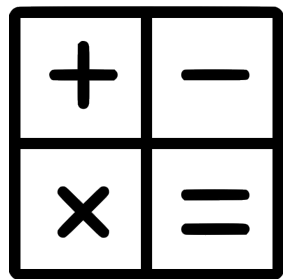
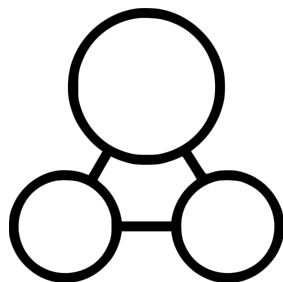


## Unsupervised Learning

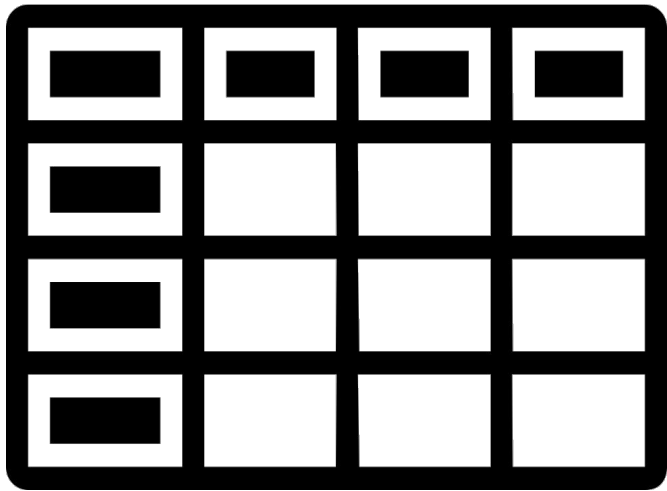


Clustering (categorical)  
& dimensionality reduction (continuous)

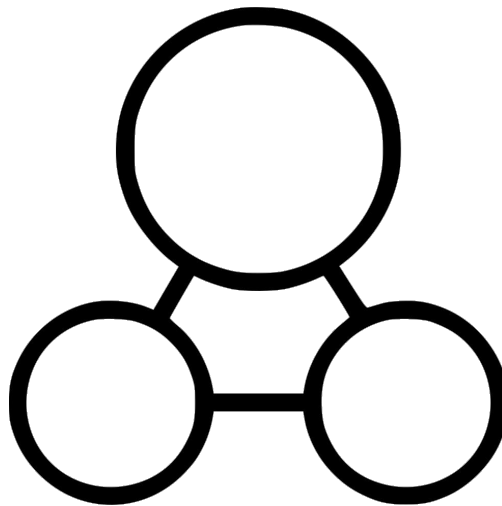
can automatically  
identify structure in  
data



model selection

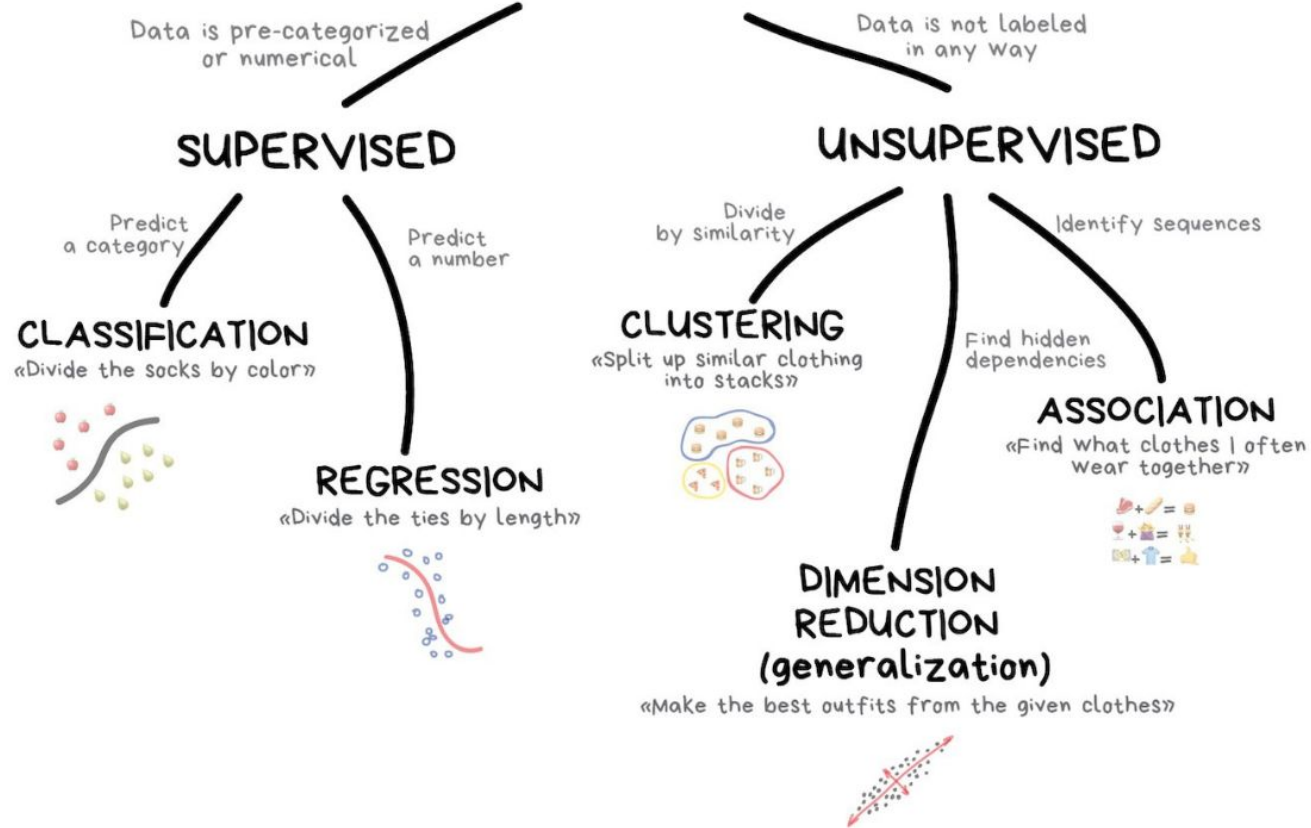


big  
datasets

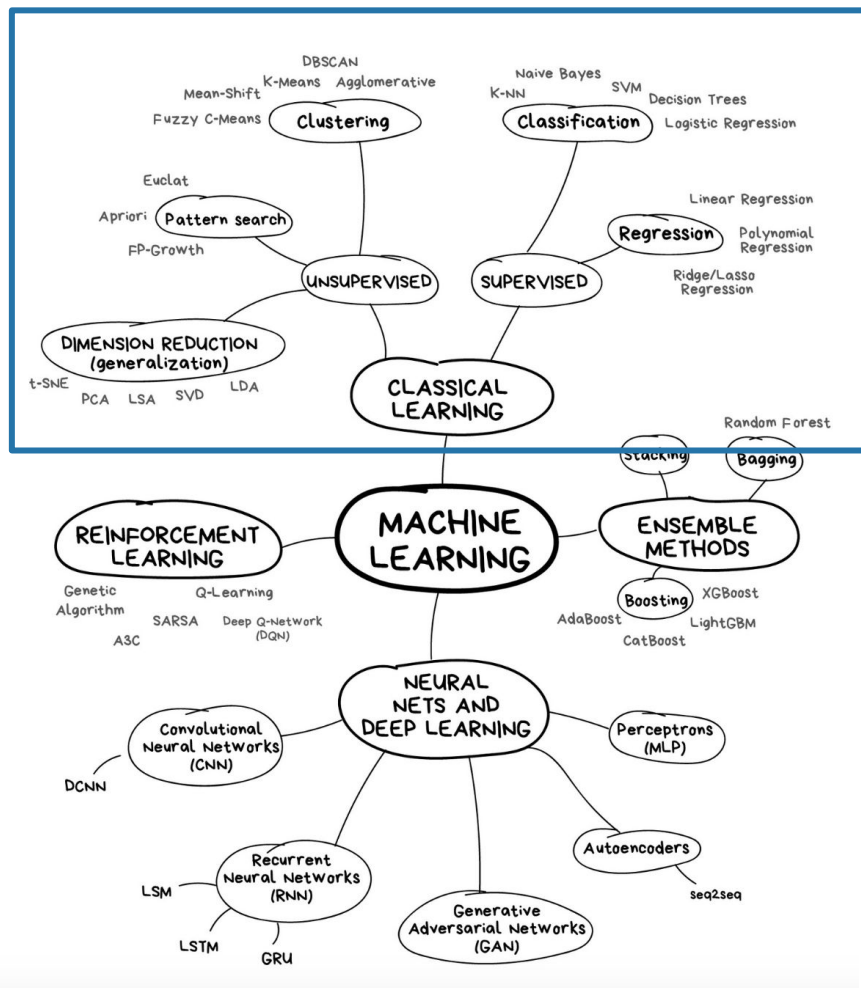


simple  
models

# CLASSICAL MACHINE LEARNING





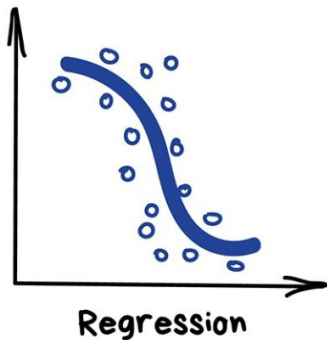


## Regression

*"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



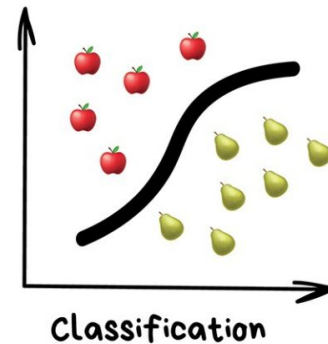
Popular algorithms are Linear and Polynomial regressions.

## Classification

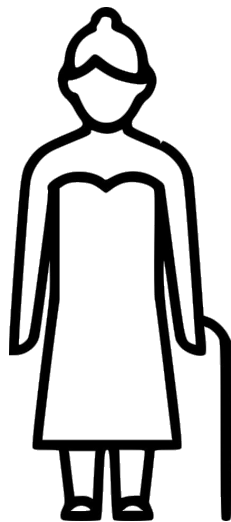
*"Splits objects based at one of the attributes known beforehand. Separate socks by based on color, documents based on language, music by genre"*

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



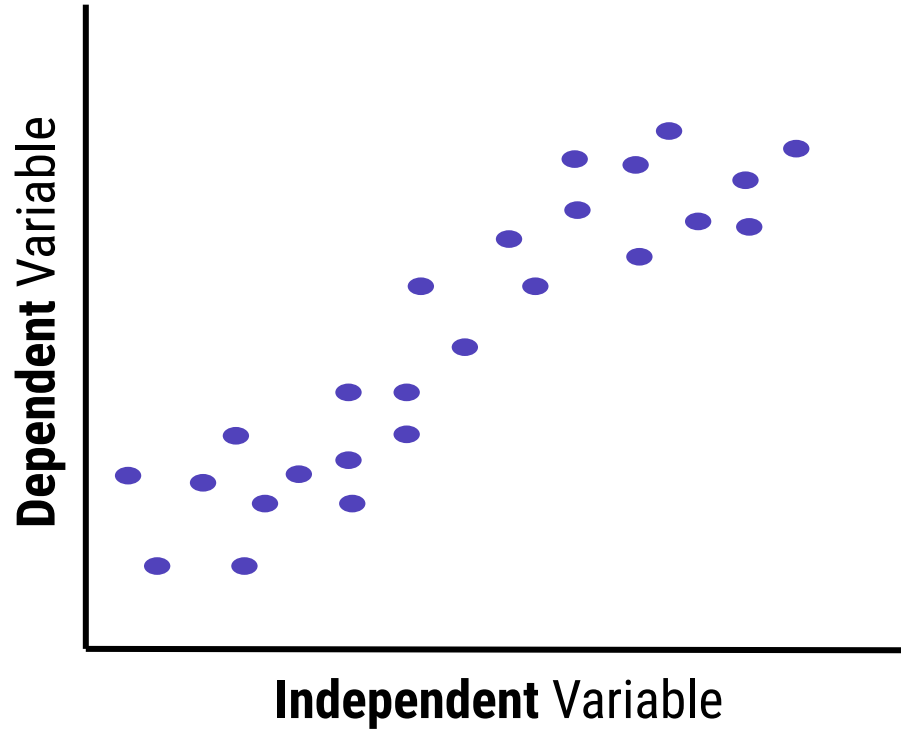
**Regression:**  
predicting continuous  
variables  
(i.e. Age)

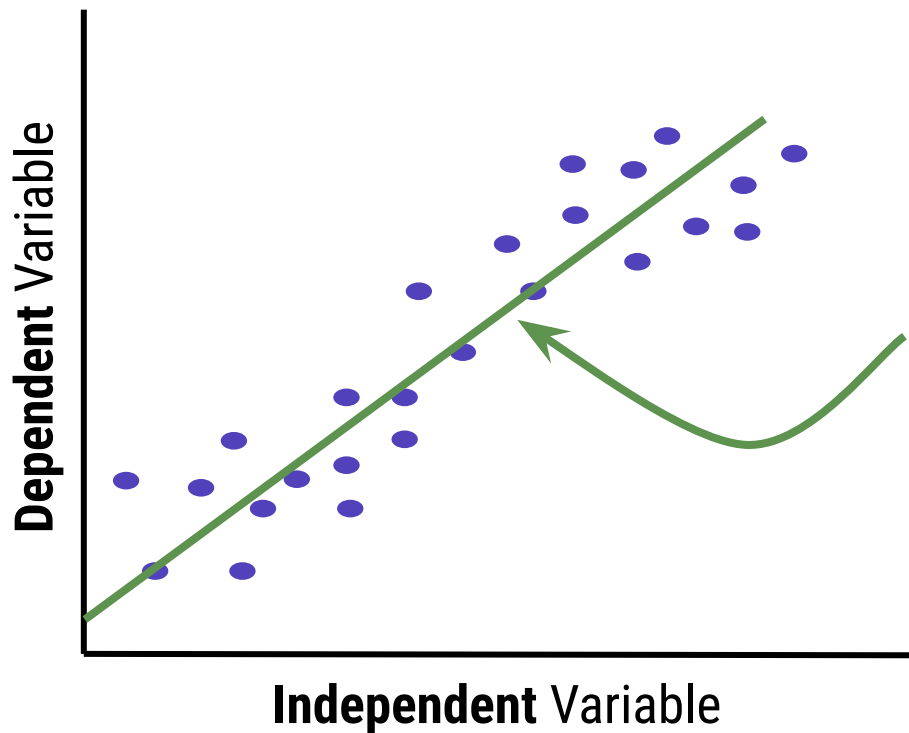
continuous variable prediction



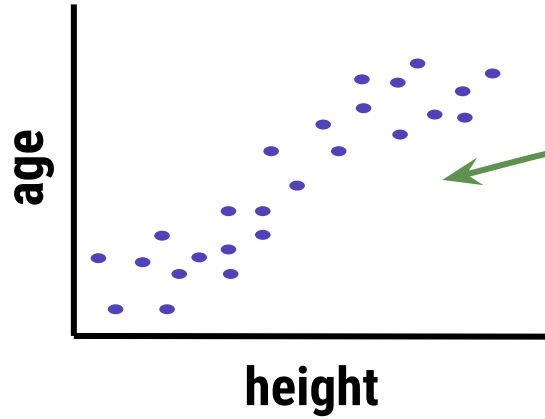
**Classification:**  
predicting categorical  
variables  
(i.e. education level)

categorical variable prediction

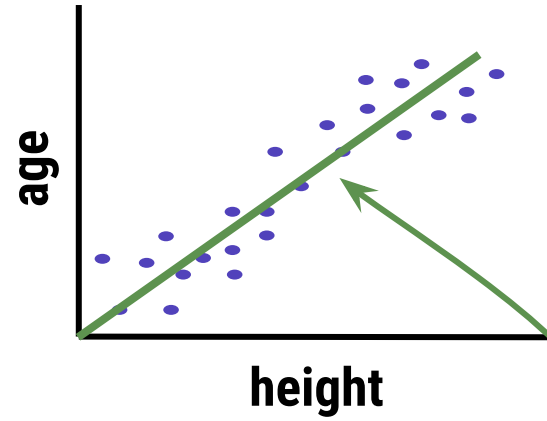
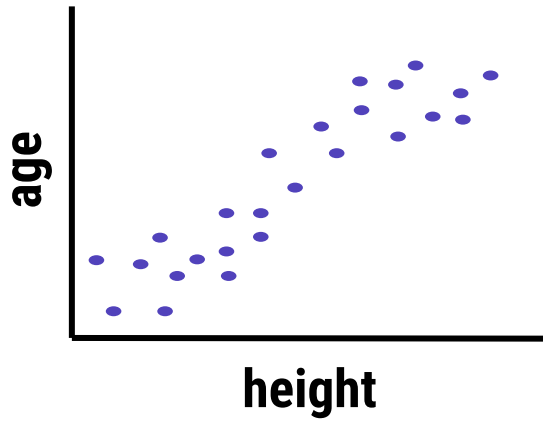




We'll use the linear relationship between variables to generate a **predictive model**



the training data will  
be used to build the  
predictive model



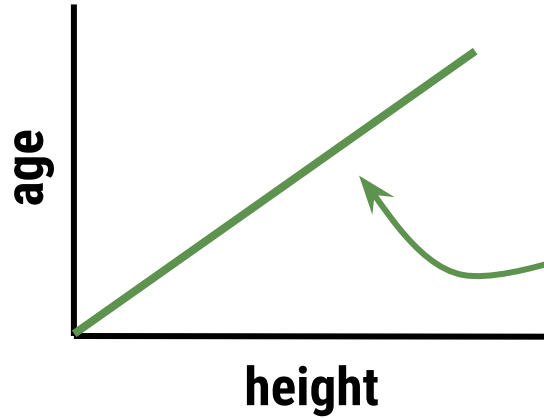
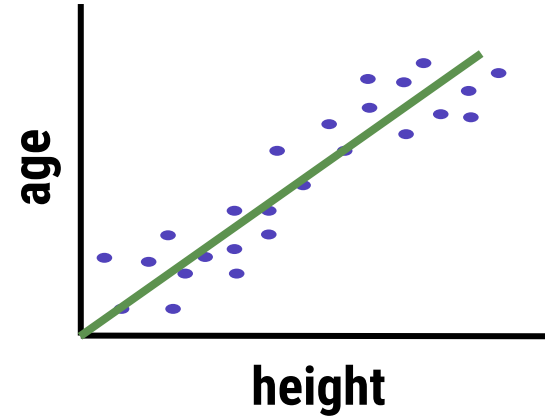
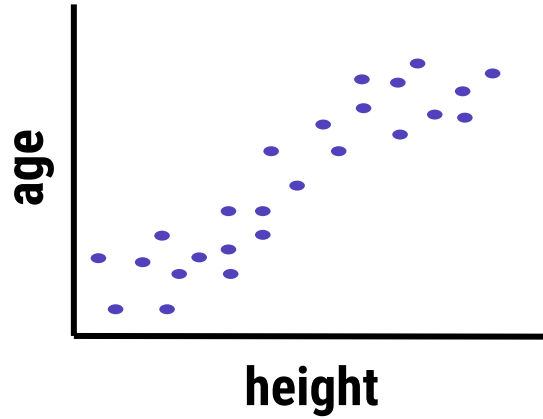
Supervised Learning

regression

continuous variable prediction

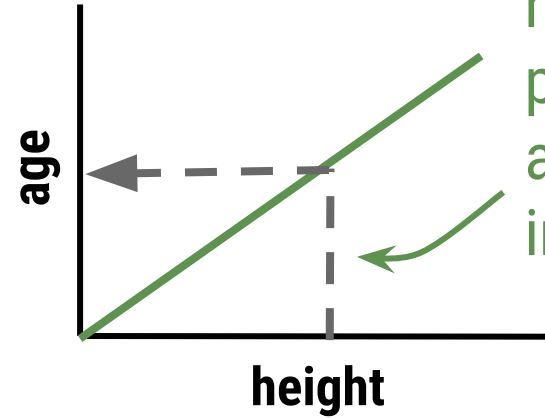
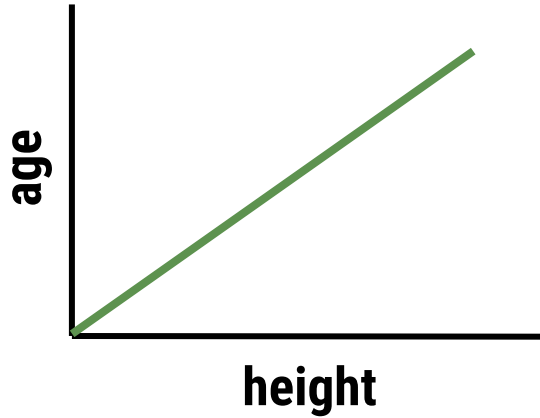
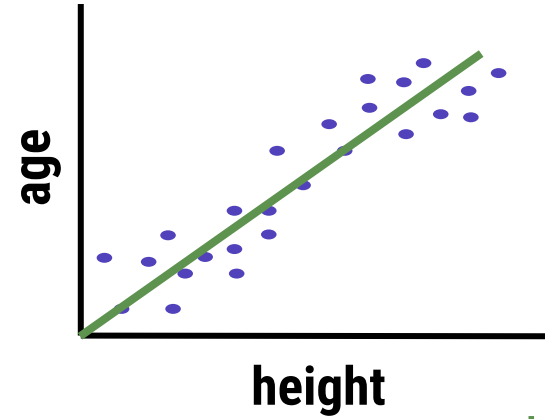
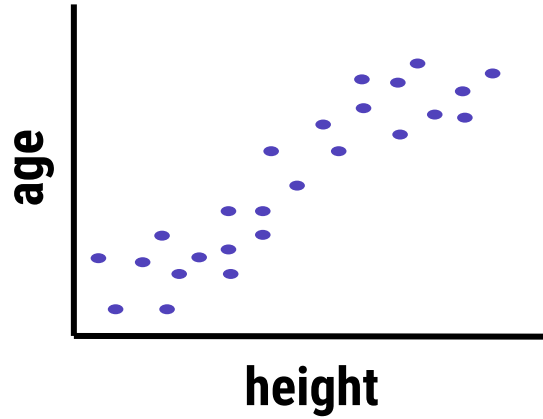


use linear regression to  
model the relationship

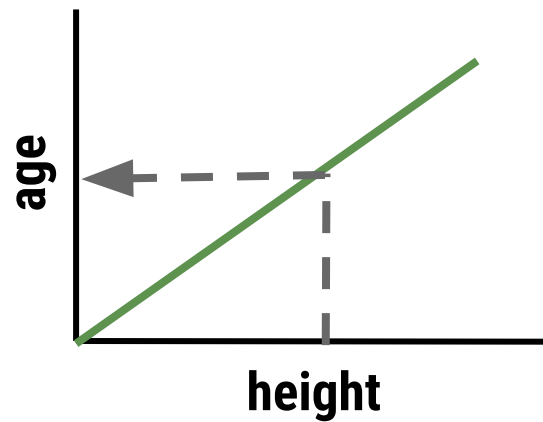
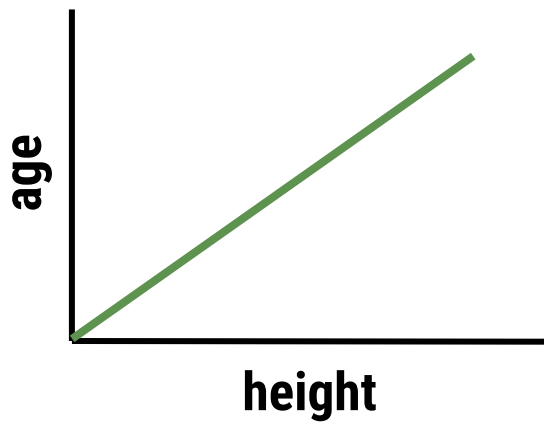
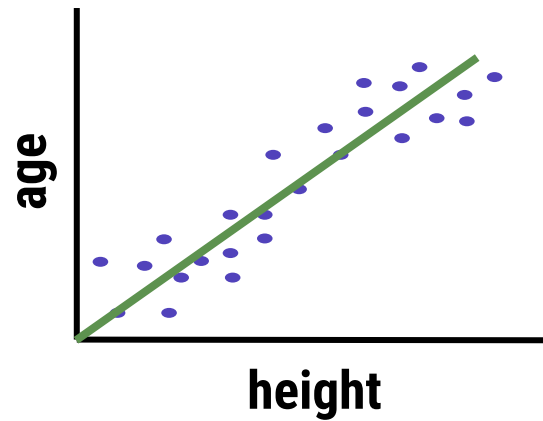
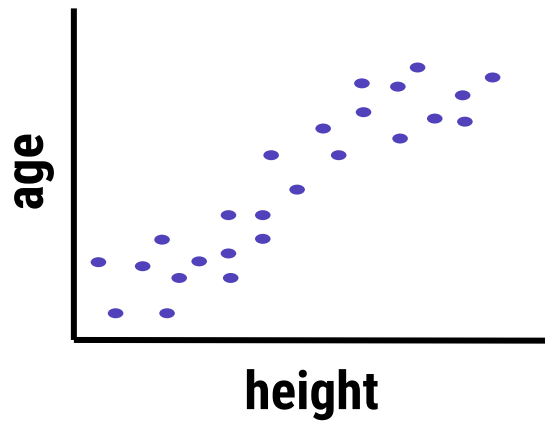


For prediction, the individual values in the training data are *not* important. We only need the model.

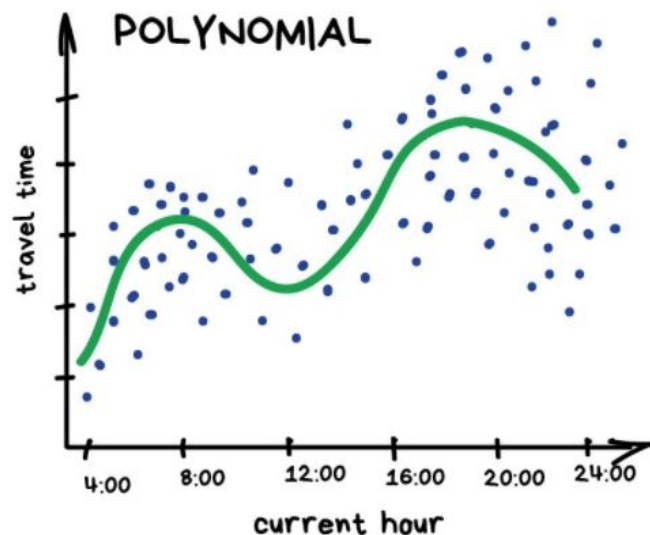
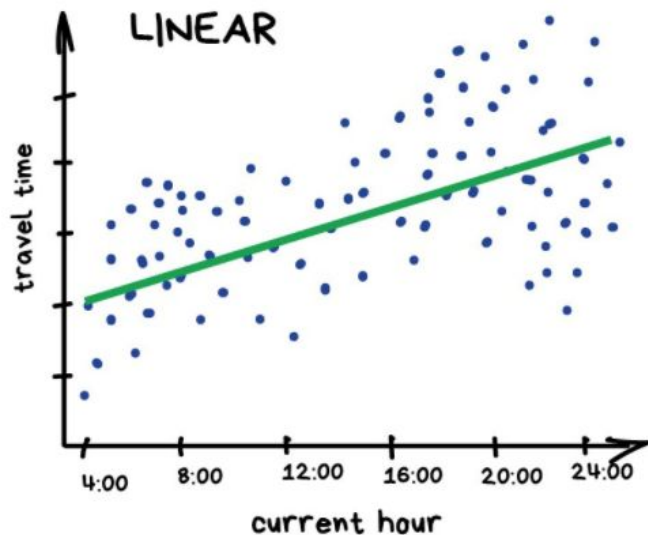




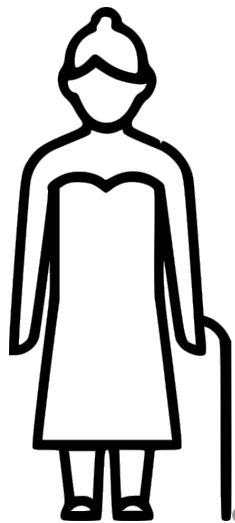
how we'll make  
predictions for  
a future  
individual



## PREDICT TRAFFIC JAMS



REGRESSION

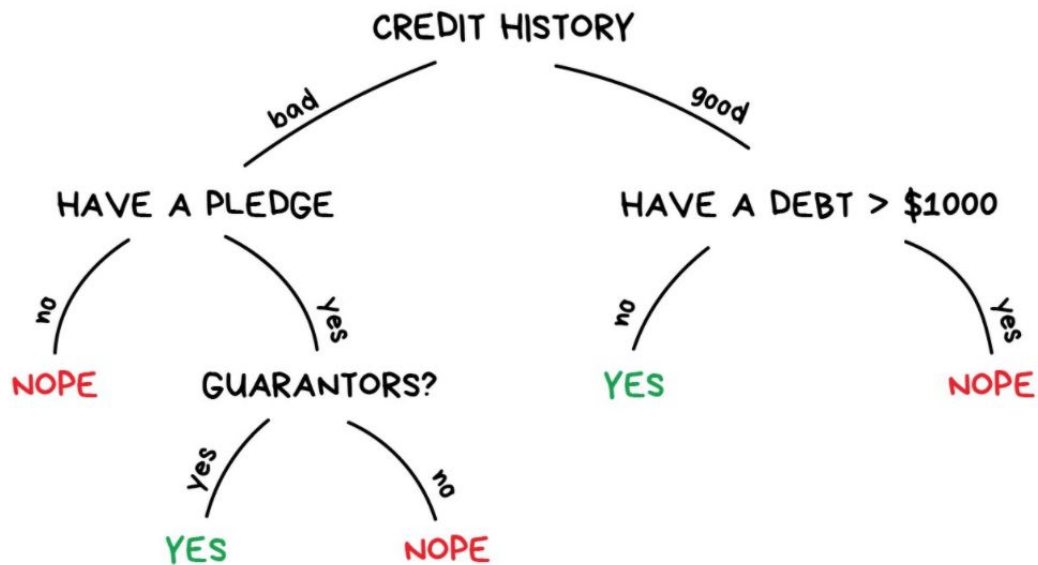


**Regression:**  
predicting continuous  
variables  
(i.e. Age)



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

# GIVE A LOAN?



DECISION TREE



training data

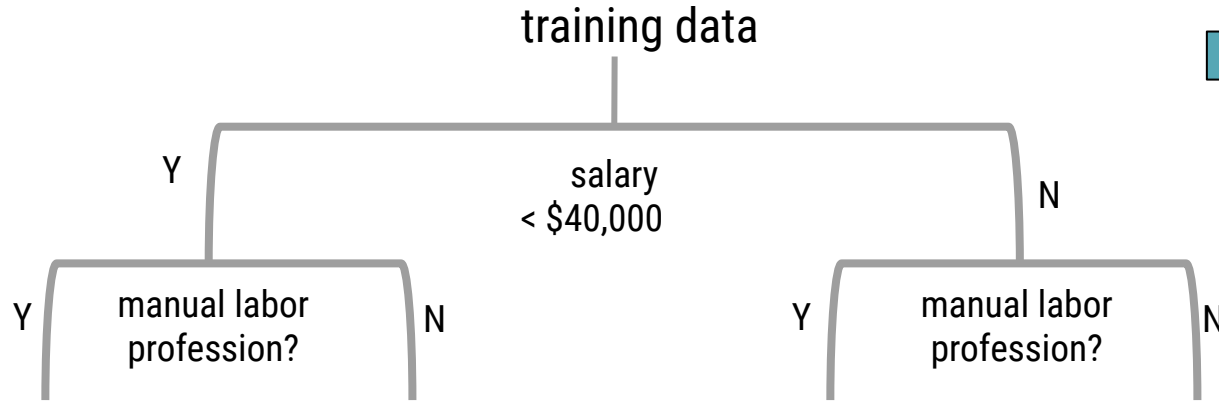
Y

salary  
< \$40,000

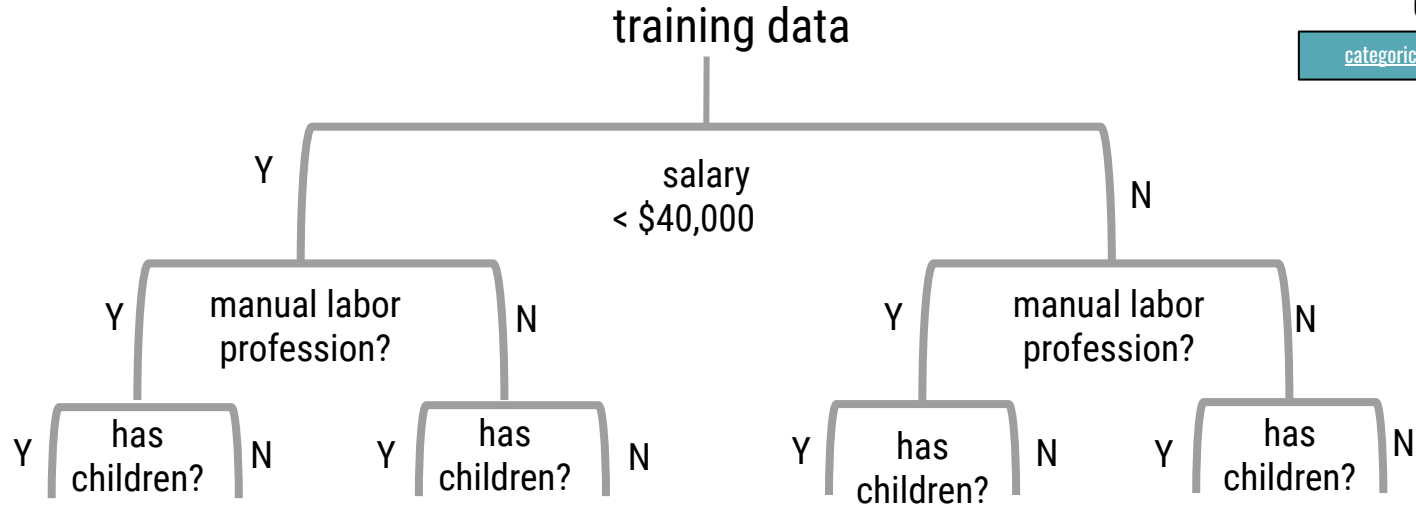
N

All the people  
who make *less*  
than 40K over  
here

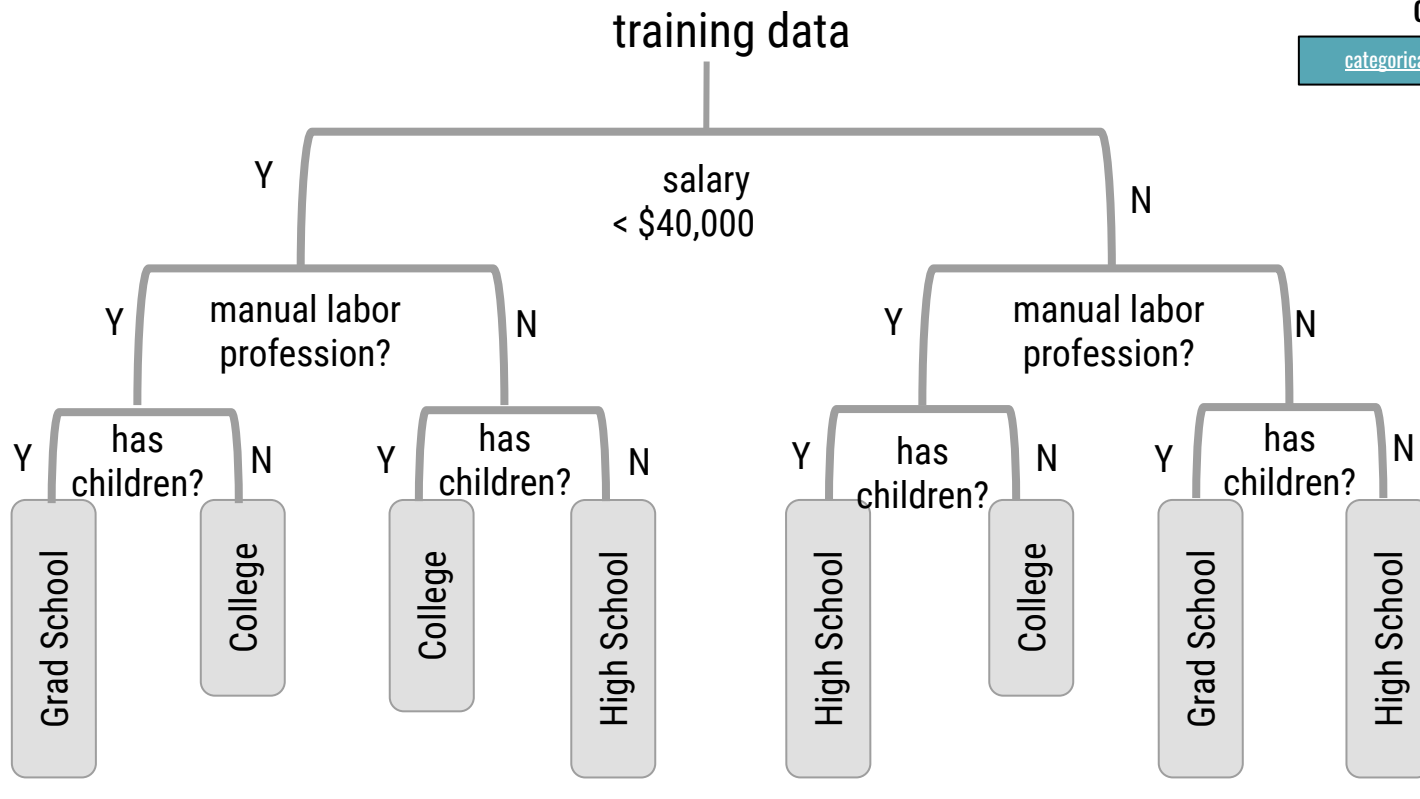
All the people  
who make *more*  
than 40K over  
here



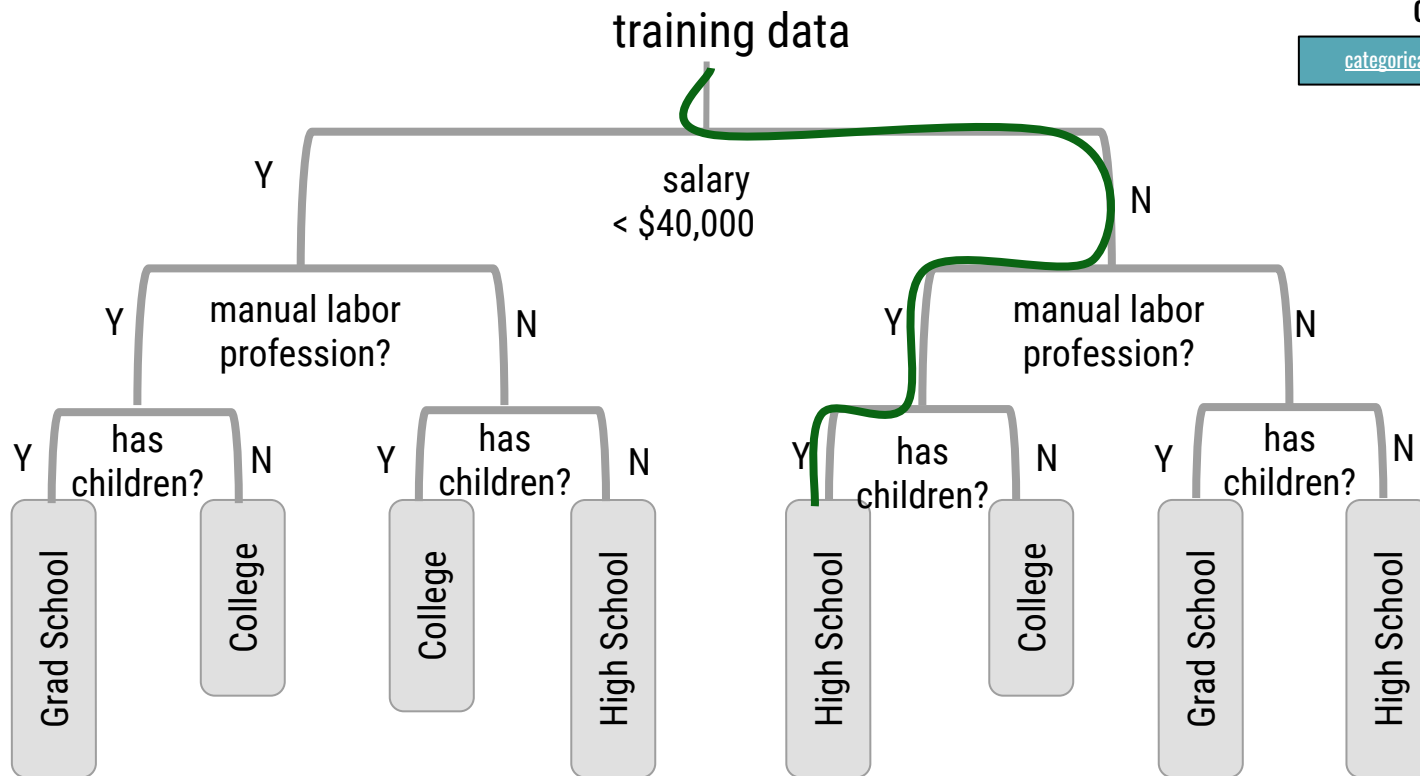
Continue adding *branches* to the **decision tree** where the variables and information in the training data decide which observations goes down which branch







At the end of the tree, labels will be applied to each *leaf* of the tree



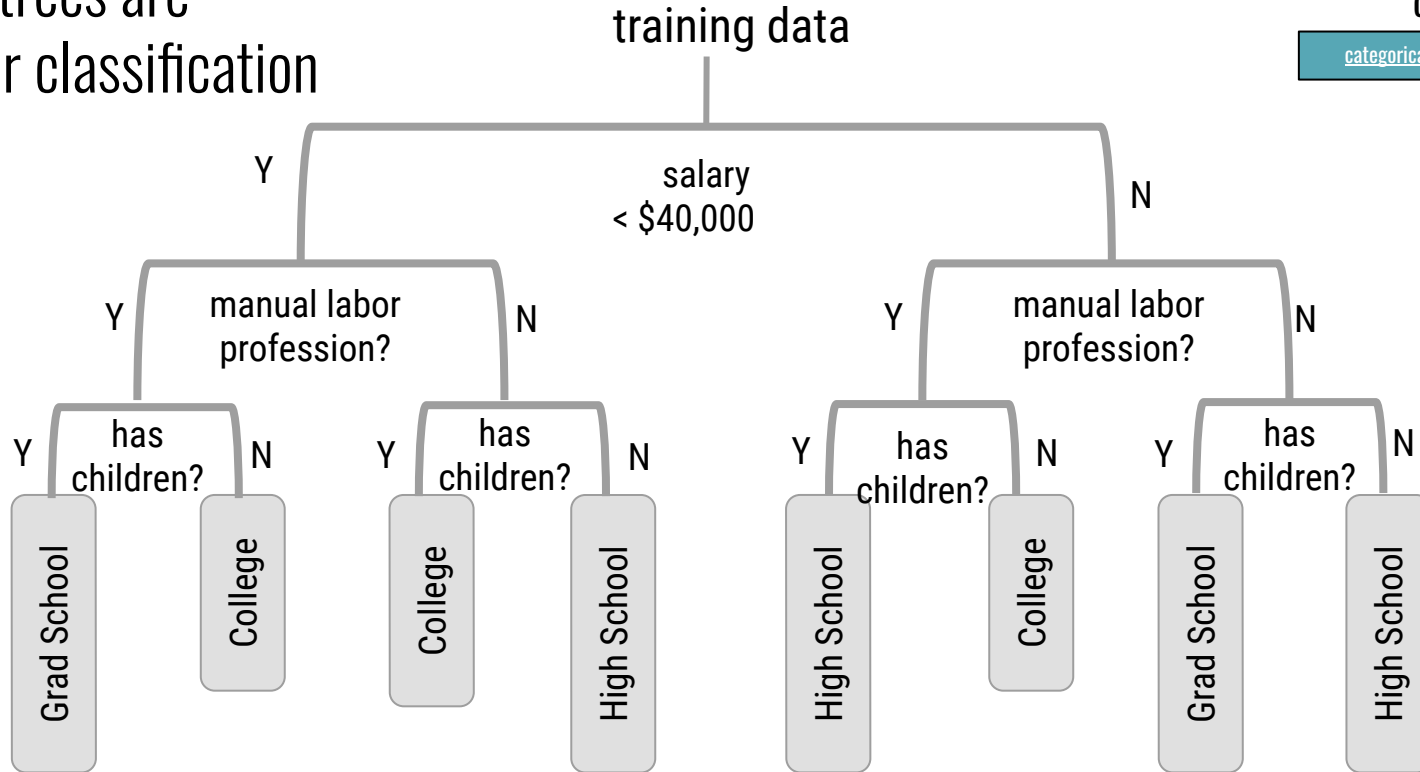
# Decision trees are helpful for classification

Supervised Learning

classification



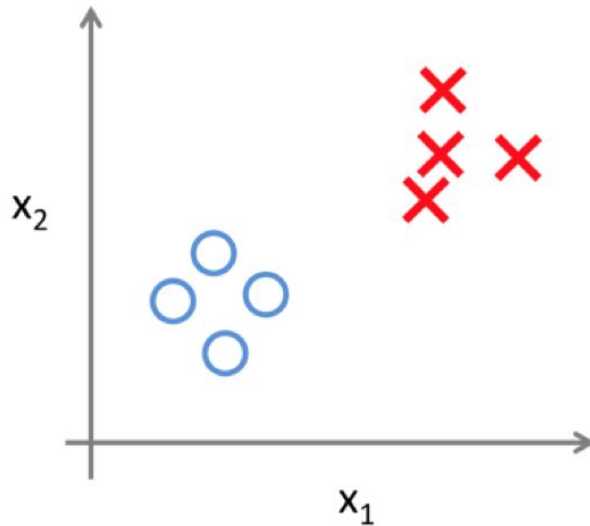
categorical variable prediction



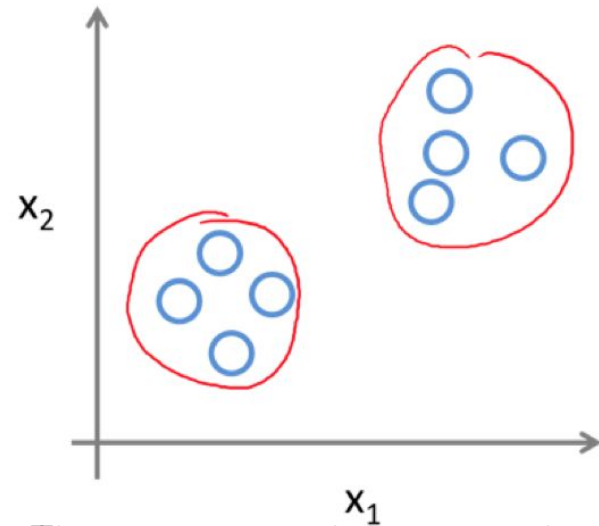
# Unsupervised Learning

# To modes of machine learning

## Supervised Learning



## Unsupervised Learning



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

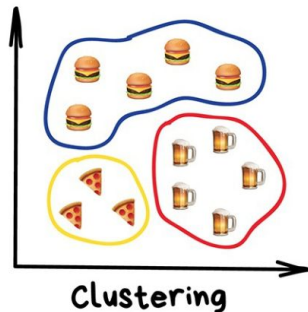
## Dimensionality Reduction (Generalization)

### Clustering

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

Nowadays used:

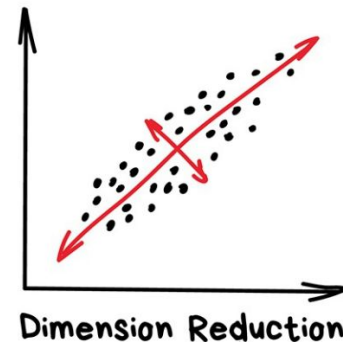
- For market segmentation (types of customers, loyalty)
- To merge close points on a map
- For image compression
- To analyze and label new data
- To detect abnormal behavior



"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

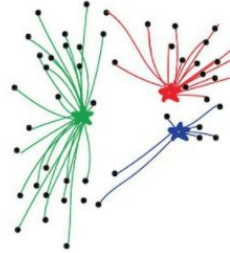
Popular algorithms: Principal Component Analysis (PCA), Singular Value Decomposition (SVD), Latent Dirichlet allocation (LDA), Latent Semantic Analysis (LSA, pLSA, GLSA), t-SNE (for visualization)

## PUT KEBAB KIOSKS IN THE OPTIMAL WAY

(also illustrating the K-means method)



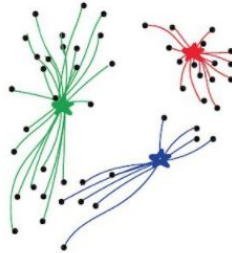
1. Put kebab kiosks in random places in city



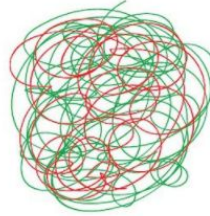
2. Watch how buyers choose the nearest one



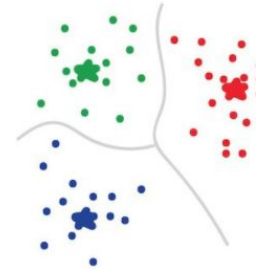
3. Move kiosks closer to the centers of their popularity



4. Watch and move again



5. Repeat a million times



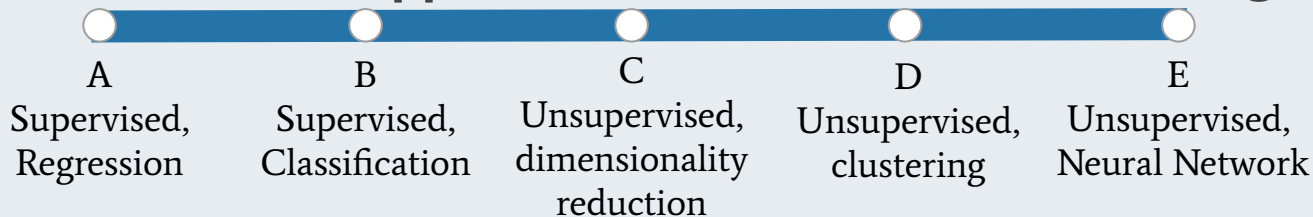
6. Done!  
You're god of kebabs!

# Prediction Approach

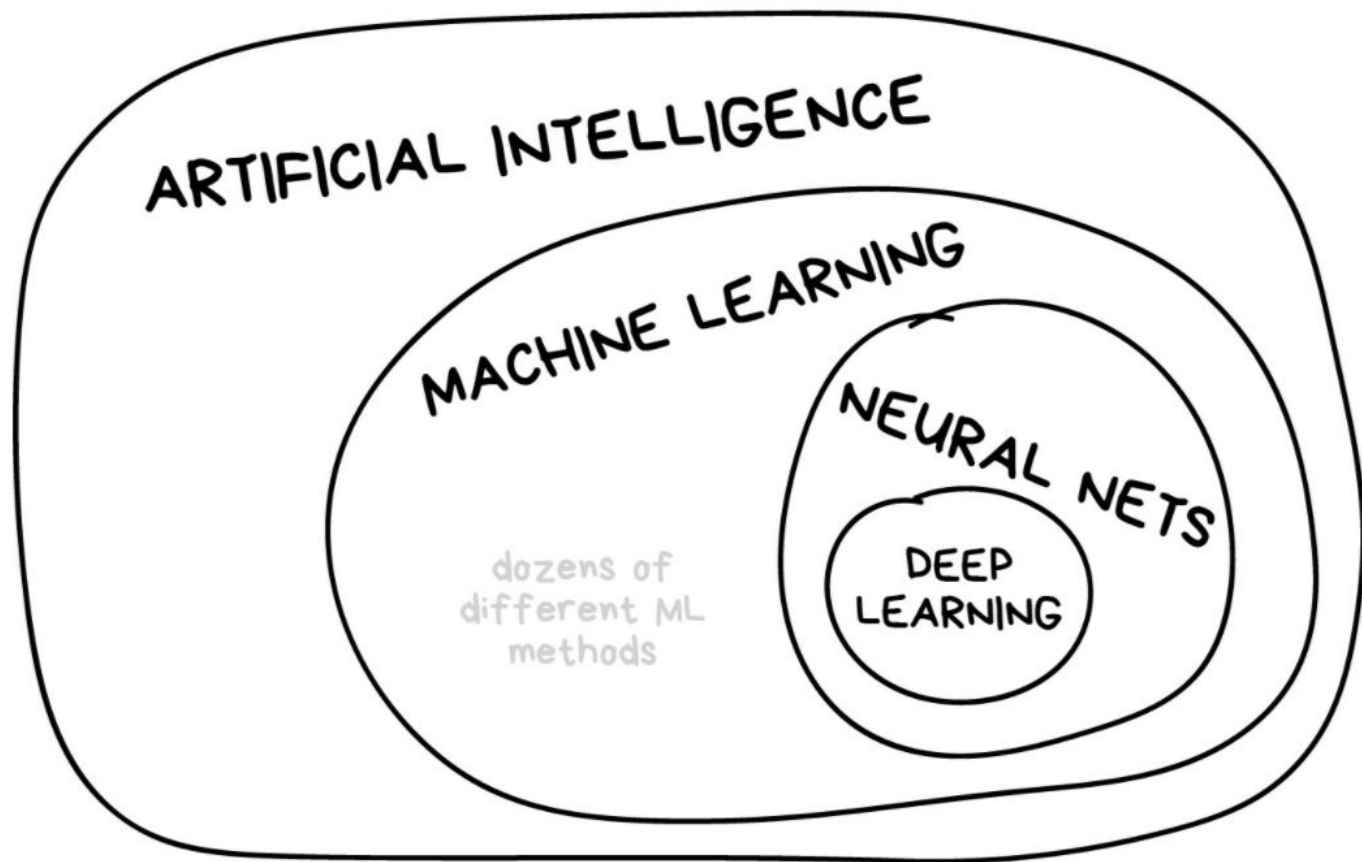


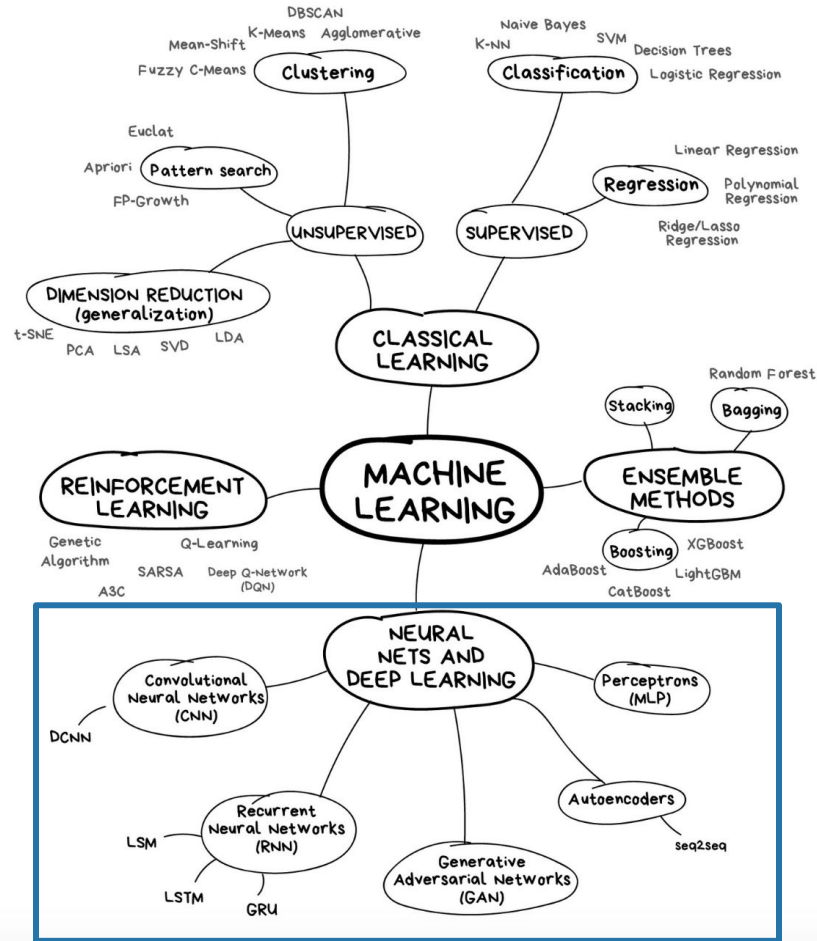
You want to predict someone's emotion based on an image.

How would you approach this with machine learning?





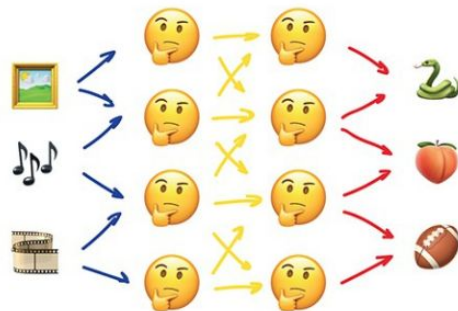




*"We have a thousand-layer network, dozens of video cards, but still no idea where to use it. Let's generate cat pics!"*

Used today for:

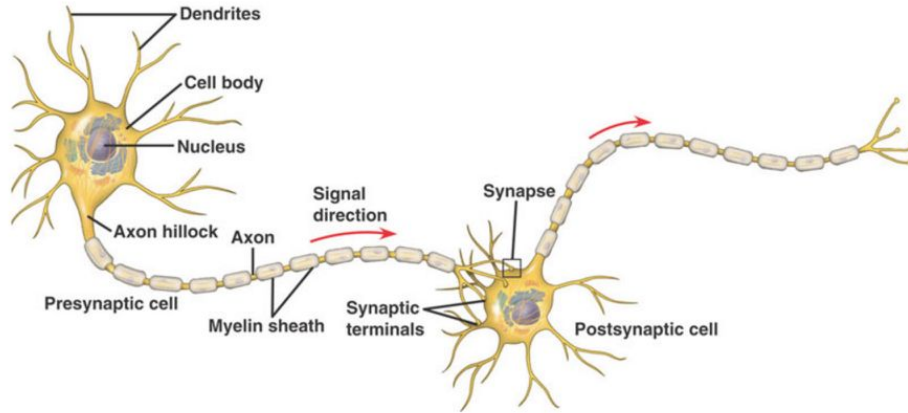
- Replacement of all algorithms above
- Object identification on photos and videos
- Speech recognition and synthesis
- Image processing, style transfer
- Machine translation



## Neural Networks

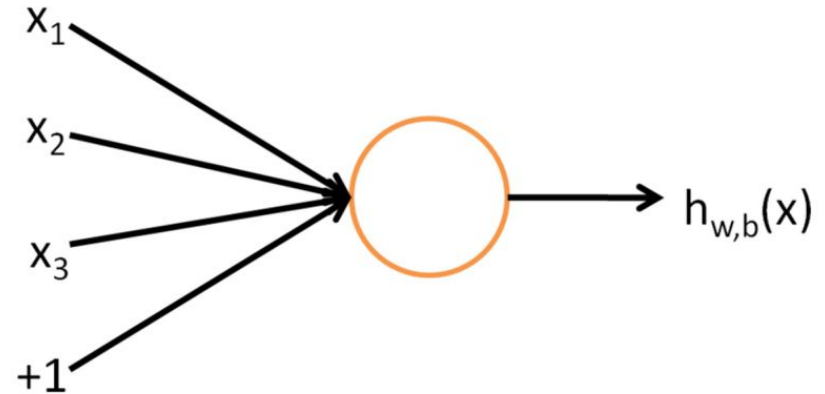
Popular architectures: Perceptron, Convolutional Network (CNN), Recurrent Networks (RNN), Autoencoders

# WHAT IS A NEURON?



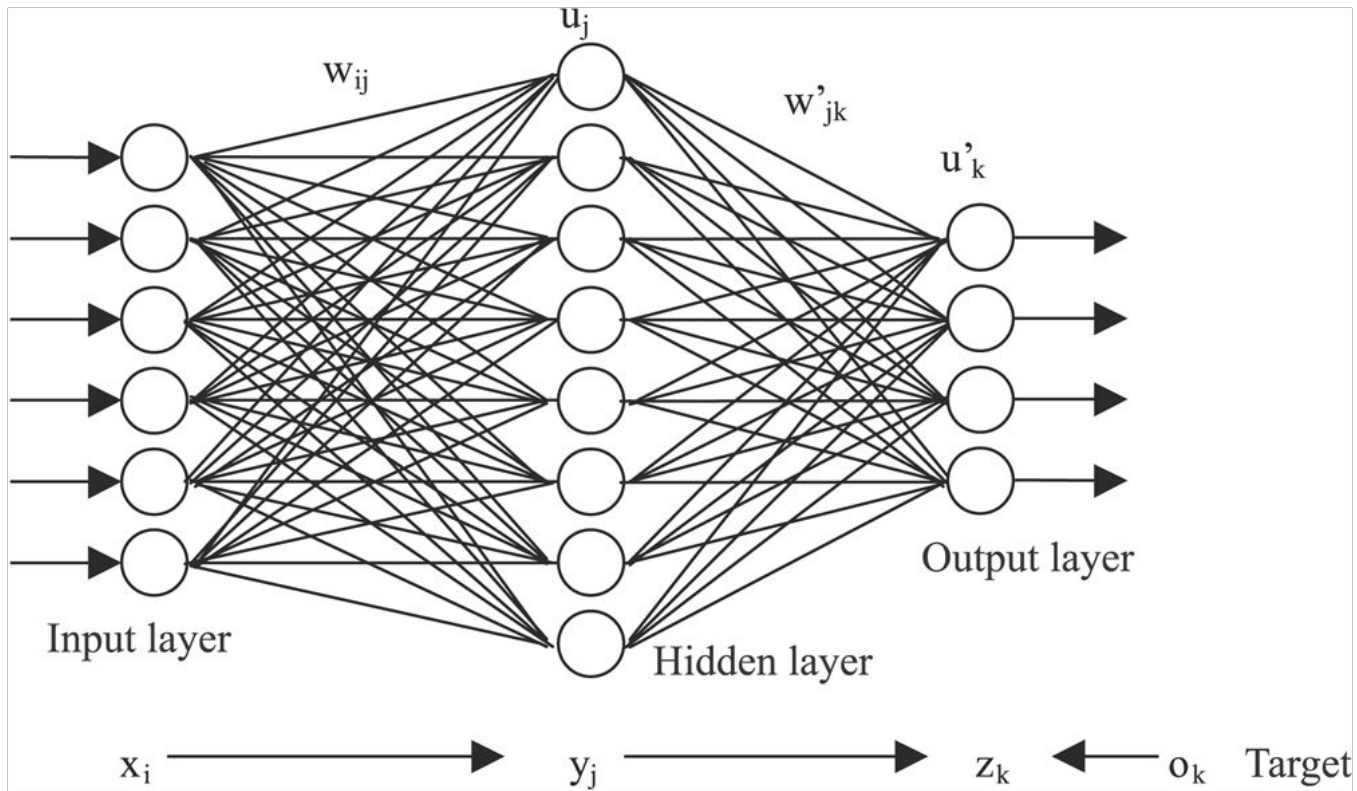
- Receives signal on synapse
- When trigger sends signal on axon

# 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

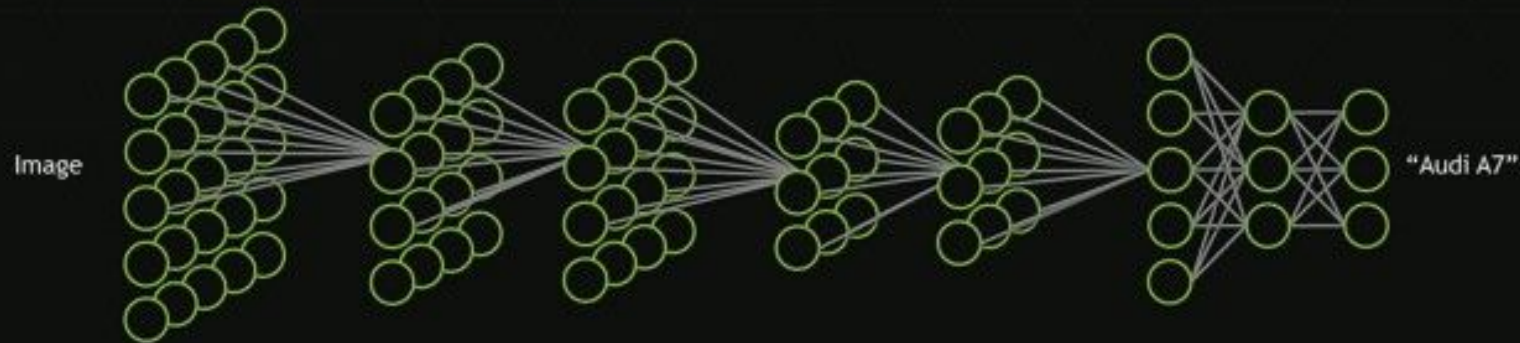
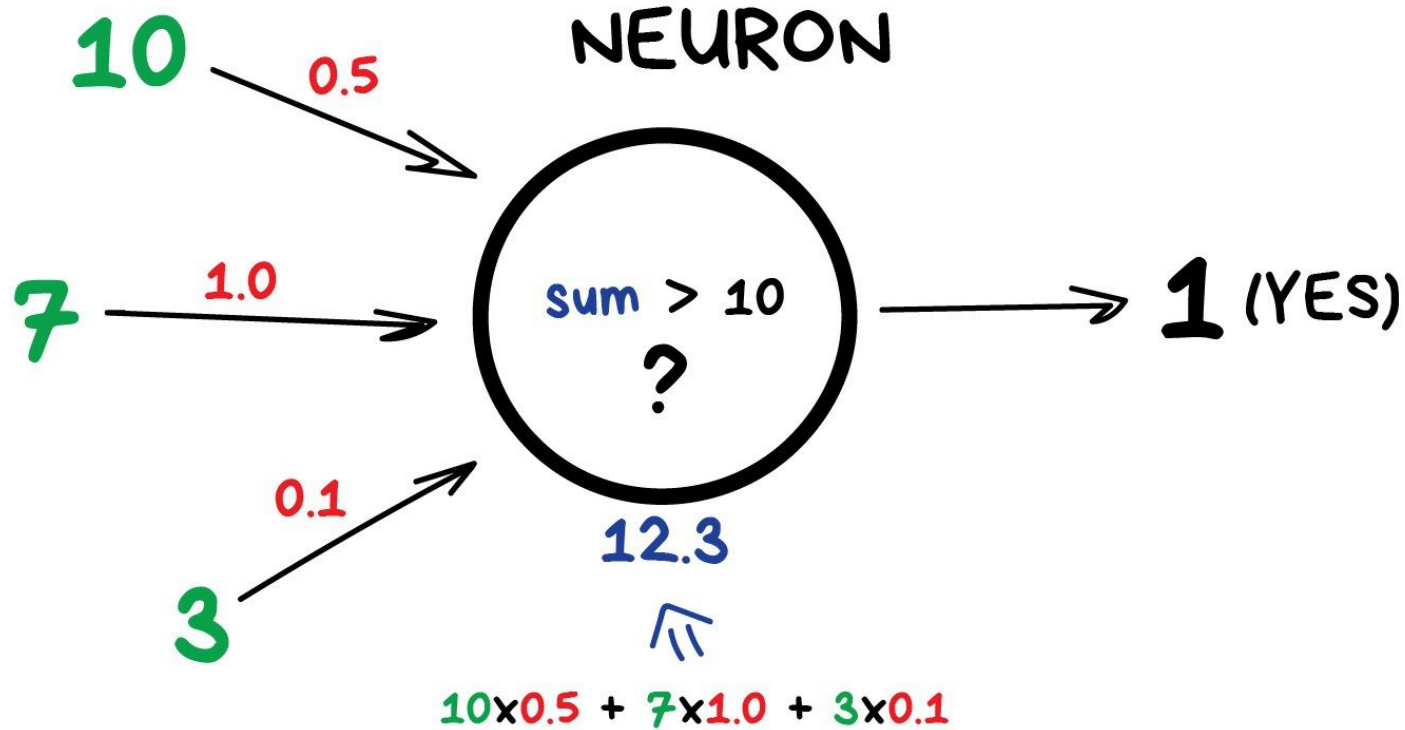
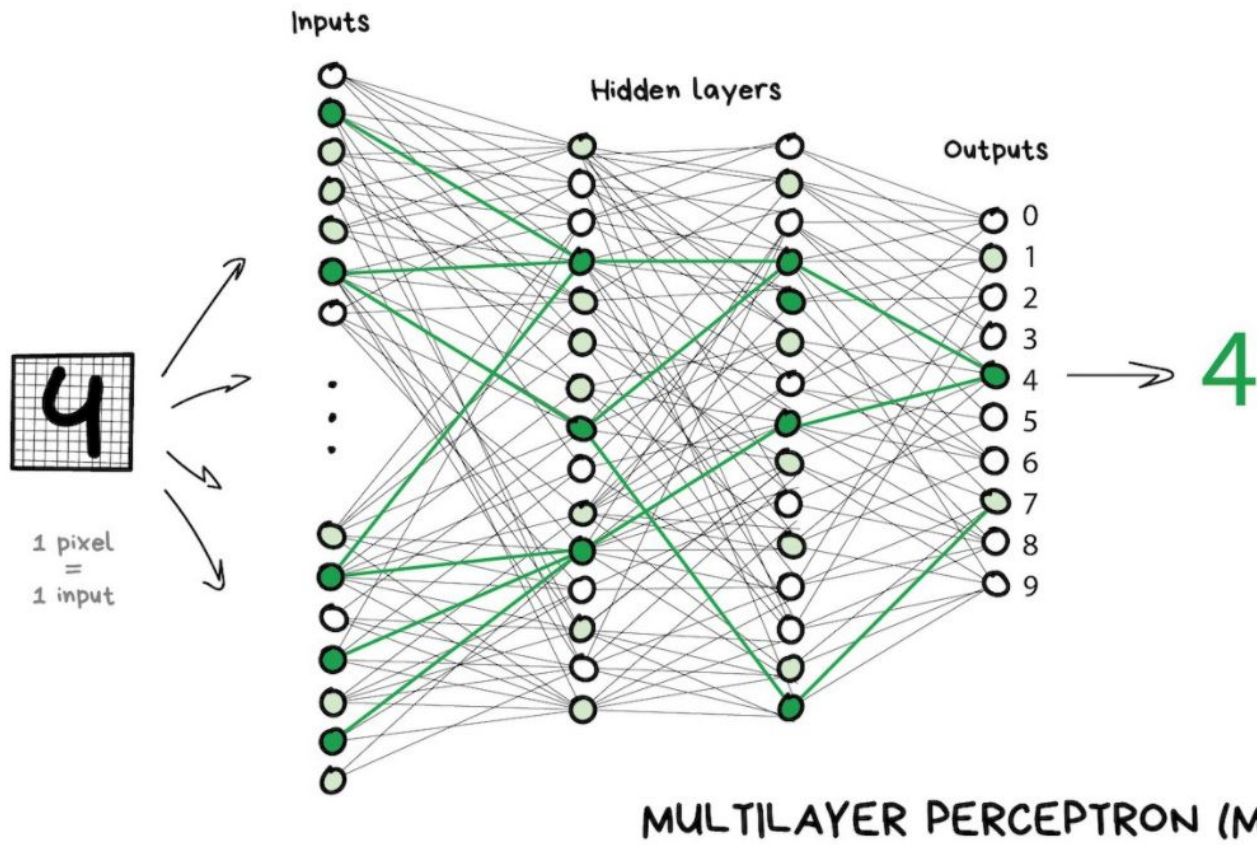


Image source: "Unsupervised Learning of Hierarchical Representations with Convolutional Deep Belief Networks" ICML 2009 & Comm. ACM 2011, Honglak Lee, Roger Grosse, Rajesh Ranganath, and Andrew Ng.

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...



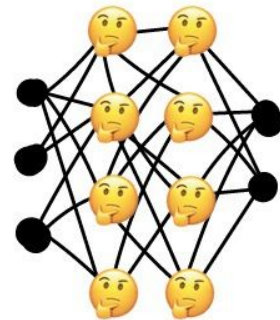
Original image



Preliminary processing



Hand-crafted  
features



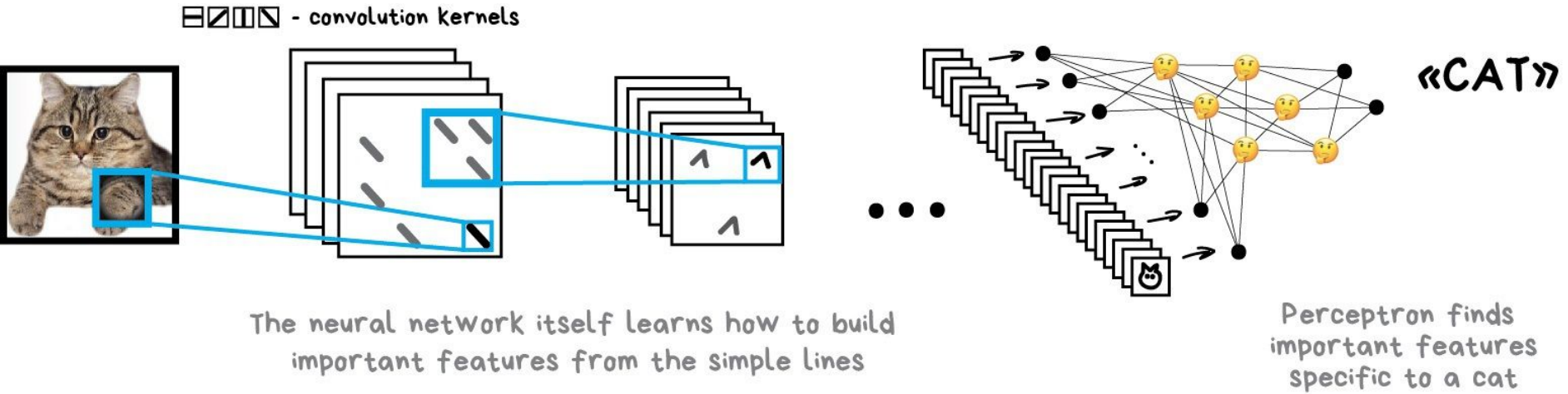
Neural Network



«cat»

Result

# 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)



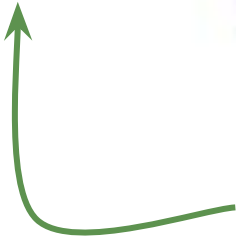


model assessment

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# Root Mean Squared Error (RMSE)

$$RMSE = \sqrt{\frac{\sum_{i=1}^N (\text{Predicted}_i - \text{Actual}_i)^2}{N}}$$



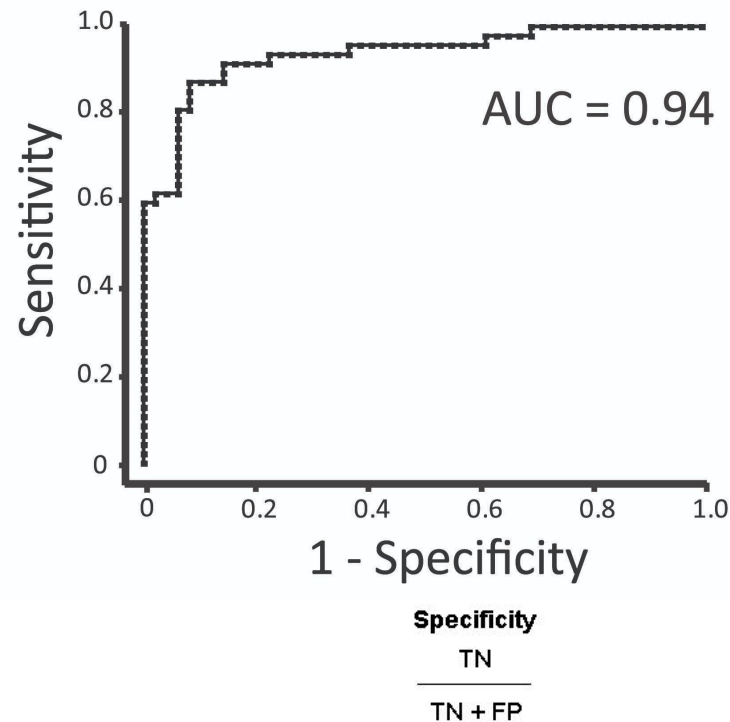
A few outliers can lead to a big increase in RMSE, even if all the other predictions are pretty good

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

		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

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



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

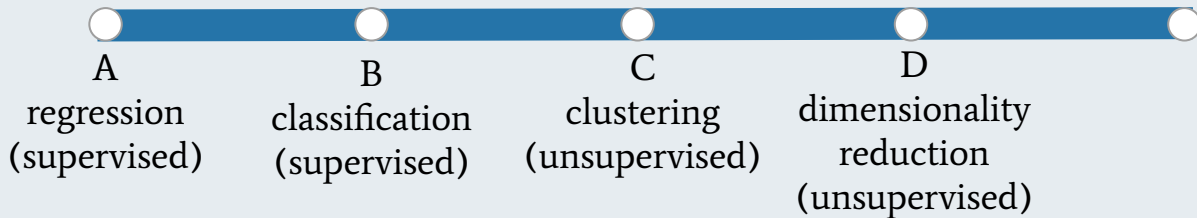


# Prediction Approach



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

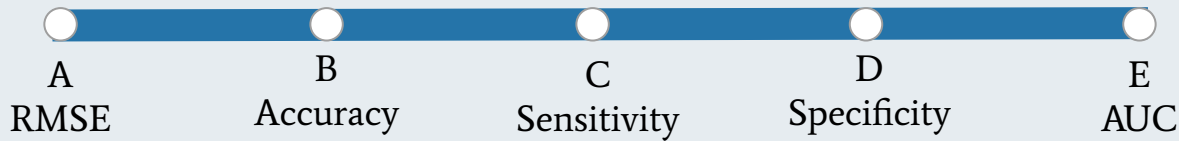
**What prediction approach would you use?**



# 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?

