What is ML? DL? RL? AI?

- There are many domains we'll cover in this course, including:
 - ML Machine Learning
 - DL Deep Learning
 - RL Reinforcement Learning
 - AI Artificial Intelligence

 This Overview Section is designed to help understand how Artificial Intelligence, Machine Learning, Deep Learning, and Reinforcement Learning are related to each other.

- We'll explore "standard" machine learning concepts, such as Supervised Learning and Unsupervised Learning.
- Then we'll see how Reinforcement Learning differs from these more traditional methods.

- By the end of these lectures, we'll understand the relationships between:
 - Machine Learning
 - Supervised Learning
 - Unsupervised Learning
 - Reinforcement Learning
 - Deep Learning

• Let's begin by exploring the domains:

• Let's begin by exploring the domains:

Artificial Intelligence

• Let's begin by exploring the domains:

Artificial Intelligence

• Intelligence demonstrated by machines.

• Let's begin by exploring the domains:

Artificial Intelligence

- Intelligence demonstrated by machines.
 - What is "intelligence"?
 - How to test for "intelligence"?

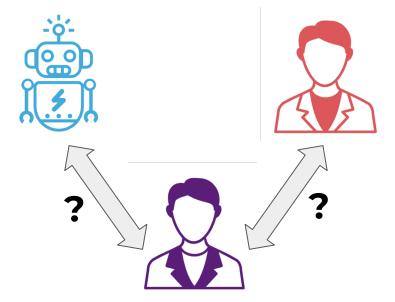












- Tests for Artificial Intelligence:
 - Marcus Test:
 - Measures a computer's ability to understand a television program.
 - Lovelace 2.0 Test:
 - Measuring a computer's ability to create artistic artifacts.

Artificial Intelligence

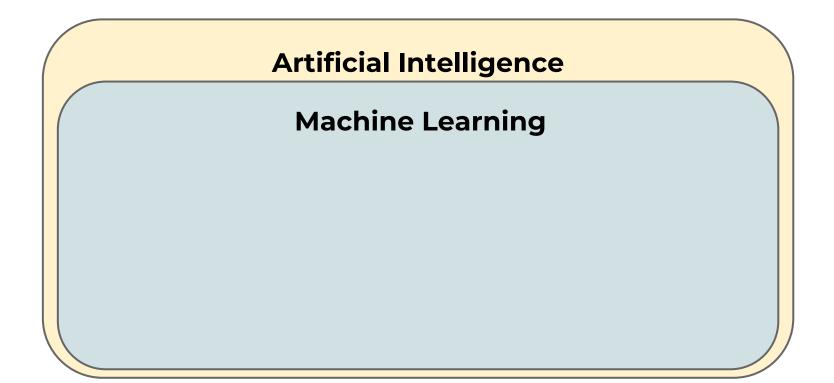
- Intelligence demonstrated by machines.
 - What is "intelligence"?
 - How to test for "intelligence"?

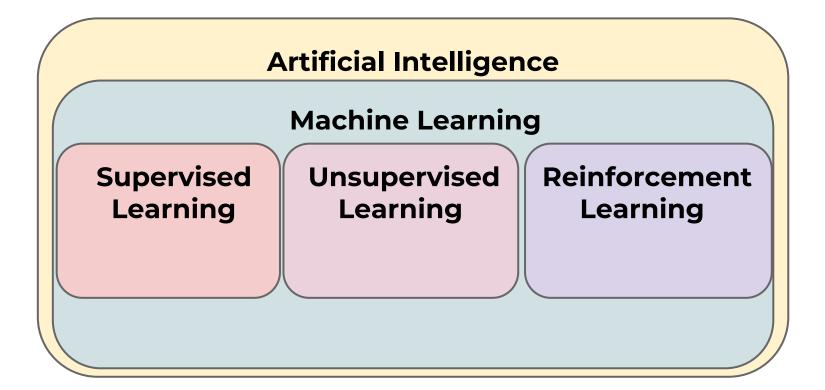
Artificial Intelligence

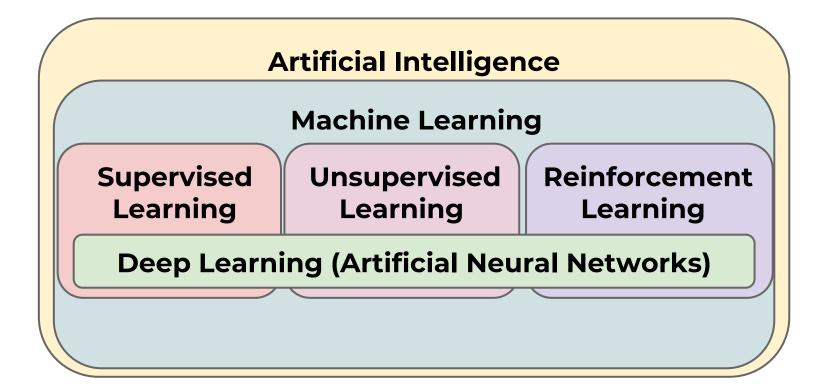
- General Artificial Intelligence
 - Human level (or better) intelligence in multiple domains.
- Narrow Artificial Intelligence
 - Human level intelligence in a specific domain (e.g. Chatbot, Image Recognition, etc...)

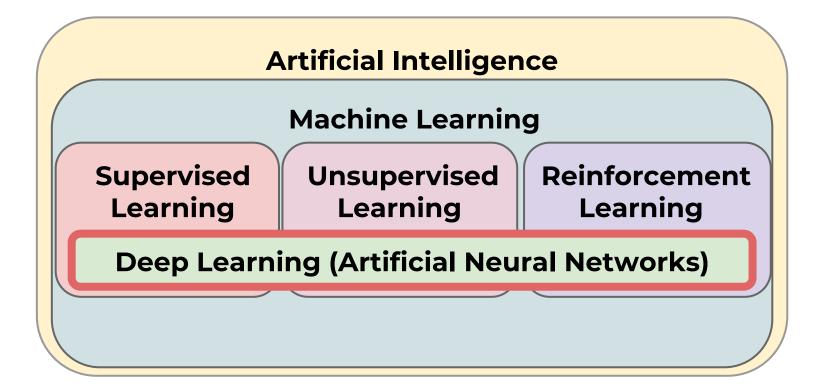
Artificial Intelligence

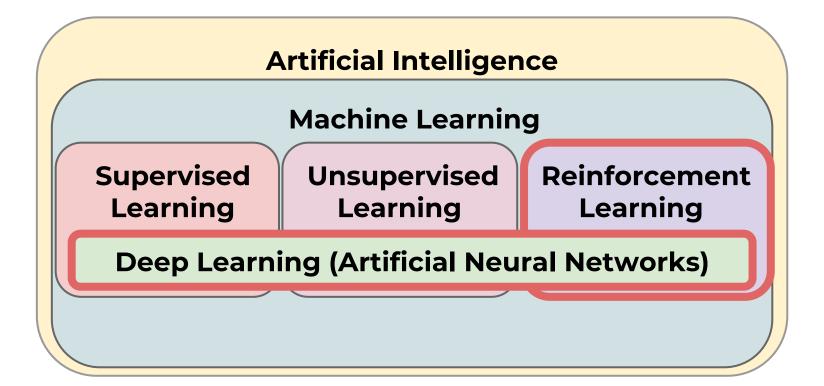
• What subdomains are necessary to create artificial intelligence?

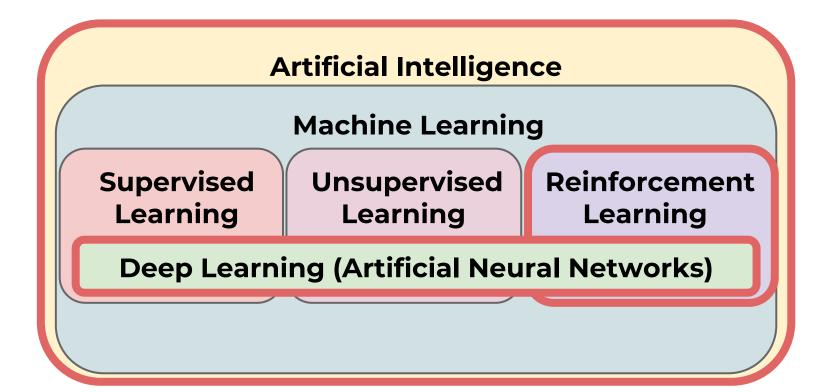












- Understand Machine Learning
 - Key Library Ideas (Pandas and Scikit-Learn)
 - Supervised Learning Process
 - Deep Learning (ANN and CNN)

Knowledge Path for Artificial Intelligence:

 Understand Reinforcement Learning
 Agent, Environment, and Policy
 Tabular Q-Learning

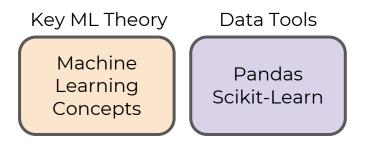
 Knowledge Path for Artificial Intelligence:

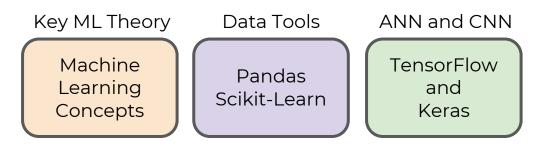
 Combine Deep Learning and Reinforcement Learning
 Combine ANN with Q-Learning

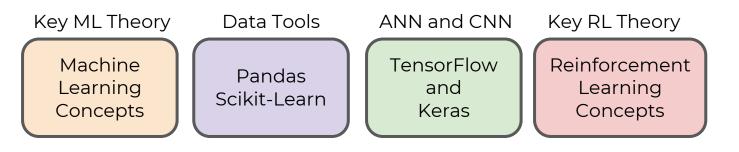
• Knowledge Path for Artificial Intelligence:

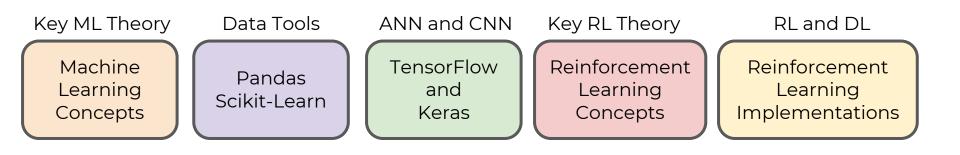
Key ML Theory

Machine Learning Concepts









Let's get started!

Environment Setup

Environment Set-up

• We use a wide variety of libraries in this course, so we'll show you how to set-up a separate virtual environment with Anaconda in order to **pip install** the libraries later on, including gym and tensorflow.

Environment Set-up

- The easiest way to do this is through the command line.
- Let's use the:
 - \circ Windows
 - Anaconda Prompt
 - MacOS/Linux
 - Terminal

Environment Set-up

Machine Learning Supervised Pathway

- Machine learning in general is the study of statistical computer algorithms that improve automatically through data.
- This means unlike typical computer algorithms that rely on human input for what approach to take, ML algorithms infer best approach from the data itself.

- Machine learning is actually a subset of Artificial Intelligence.
- ML algorithms are not explicitly programmed on which decisions to make.
- Instead the algorithm is designed to infer from the data the most optimal choices to make.

• What kinds of problems can ML solve?

- Credit Scoring
- Insurance Risk
- Price Forecasting
- Spam Filtering
- Customer Segmentation
- Much more!

"Standard" ML Pathway

Problem to Solve





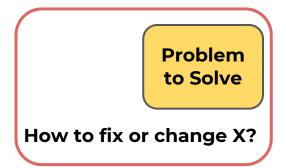


How to fix or change X?





How does a change in X affect Y?





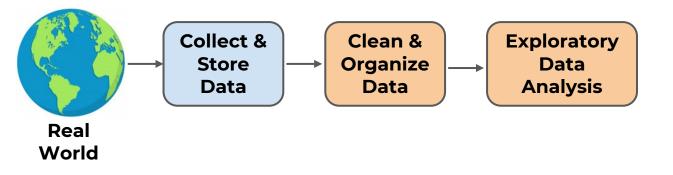


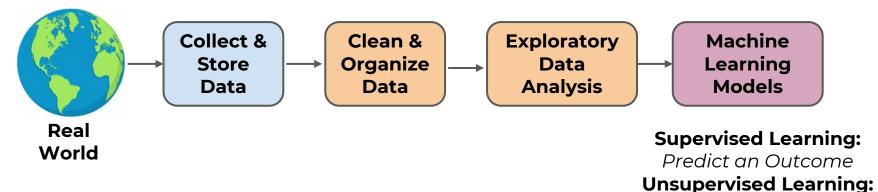
How does a change in X affect Y?



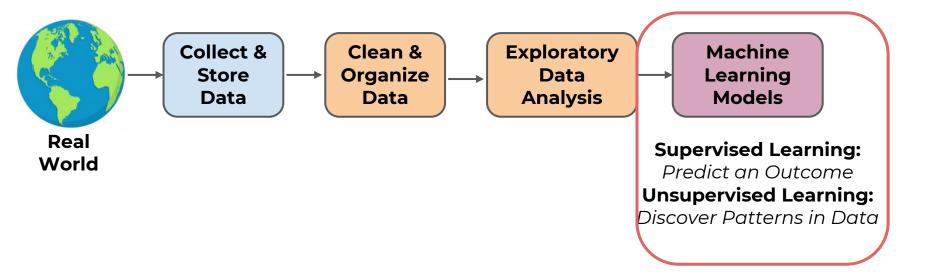








Discover Patterns in Data



Why Machine Learning?

• Structure of ML Problem framing:

- Given features from a data set obtain a desired label.
- ML algorithms are often called "estimators" since they are estimating the desired **label** or output.

- How can ML be so robust in solving all sorts of problems?
- Machine learning algorithms rely on data and a set of statistical methods to learn what features are important in data.

- Simple Example:
 - Predict the price a house should sell at given its current features (Area,Bedrooms,Bathrooms,etc...)

 House Price Prediction

 Typical Algorithm
 Human user defines an algorithm to manually set values of importance for each feature.

 House Price Prediction

 ML Algorithm
 Algorithm automatically determines importance of each feature from existing data

- Why machine learning?
 - Many complex problems are only solvable with machine learning techniques.
 - Problems such as spam email or handwriting identification require ML for an effective solution.

- Why not just use machine learning for everything?
 - Major caveat to effective ML is good data.
 - Majority of development time is spent cleaning and organizing data, **not** implementing ML algorithms.

 Do we develop our own ML algorithms?

 Rare to have a need to manually develop and implement a new ML algorithm, since these techniques are well documented and developed.

• Let's continue this discussion by exploring the types of machine learning algorithms!

Types of Machine Learning

- There are three main types of Machine Learning:
 - Supervised Learning
 - Unsupervised Learning
 - Reinforcement Learning

- Supervised Learning

 Using historical and labeled data, the machine learning model predicts a value.
- Unsupervised Learning
 - Applied to **unlabeled** data, the machine learning model discovers possible patterns in the data.

Supervised Learning

 Requires historical labeled data:
 Historical

- Known results and data from the past.
- Labeled
 - The desired output is known.

Supervised Learning

 Two main label types
 Categorical Value to Predict
 Classification Task
 Continuous Value to Predict

Regression Task

- Supervised Learning

 Classification Tasks
 - Predict an assigned category
 - Cancerous vs. Benign Tumor
 - Fulfillment vs. Credit Default
 - Assigning Image Category

 Handwriting Recognition

- Supervised Learning

 Regression Tasks
 Predict a continuous value
 - Future prices
 - Electricity loads
 - Test scores

• Unsupervised Learning

- Group and interpret data without a historical label.
- Example:
 - Clustering customers into separate groups based off their behaviour features.

 Unsupervised Learning

 Major downside is because there was no historical "correct" label, it is much harder to evaluate performance of an unsupervised learning algorithm.

• Reinforcement Learning

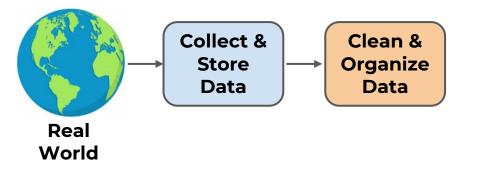
 Finally, before we dive into coding and linear regression in the next section, let's have a deep dive into the entire Supervised Machine Learning process to set ourselves up for success!

Supervised Machine Learning Process

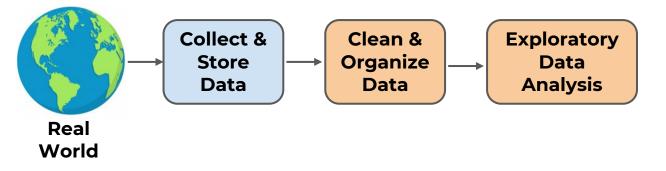
• Machine Learning Pathway



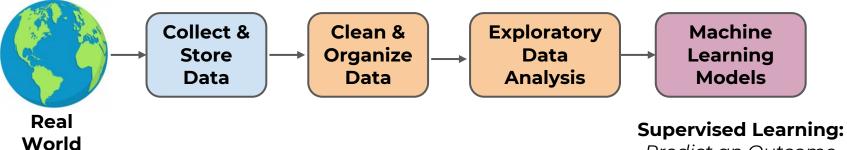
• Machine Learning Pathway



• Machine Learning Pathway



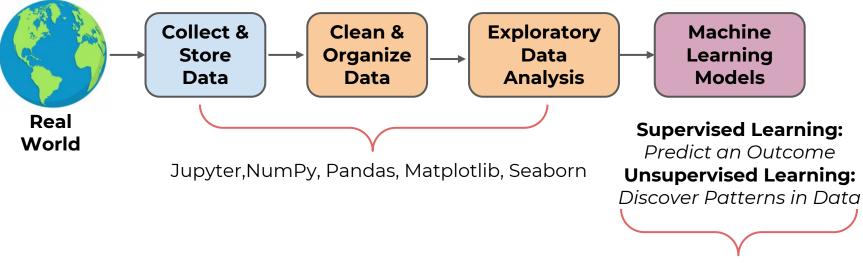
• Machine Learning Pathway



Predict an Outcome Unsupervised Learning:

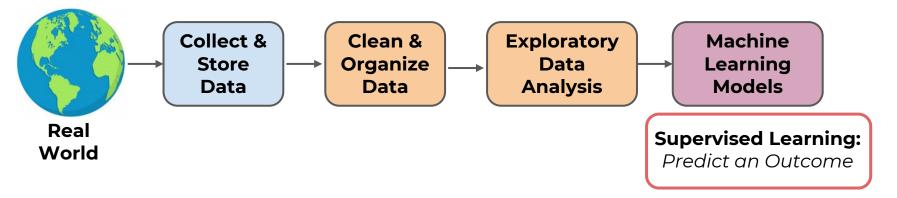
Discover Patterns in Data

• Machine Learning Pathway

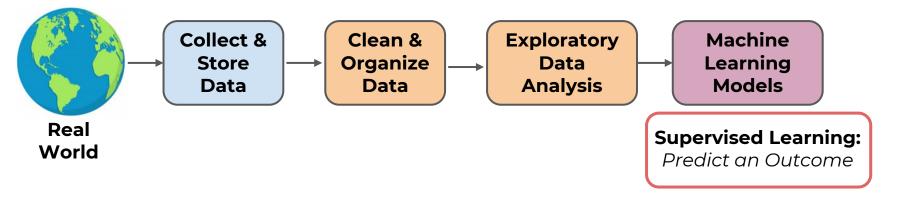


Scikit-learn or TensorFlow

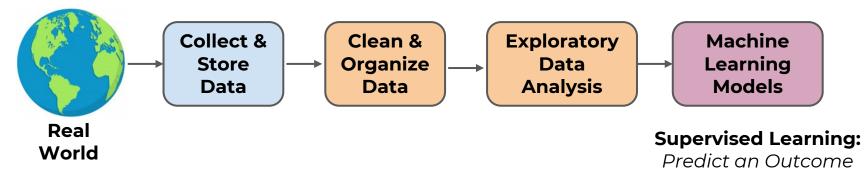
• Machine Learning Pathway



• ML Process : Supervised Learning Tasks



• Predict price a house should sell at.



- **Supervised** Machine Learning Process
- Start with collecting and organizing a data set based on history:

•			
Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• **Historical labeled** data on previously sold houses.

Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• If a new house comes on the market with a known Area, Bedrooms, and Bathrooms: *Predict what price should it sell at.*

Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

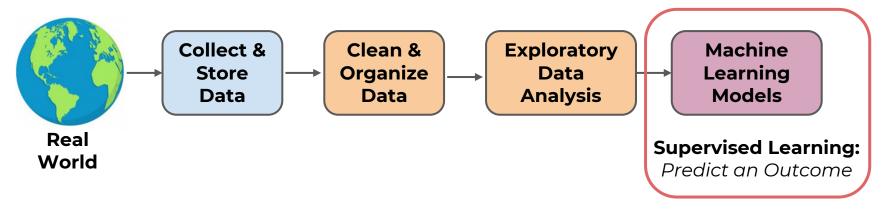
- Data Product:
 - Input house features
 - Output predicted selling price

· · ·	-		
Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

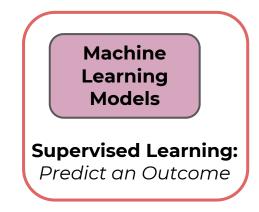
• Using **historical**, **labeled** data predict a future outcome or result.

Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Predict price a house should sell at.



• Predict price a house should sell at.



• Predict price a house should sell at.

Machine Learning Models

Supervised Learning: *Predict an Outcome*

• Predict price a house should sell at.

Machine Learning Models

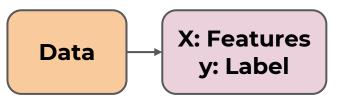
Supervised Learning: *Predict an Outcome*

Data

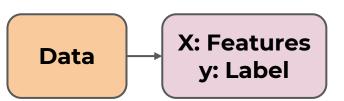
• Supervised Machine Learning Process



• **Supervised** Machine Learning Process

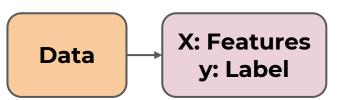


• **Supervised** Machine Learning Process



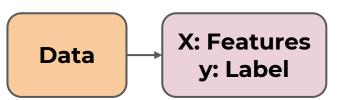
Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Label is what we are trying to predict



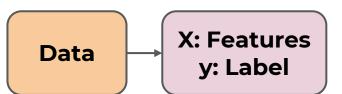
Area m ²	Bedrooms	Bathrooms	Price
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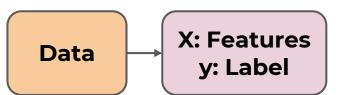
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200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Features are known characteristics or components in the data



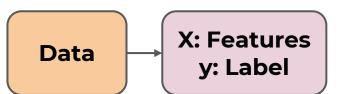
Area m ²	Bedrooms	Bathrooms	Price
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• Features are known characteristics or components in the data



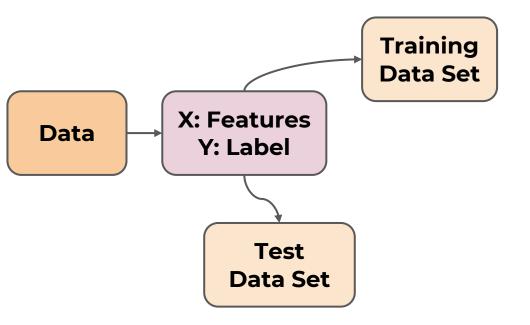
Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Features and Label are identified according to the problem being solved.

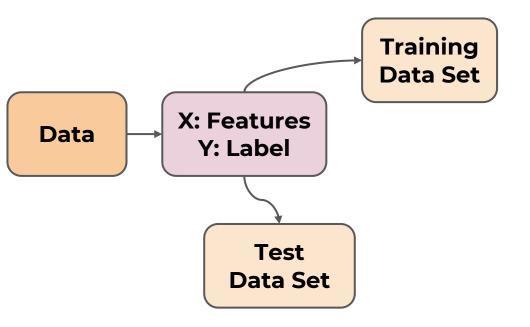


Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

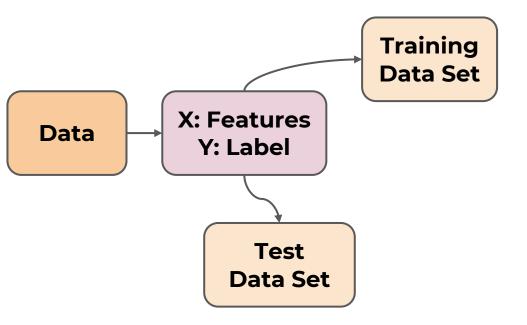
• Split data into training set and test set



• Later on we will discuss cross-validation



• Why perform this split? How to split?



• Why perform this split? How to split?

Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• How would you judge a human realtor's performance?



Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Ask a human realtor to take a look at historical data...



Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Then give her the features of a house and ask her to predict a selling price.



Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

 But how would you measure how accurate her prediction is? What house should you choose to test on?



Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

 You can't judge her based on a new house that hasn't sold yet, you don't know it's true selling price!



Area m ²	Bedrooms Bathrooms		Price	
200	3	2	\$500,000	
190	2	1	\$450,000	
230	3	3	\$650,000	
180	1	1	\$400,000	
210	2	2	\$550,000	

 You shouldn't judge her on data she's already seen, she could have memorized it!



Area m ²	Bedrooms	Bathrooms	Price	
200	3	2	\$500,000	
190	2	1	\$450,000	
230	3	3	\$650,000	
180	1	1	\$400,000	
210	2	2	\$550,000	

• Thus the need for a Train/Test split of the data, let's explore further...



Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

 We already organized the data into Features (X) and a Label (y)

Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Now we will split this into a training set and a test set:

	Area m ²	Bedrooms	Bathrooms	Price
	200	3	2	\$500,000
TRAIN	190	2	1	\$450,000
	230	3	3	\$650,000
	180	1	1	\$400,000
	210	2	2	\$550,000

• Now we will split this into a training set and a test set:

	Area m ²	Bedrooms	Bathrooms	Price
	200	3	2	\$500,000
TRAIN	190	2	1	\$450,000
	230	3	3	\$650,000
TEST	180	1	1	\$400,000
	210	2	2	\$550,000

• Notice how we have 4 components



• Let's go back to fairly testing our human realtor....



Area m ²	Bedrooms	Bathrooms	Price
200	3	2	\$500,000
190	2	1	\$450,000
230	3	3	\$650,000
180	1	1	\$400,000
210	2	2	\$550,000

• Let's go back to fairly testing our human realtor....

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N	

	Area m ²	Bedrooms	Bathrooms	Price
	200	3	2	\$500,000
TRAIN	190	2	1	\$450,000
	230	3	3	\$650,000
TEST	180	1	1	\$400,000
	210	2	2	\$550,000

• Let her study and learn on the training set getting access to both X and y.

			Area m ²	Bedrooms	Bathrooms	Price
5	2		200	3	2	\$500,000
		TRAIN	190	2	1	\$450,000
Π		230	3	3	\$650,000	

• After she has "learned" about the data, we can test her skill on the test set.

		[Area m ²	Bedrooms	Bathrooms
5	2	TEST	180	1	1
			210	2	2
Π	11	-			

• Provide only the X test data and ask for her predictions for the sell price.

			Area m ²	Bedrooms	Bathrooms
5	2	TEST	180	1	1
			210	2	2
	11				

• This is new data she has never seen before! She has also never seen the real sold price.

		Area m ²	Bedrooms	Bathrooms
5	TEST	180	1	1
	1201	210	2	2

• Ask for predictions per data point.

5	
Ν	π

Predictions	Area m ²	Bedrooms	Bathrooms
\$410,000	180	1	1
\$540,000	210	2	2

• Then bring back the original prices.



Predictions	Area m ²	Bedrooms	Bathrooms	Price
\$410,000	180	1	1	\$400,000
\$540,000	210	2	2	\$550,000

• Finally compare predictions against true test price.



Predictions	Price
\$410,000	\$400,000
\$540,000	\$550,000

• This is often labeled as **ŷ** compared again **y**



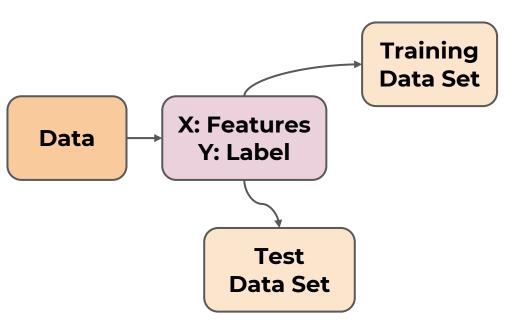


• Later on we will discuss the many methods of evaluating this performance!

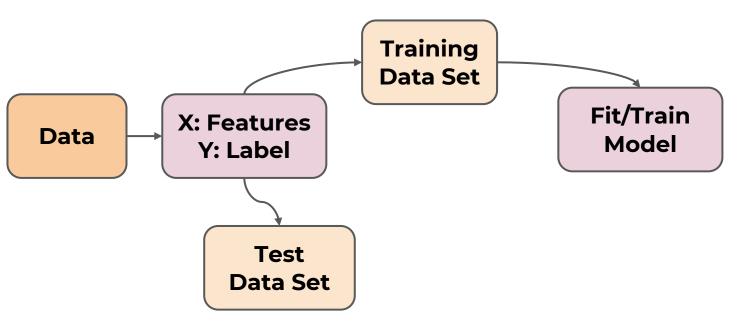


Predictions	Price
\$410,000	\$400,000
\$540,000	\$550,000

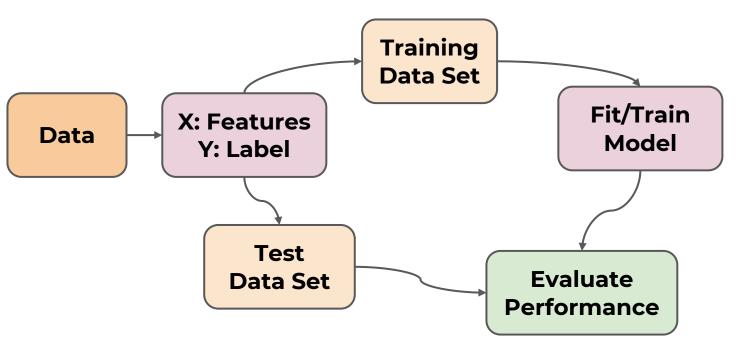
• Split Data



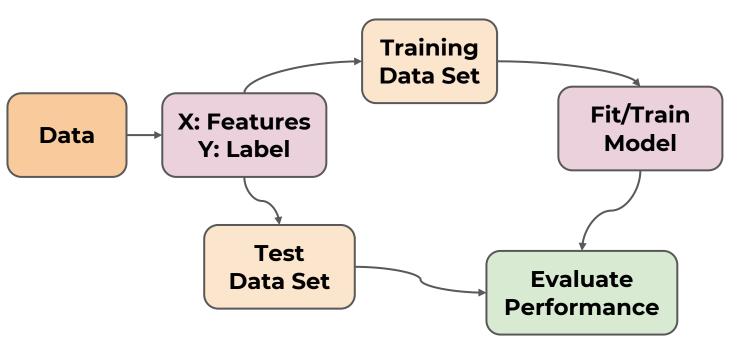
• Split Data, Fit on Train Data



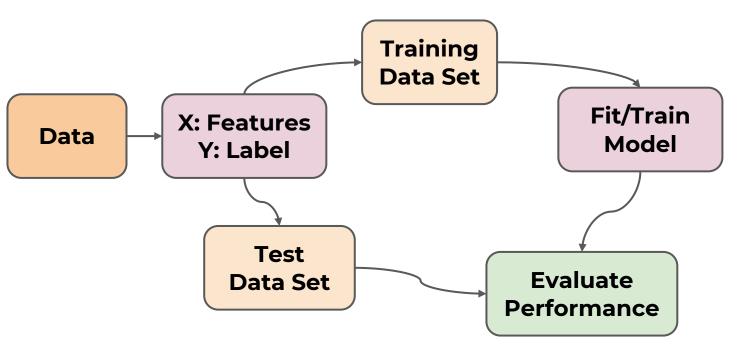
• Split Data, Fit on Train Data, Evaluate Model



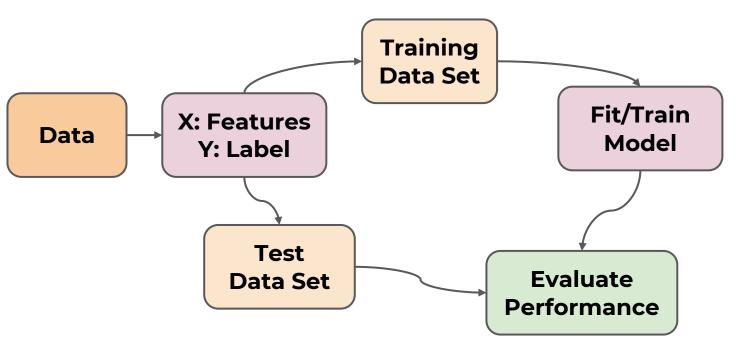
• What happens if performance isn't great?



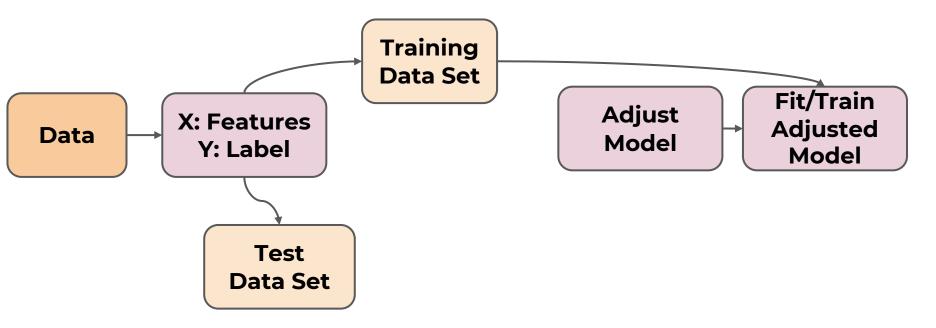
• We can adjust model **hyperparameters**



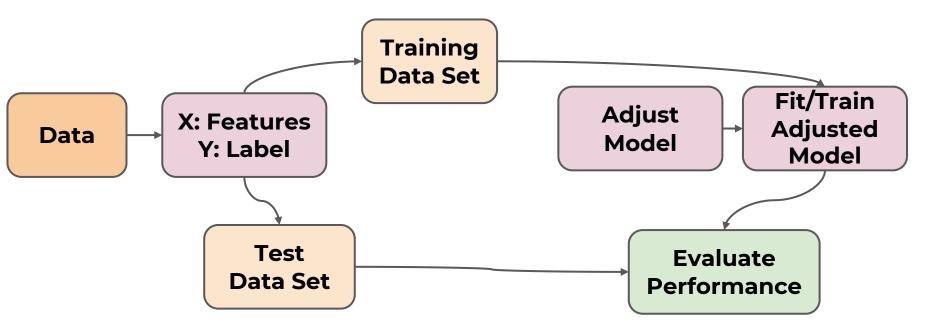
• Many algorithms have adjustable values



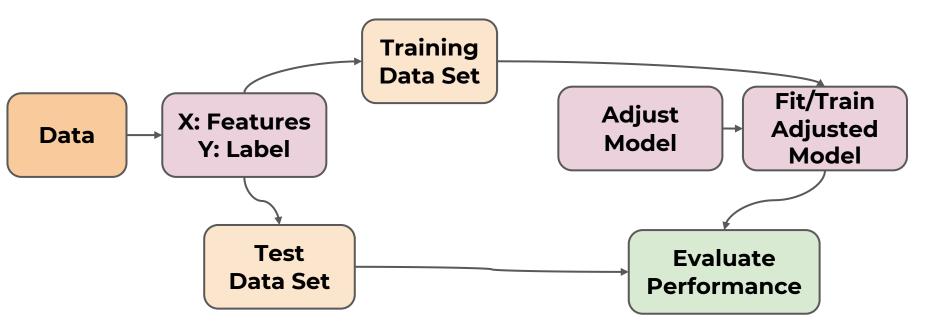
• Many algorithms have adjustable values



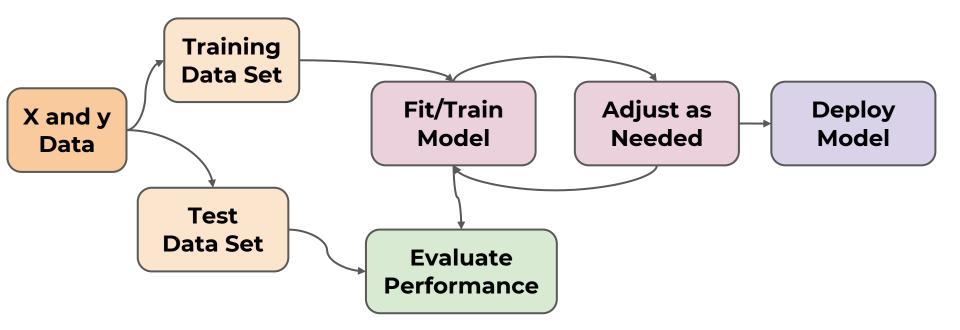
• Evaluate adjusted model



• Can repeat this process as necessary



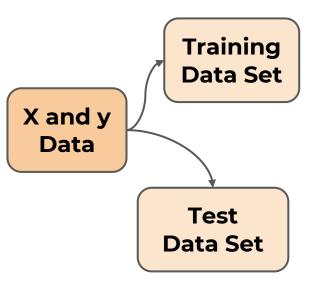
• Full and Simplified Process



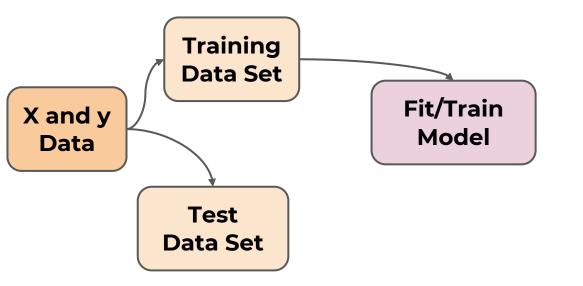
• Get X and y data



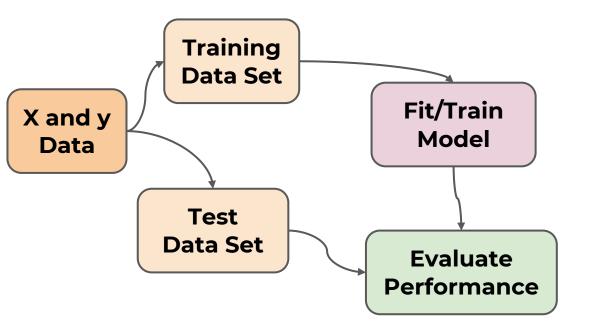
• Split data for evaluation purposes



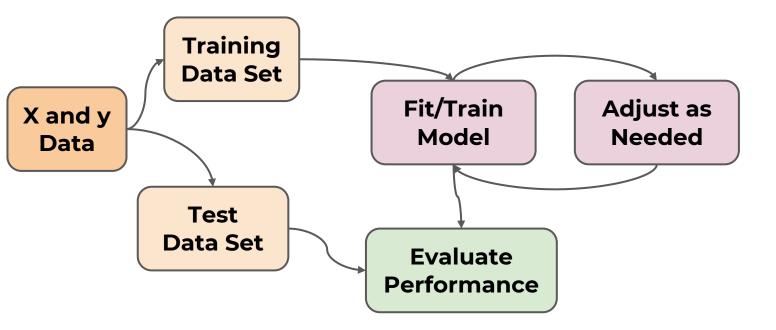
• Fit ML Model on Training Data Set



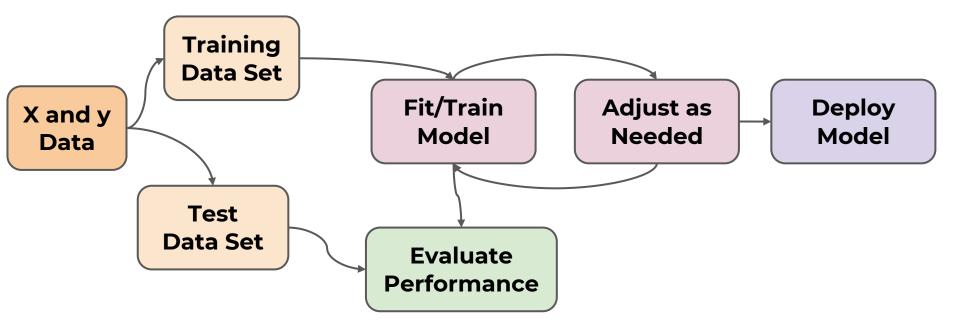
• Evaluate Model Performance



• Adjust model hyperparameters as needed

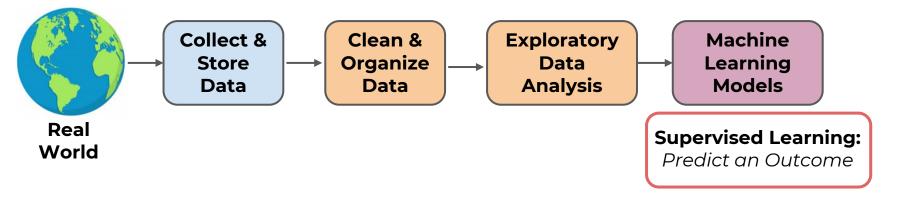


• Deploy model to real world



Machine Learning

• ML Process : Supervised Learning Tasks



ML Pathway

