

# Classification



# Classification Introduction



- Credit Scoring
  - Input: income, savings, profession, age and past financial history
  - Output: accept or refuse
- Medical Diagnosis
  - Input: current symptoms, age, gender and past medical history
  - Output: type of diseases
- Handwritten Character Recognition
- Face Recognition
  - Input: image of a face
  - Output: person

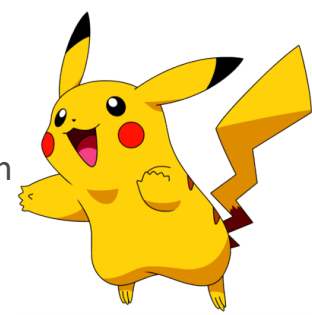
# Example Application



$$f(\text{Pikachu}) = \text{Electric} \quad f(\text{Turtle}) = \text{Water} \quad f(\text{Rhydon}) = \text{Grass}$$

## Example Application, Continued

Pokémon games (*not* Pokémon cards or Pokémon Go)



- **HP:** hit points or health. Defines how much damage a Pokémon can withstand before fainting
- **Attack:** the base modifier for normal attacks (e.g., Scratch, Punch)
- **Defense:** the base damage resistance against normal attacks
- **SP Attack:** the base modifier for special attacks (e.g., fire blast, bubble beam)
- **SP Def:** the base damage resistance against special attacks
- **Speed:** determines which Pokémon attacks first during each round

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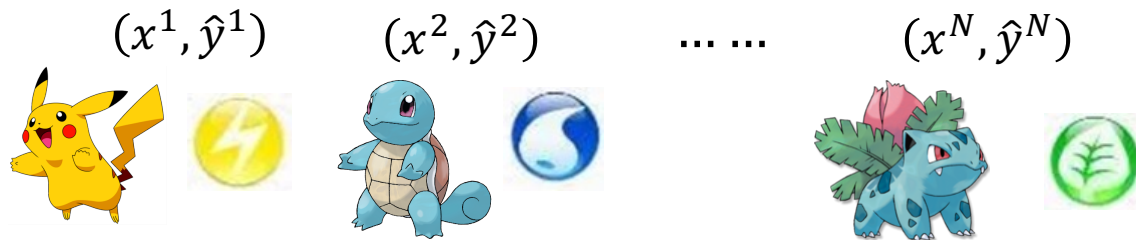
90

Can we predict the “type” of Pokémon based on the information?



# How to do Classification

## Training Data for Classification



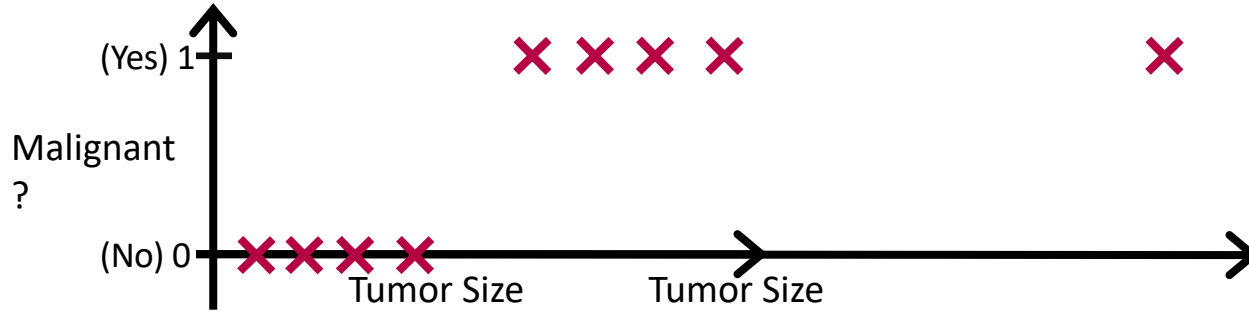
## Classification as Regression?

Binary classification as example

Training: Class 1 means the target is 1; Class 2 means the target is -1

Testing: closer to 1  $\rightarrow$  class 1; closer to -1  $\rightarrow$  class 2

# Using Regression for Classification



Threshold classifier output  $h_{\theta}(x)$  at 0.5:

If  $h_{\theta}(x) \geq 0.5$ , predict “y = 1”

If  $h_{\theta}(x) < 0.5$ , predict “y = 0”

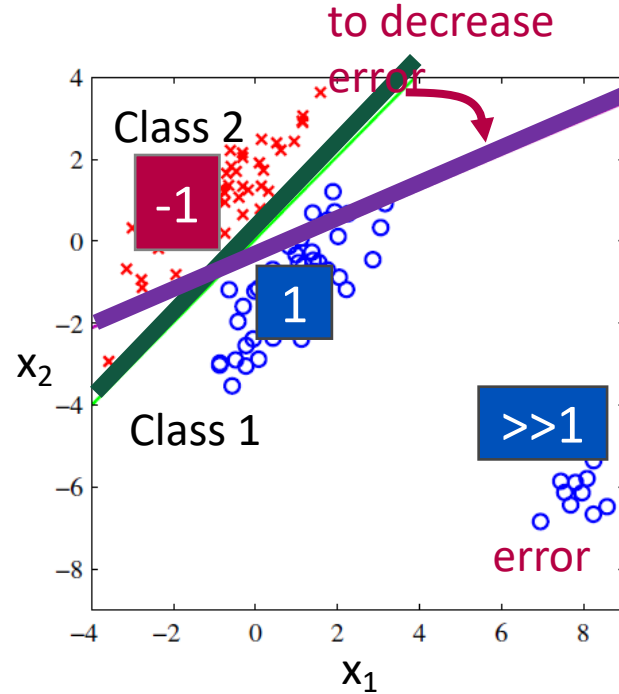
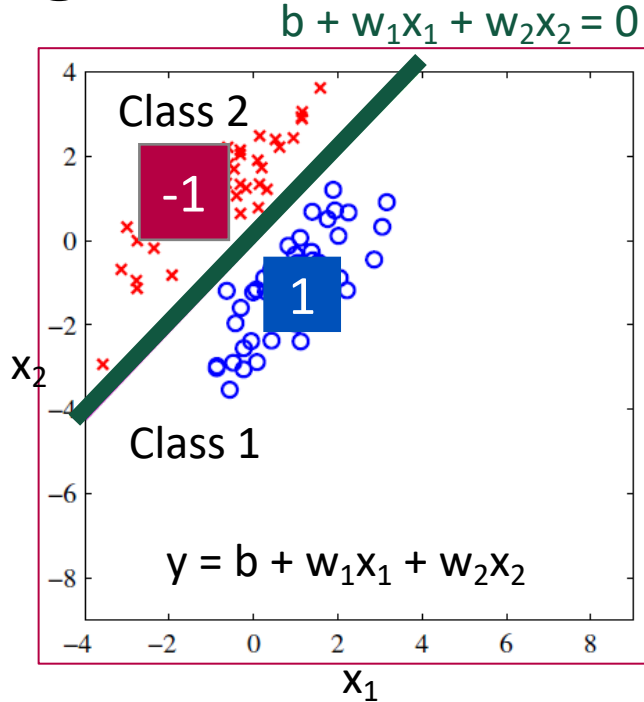
# Using Regression for Classification? Continued

## Multiple Classes:

Class 1 means  
the target is 1

Class 2 means  
the target is 2

Class 3 means  
the target is  
3 —problematic



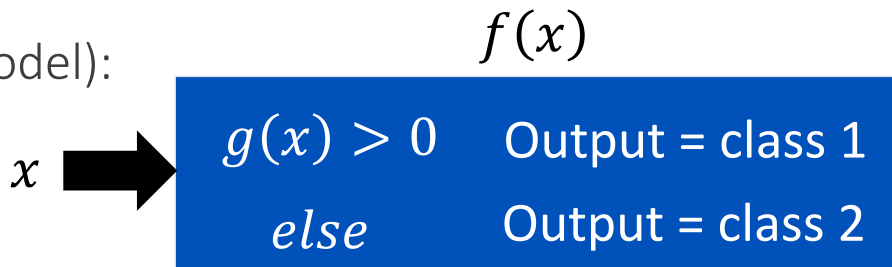
Penalize to the examples that are “too correct”.

(Bishop, 2006, p 186)



# Ideal Alternatives

- Function (Model):



- Loss function:

$$L(f) = \sum_n \delta(f(x^n) \neq \hat{y}^n)$$

The number of times  $f$  get incorrect results on training data.

- Find the best function:

Gradient  
Descent

