

# *Artificial Intelligence: Modeling Human Intelligence with Networks*

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# Machine Learning and Linear Classifiers

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# Machine Learning and Classification

## Machine Learning

- Studies how to automatically learn to make accurate predictions based on past observations.
- Two kinds kinds of problems in ML:
  - ▶ Supervised Learning: Observations + desired outputs. Ex.: Spam detection.
  - ▶ Unsupervised Learning: Only the observations. Ex.: Clustering.
- Our focus in this course: Supervised Learning

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# Classification Example

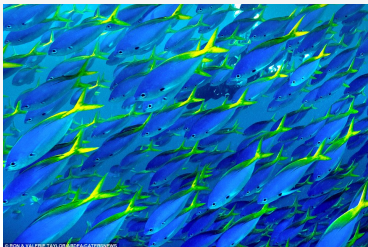
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- You are given a bunch of fish!



# Classification Example

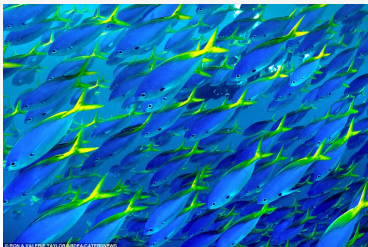
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(a) “Look, these ones are *bass*...”

# Classification Example

- You are given a bunch of fish! And someone tells you:



(a) “Look, these ones are *bass*...



(b) and these are *tuna*.”

- He asks you, “Would you mind tell me which kind of fish Jim is bringing?”.
- What would you do?

# Classification Example

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- Possible steps:

- ① Extract information about the fish you have.

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- Possible steps:

- ① Extract information about the fish you have. *What info?*

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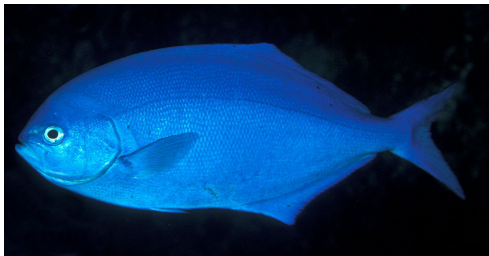
- ① Extract information about the fish you have. *What info? How about possible info?*
- ② Propose a rule based on that information! *Which rule?*
- ③ (Pray so that the you were given everything you needed...)
- ④ Classify the new fish!

# Classification Example

- Possible steps:

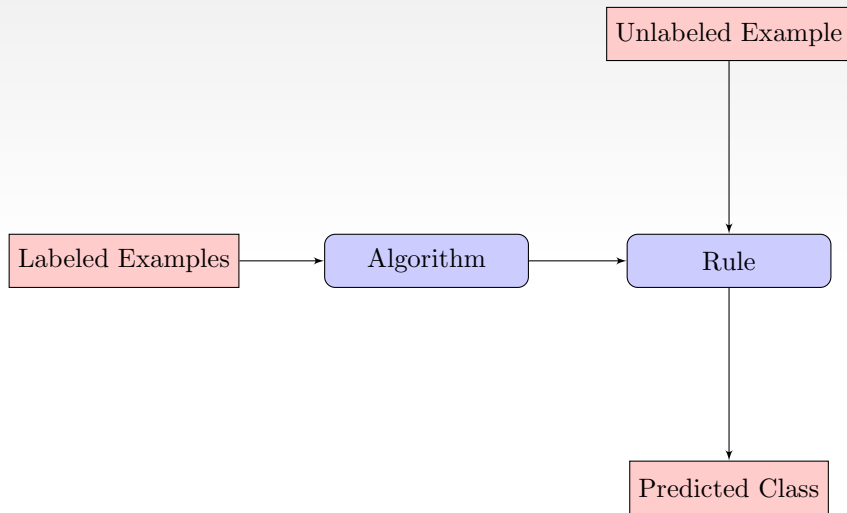
- ① Extract information about the fish you have. *What info? How about possible info?*
- ② Propose a rule based on that information! *Which rule?*
- ③ (Pray so that the you were given everything you needed...)
- ④ Classify the new fish!

- Then, Jim brings you this:

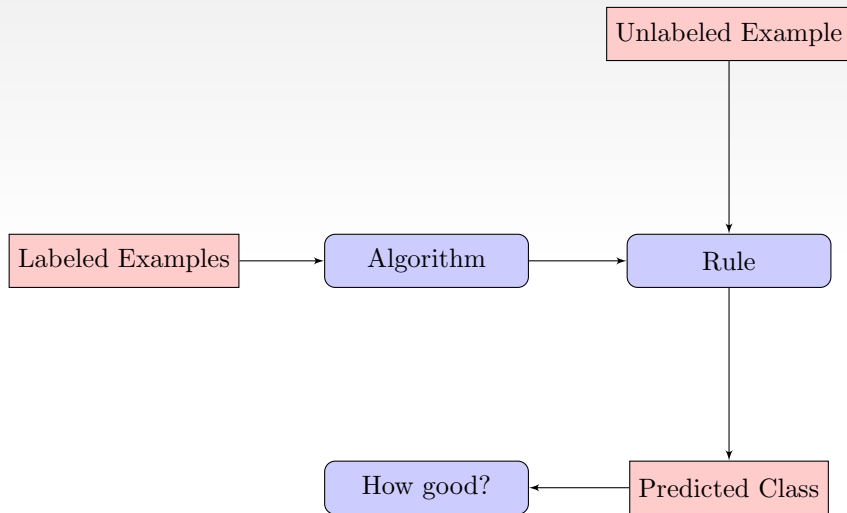


- What do you do?

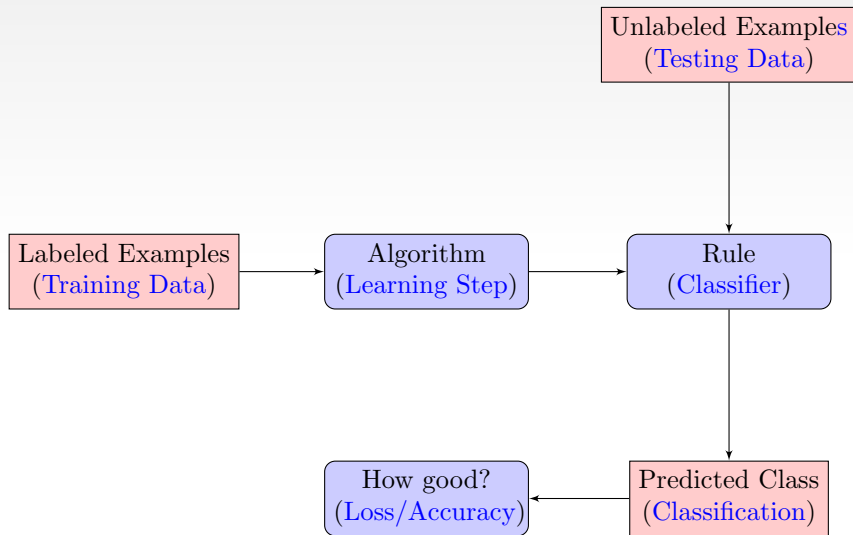
# Classification Pipeline



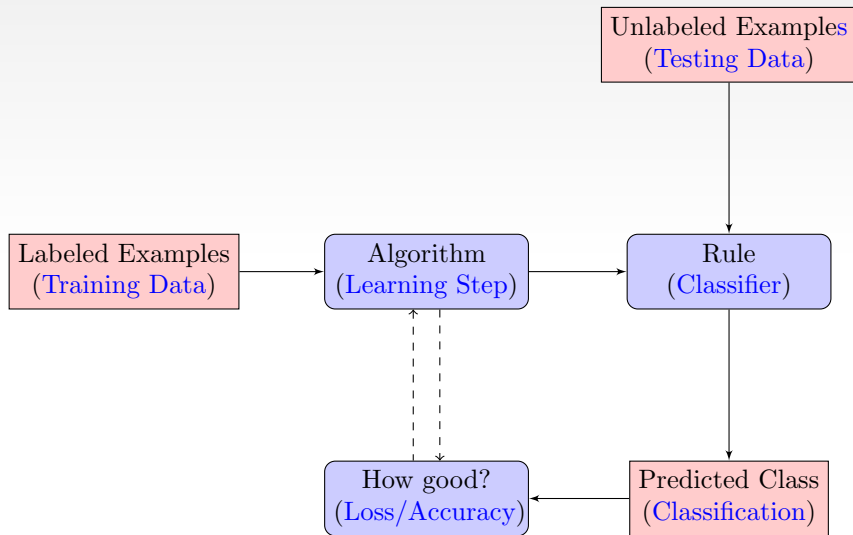
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# Classification Pipeline

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This week for us!

- **Data** (*for testing and for training*) – Points in  $\mathbb{R}^2$ .
- **Algorithm** – Perceptron (tomorrow!)
- **Classification** – In two classes (tomorrow), more than two (Thursday).
- **Classifier** – Decision boundaries given by lines in  $\mathbb{R}^2$ .
- **Loss Function** – Zero-one loss.

# Linear Classifiers

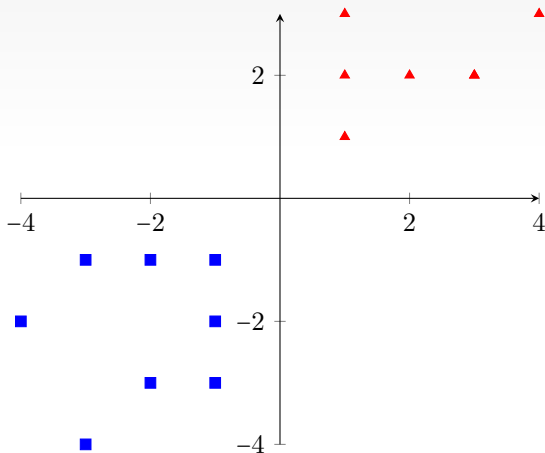
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- Our problem:  $m$  points in the plane, i.e., points  $p^i = [x_1^i, x_2^i]$ .



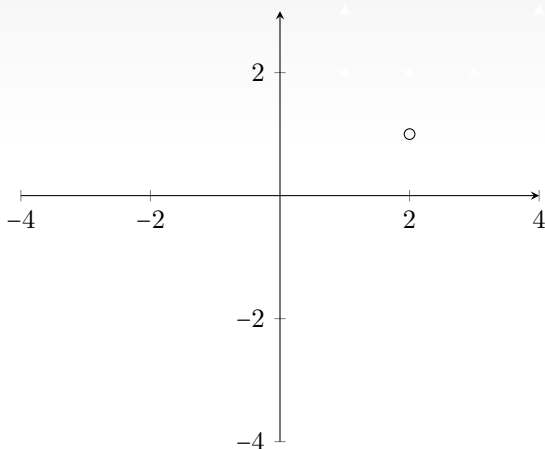
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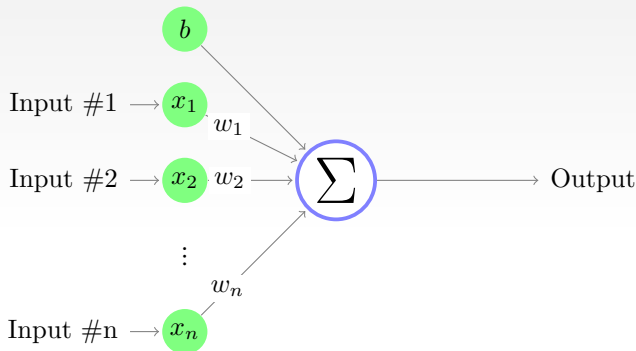


# Linear Classifiers

- Now, what is the class of this point? Blue square or red triangle?



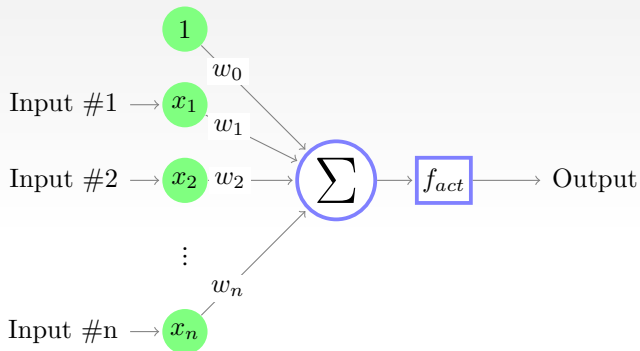
# The Neuron Model!



$$x = [x_1, x_2, \dots, x_n], \quad w = [w_1, w_2, \dots, w_m]$$

$$f_{act}(x \cdot w + b) = \text{sign}(x \cdot w + b) = \begin{cases} +1, & x \cdot w + b > 0 \\ -1, & x \cdot w + b < 0 \end{cases}$$

# The Neuron Model!

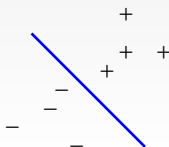


$$x = [1, x_1, x_2, \dots, x_n], \quad w = [w_0, w_1, w_2, \dots, w_m]$$

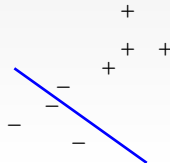
$$f_{act}(x \cdot w) = \text{sign}(x \cdot w) = \begin{cases} +1, & x \cdot w > 0 \\ -1, & x \cdot w < 0 \end{cases}$$

# Loss Function

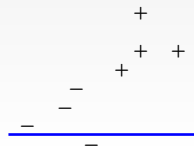
- Which of the following is the best?



(a)



(b)

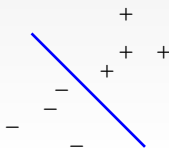


(c)

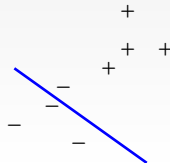
- Why?

# Loss Function

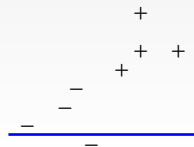
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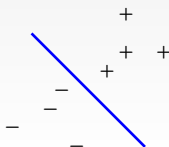


(c)

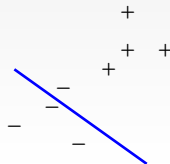
- Why? Which one is the second best?

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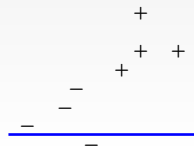
- Which of the following is the best?



(a)



(b)



(c)

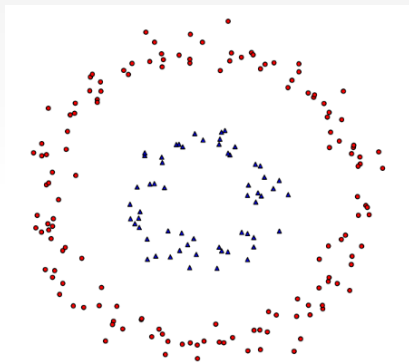
- Why? Which one is the second best?
- Zero-one loss: # Number of misclassified examples.

$$\mathcal{L}_w(X) = \sum_{i=1}^n [f_w(x^i) \neq y_i]$$

$$[\cdot] : \{\text{False}, \text{True}\} \rightarrow \{0, 1\}$$

# Non-Linearly separable data

- What is decision boundary of the this:



- Could I use straight lines to separate it??
- What is Non-Linearly separable dataset?



# Python Notebooks!

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