



Machine Learning: What's The Challenge?



Goals of the course

- Identify a machine learning problem
- Use basic machine learning techniques
- Think about your data/results



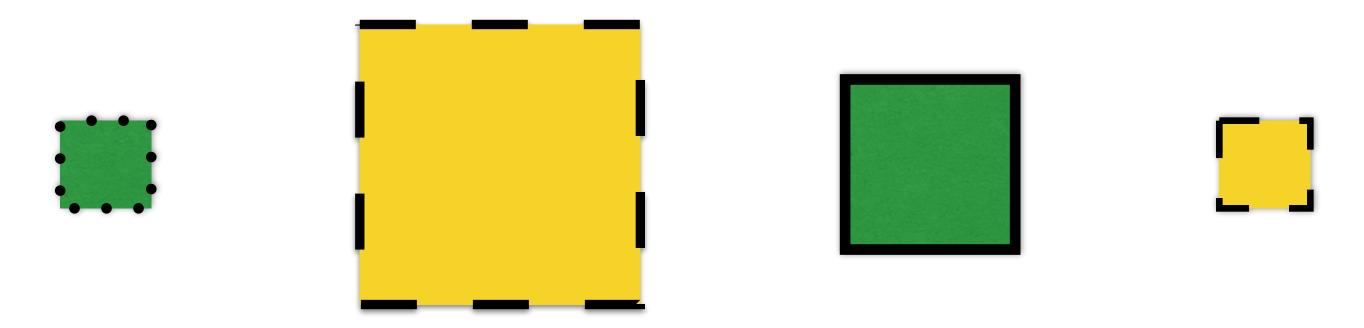
What is Machine Learning?

- Construct/use algorithms that learn from data
- More information
 Higher performance
- Previous solutions
 Experience



Example

- Label squares: size and edge —— color
- Earlier observations (labeled by humans):



• Task for computer = label unseen square:



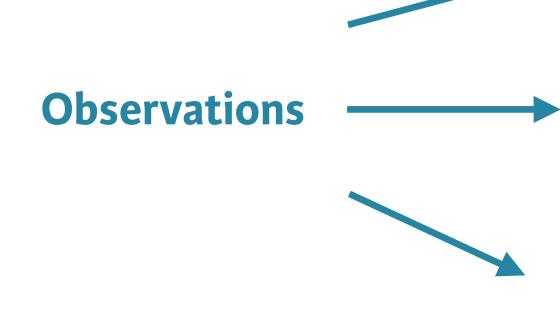
• Result: right or wrong!



Input Knowledge

In example: pre-labeled squares

In R - use data.frame()



Feat	ures	Label ↓
size	edge	color
small	dotted	green
big	striped	yellow
medium	normal	green

```
> squares <- data.frame(
    size = c("small", "big", "medium"),
    edge = c("dotted", "striped", "normal"),
    color = c("green", "yellow", "green"))</pre>
```

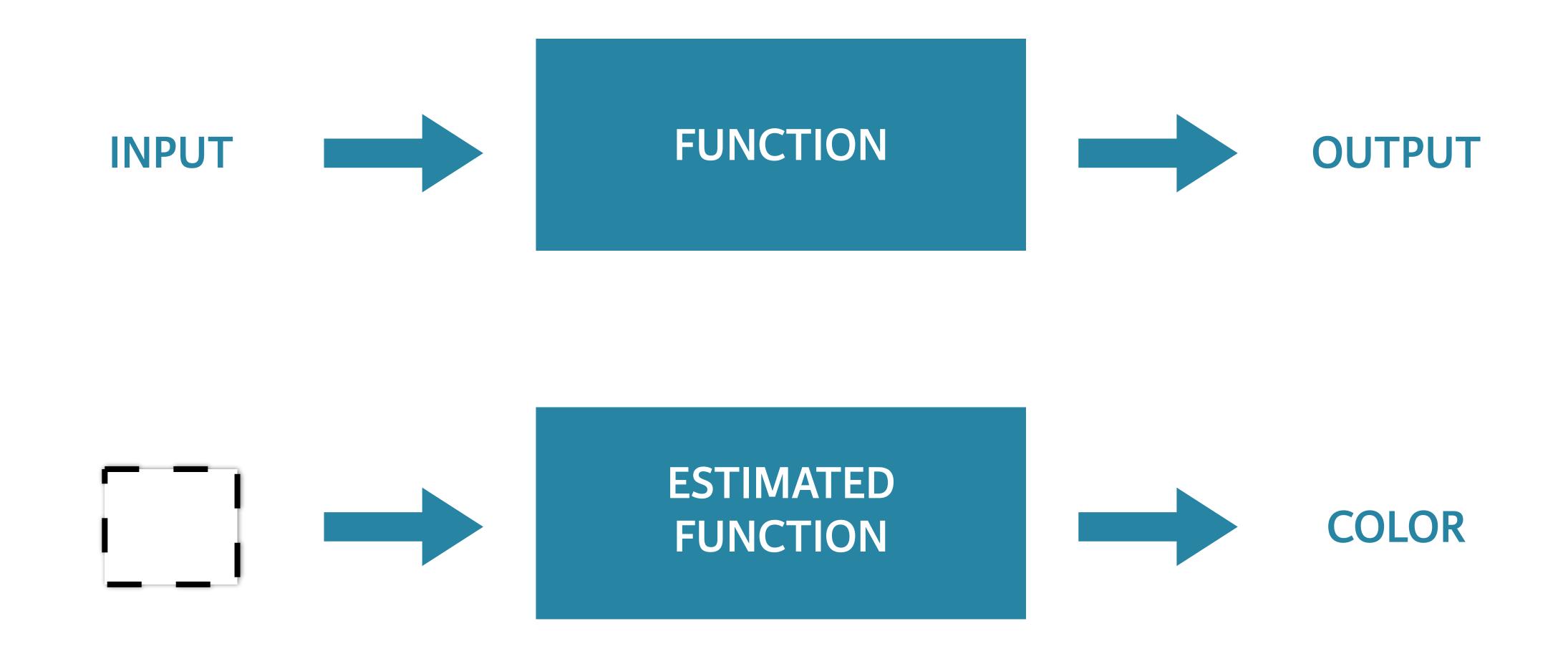


Data Frame Functions

> dim(squares) #Observations, #Features



Formulation





ML: What It Is Not

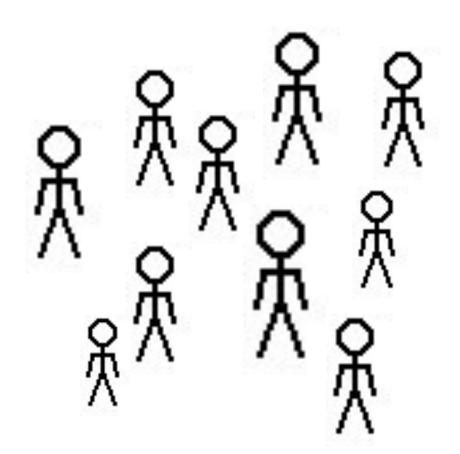
- Determining most occurring color
- Calculating average size

NOT Machine Learning

Goal: Building models for prediction!



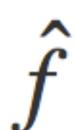
Regression



INPUT: Weight OUTPUT: Height



Estimated function:





Weight





More Applications!

- Shopping basket analysis
- Movie recommendation systems
- Decision making for self-driving cars
- and many more!





Let's practice!





Classification Regression Clustering



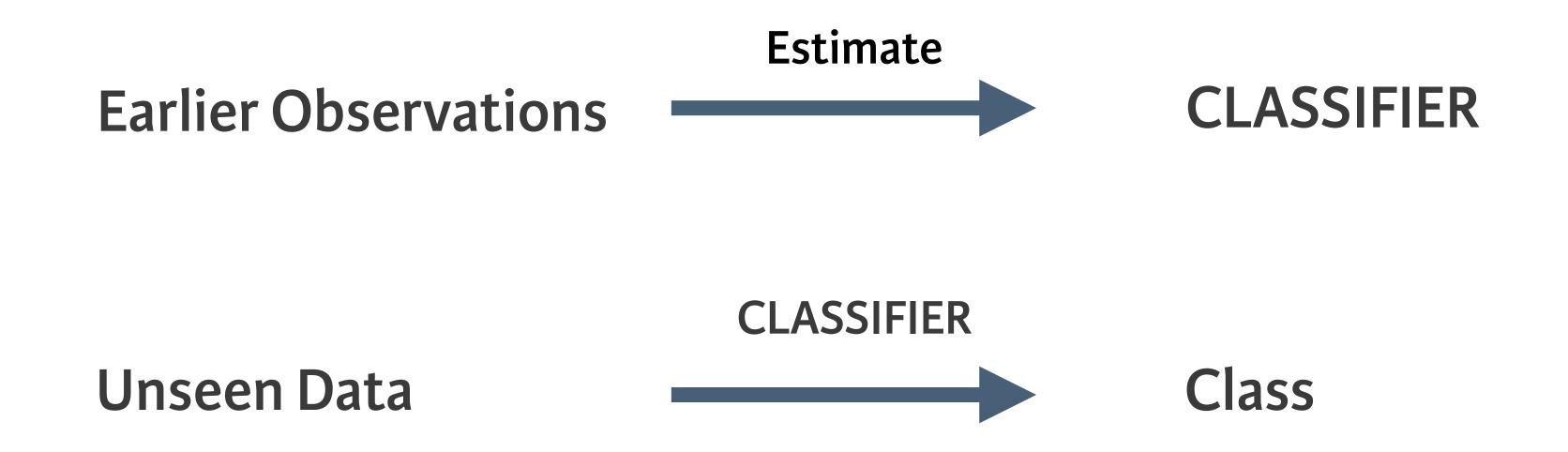
Common ML Problems

- Classification
- Regression
- Clustering



Classification Problem

Goal: predict category of new observation





Classification Applications

Medical Diagnosis
 Sick and Not Sick

Animal Recognition Dog, Cat and Horse

Important:

- Qualitative Output
- Predefined Classes

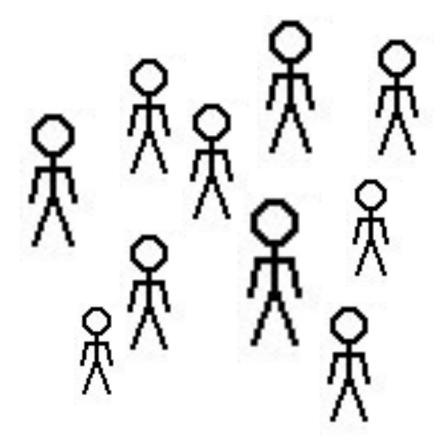


Regression



REGRESSION FUNCTION





- Relationship: Height Weight?
- Linear?



Regression Model

Fitting a linear function

Height $\approx \beta_0 + \beta_1 \times \text{Weight}$

- Predictor: Weight
- Response: Height
- Coefficients: β_0, β_1

Estimate on previous input-output

> lm(response ~ predictor)





Regression Applications

- Payments —— Credit Scores
- TimeSubscriptions
- Grades
 Landing a Job

- Quantitative Output
- Previous input-output observations



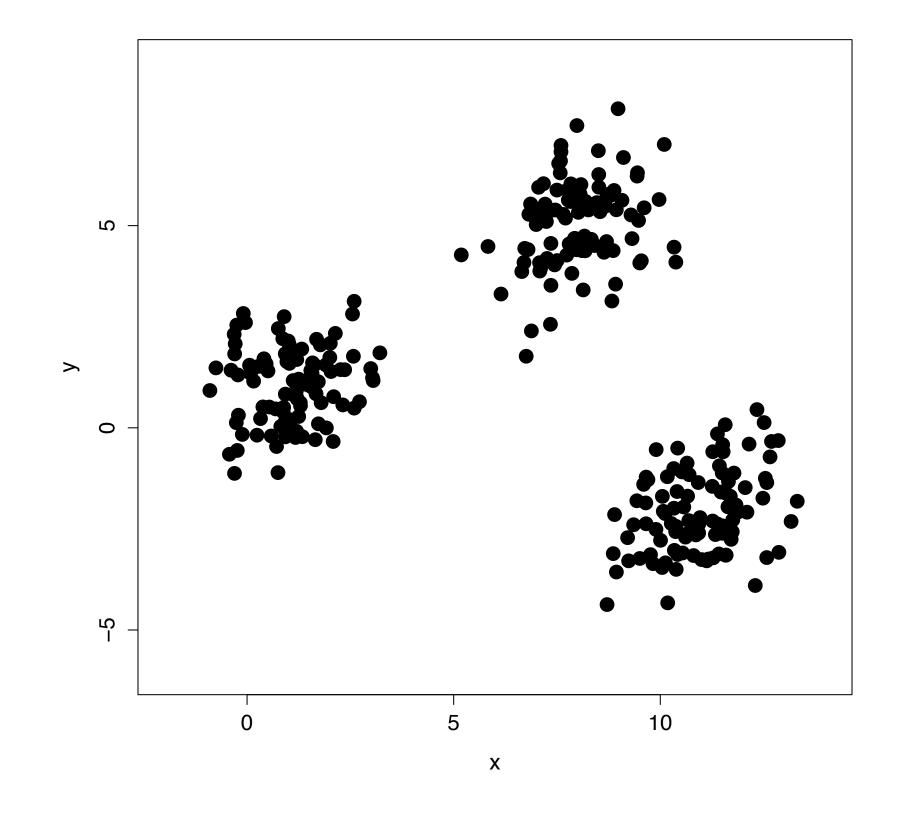
Clustering

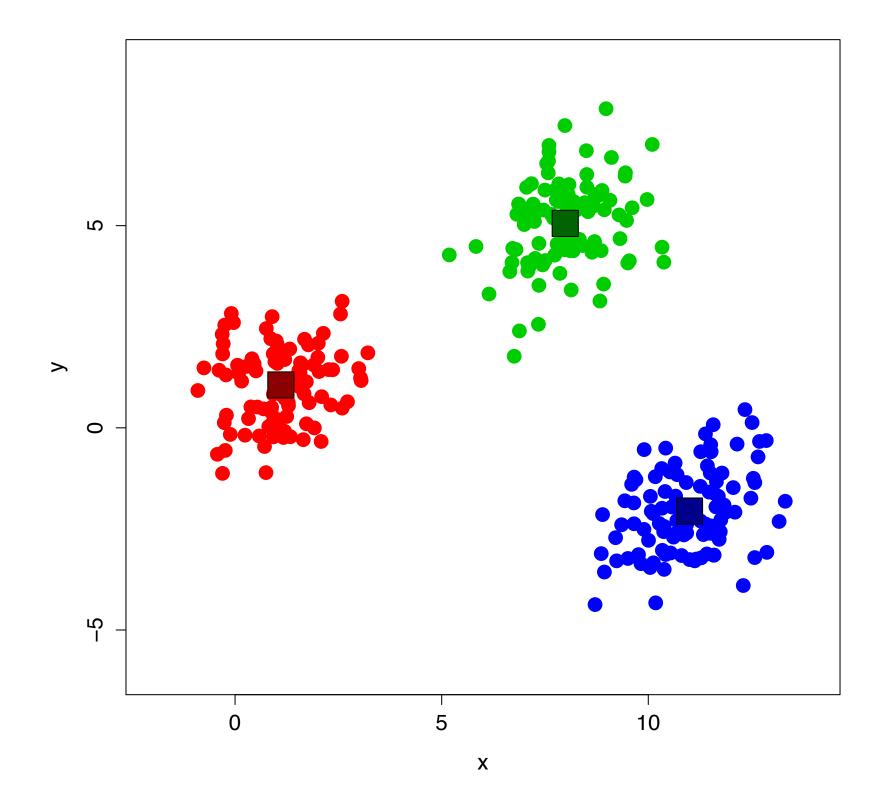
- Clustering: grouping objects in clusters
 - Similar within cluster
 - Dissimilar between clusters
- Example: Grouping similar animal photos
 - No labels
 - No right or wrong
 - Plenty possible clusterings



k-Means

Cluster data in k clusters!









Let's Practice





Supervised vs. Unsupervised



Machine Learning Tasks

Classification
Regression

quite similar

Clustering



Supervised Learning

Find: function **f** which can be used to assign a class or value to unseen observations.

Given: a set of labeled observations

Supervised Learning



Unsupervised Learning

- Labeling can be tedious, often done by humans
- Some techniques don't require labeled data
- Unsupervised Learning
 - Clustering: find groups observation that are similar
 - Does not require labeled observations



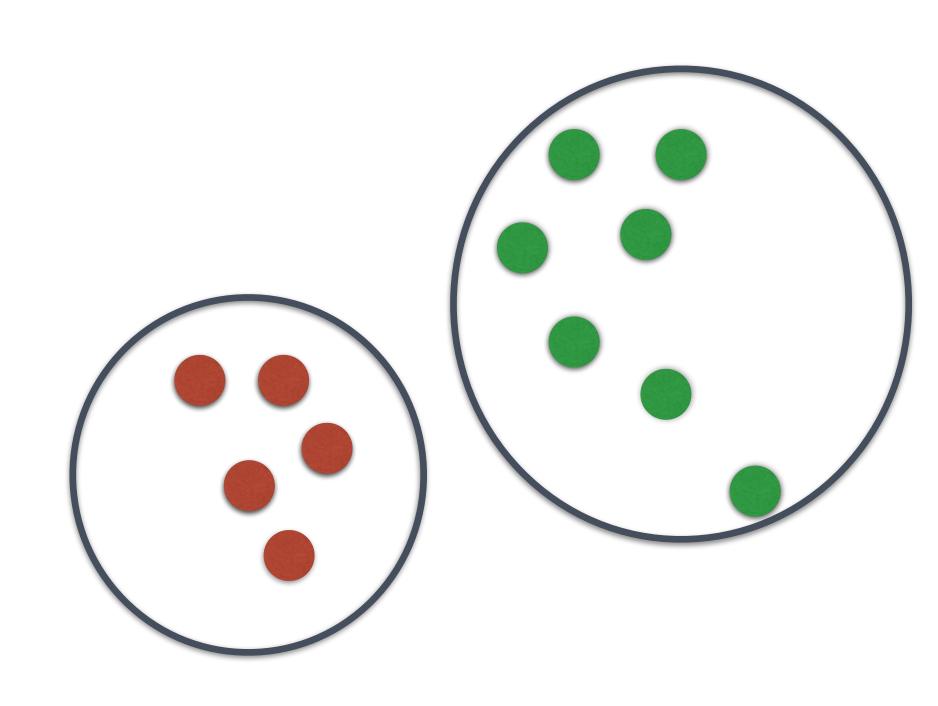
Performance of the model

- Supervised Learning
 - Compare real labels with predicted labels
 - Predictions should be similar to real labels
- Unsupervised Learning
 - No real labels to compare
 - Techniques will be explained in this course



Semi-Supervised Learning

- A lot of unlabeled observations
- A few labeled
- Group similar observations using clustering
- Use clustering information and classes of labeled observations to assign a class to unlabelled observations
- More labeled observations for supervised learning







Let's practice!