

Machine Learning 101

Data as Points in Space

Syllabus

- What is Machine Learning?
- Data as Points in Space
- Basic Relationships (magnitude and angle)
- Objective Functions (what is the good?)
- Two Basic Models
- Introduction to Decision Trees
- Overview of VenueSearch

What is Machine Learning?

- Means of using the **observed** to make inferences about the **unobserved**.
- Means of enabling computers to **program themselves**.
- Used at **Foursquare** to:
 - Determine venue visits from phone signals
 - Aggregate and reconcile venue data from across the web
 - Identify tip sentiment (positive or negative)
 - Predict user age/gender from past visits
 - And more!

Key Concepts (to look out for)

- Data
 - Features
 - Labels
- Model
- Objective function
- Model fitting
 - Training data
 - Testing data
 - Overfitting

Representation



Analysis

Big idea

metadata

metadata

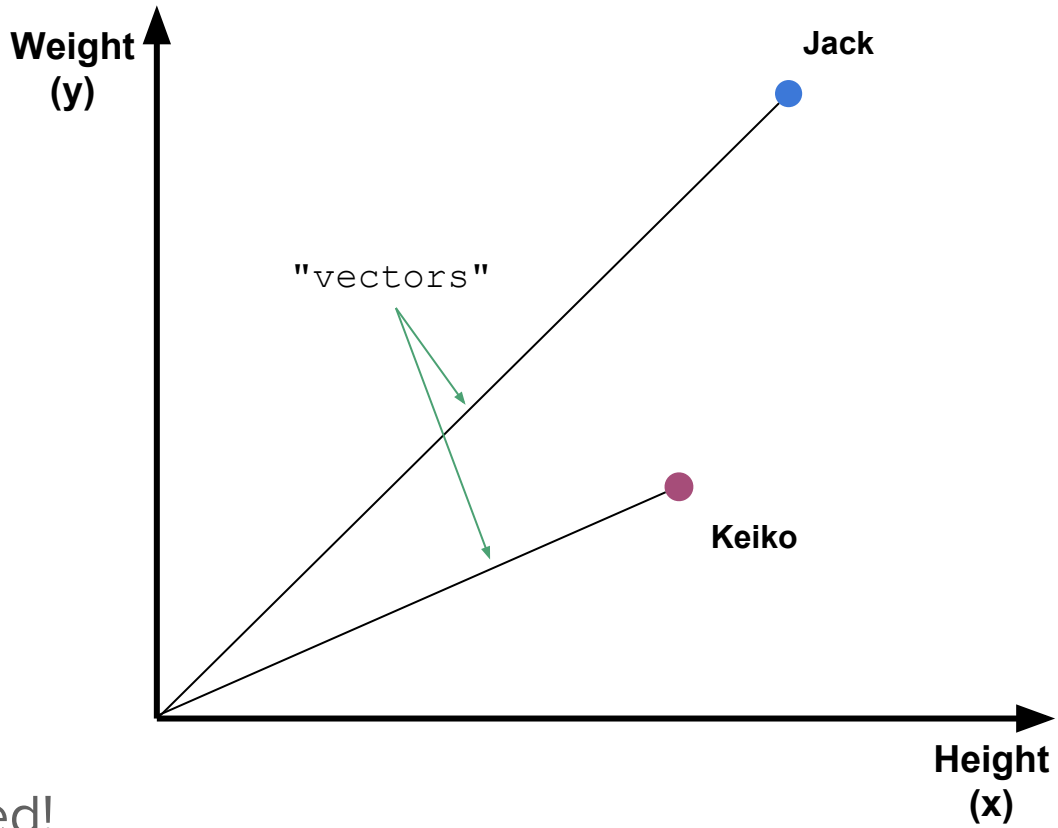
label

feature

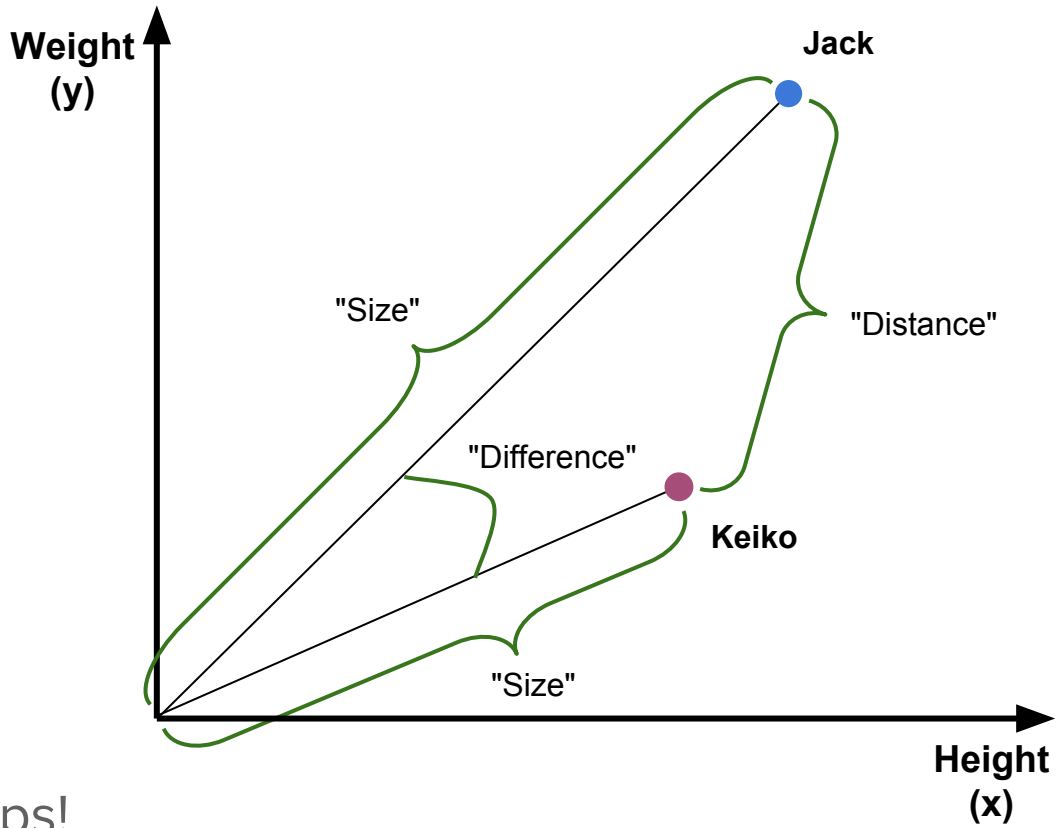
feature

ID	Name	Gender	Height (in)	Weight (lb)
1	Jack	M	70	180
2	Keiko	F	62	130
...

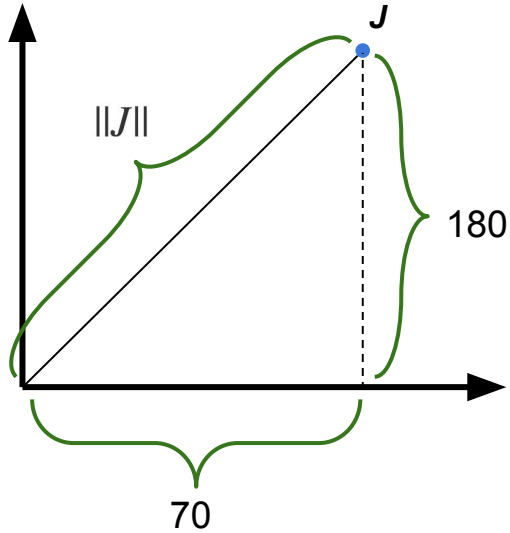
Data!



Transformed!



Relationships!



$$J = [70, 180]$$

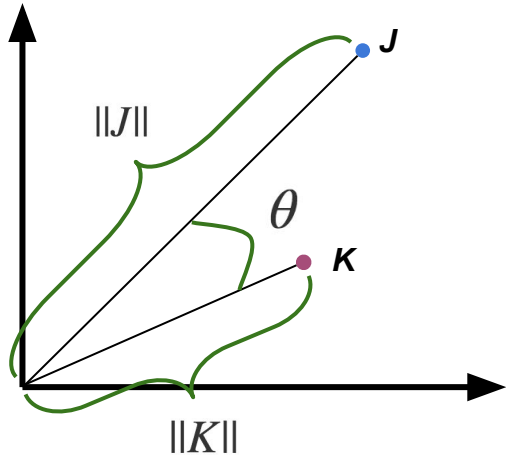
$$J \in \mathbb{R}^2$$

$$\|J\| = \sqrt{\sum_{i=1}^d J_i^2}$$

A green arrow points from the text below to the square root symbol in the equation above.

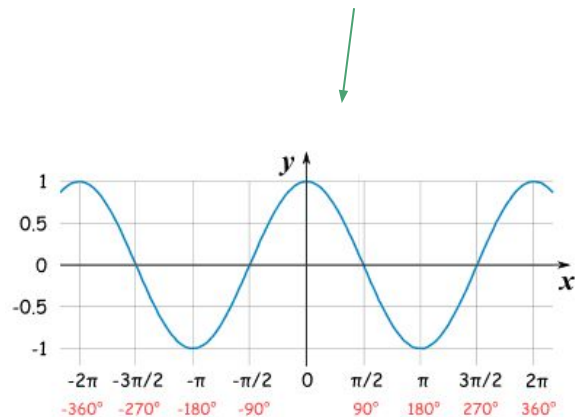
Meet the Vector

this is just the pythagorean function ;)



$$J^T K \triangleq \sum_{i=1}^d J_i \cdot K_i$$

$$= \|J\| \cdot \|K\| \cdot \cos(\theta)$$



Meet the Dot Product

So.

- Data are a set of points (vectors) in (some dimensional) space.
 - We call a vector of vectors a matrix.
- These points have (numeric) properties, mostly from trigonometry.
- Machine learning "models" are functions of these properties:


$$f(D)$$

*To know the good is
to desire the good.
- Socrates*

What is "the good" ?

$$\mathcal{L}(f, D)$$

The "Objective Function" (aka "Loss Function")

Idea:

- The objective function tells us how "good" the model is.
- These functions come in many flavors. Two common ones are:
 - Measurements of **error** (which we want to **minimize**).
 - Measurements of **likelihood** (which we want to **maximize**).
- We can use various techniques to **optimize** the model w.r.t. the objective.
- We call this process "**fitting**" a model (to data).

WIZARD PARTY

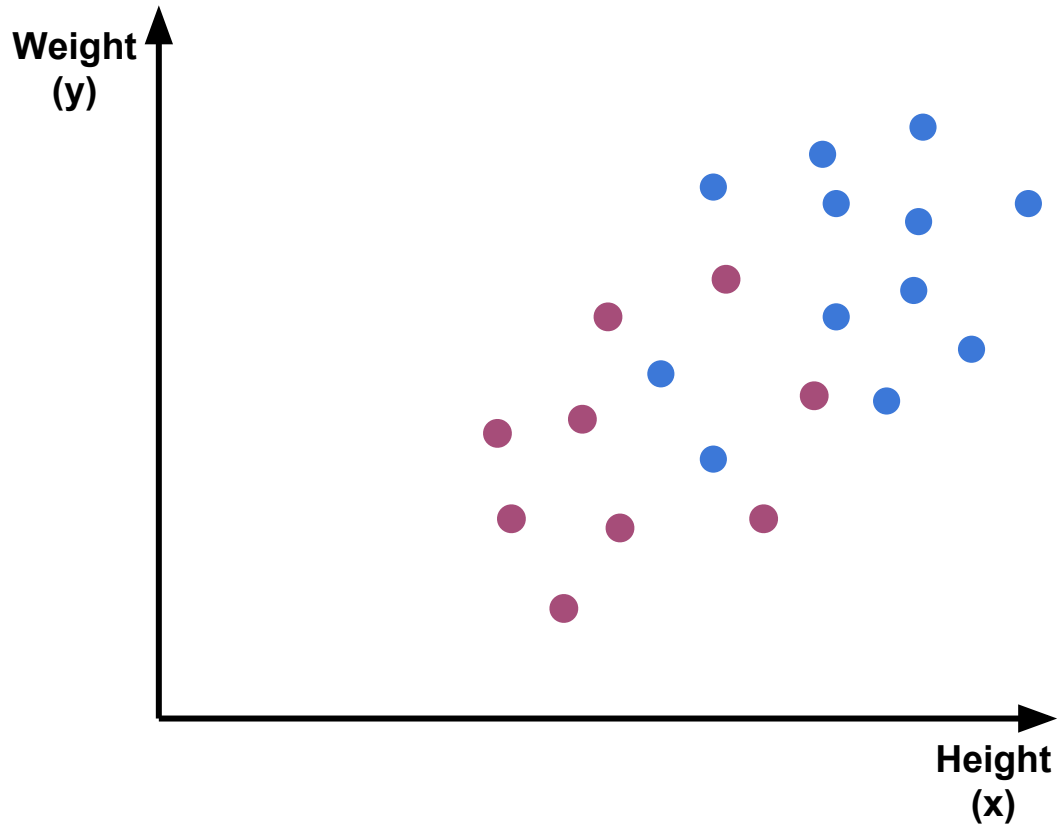


Example: Least-Squares Regression

- **Idea:** features of a data point are related in some way.
- Example: predict MPG for a car using year, weight, engine size, etc.
- Objective function:

$$\mathcal{L} = \sum_{i=1}^n \overbrace{(y_i - f(x_i))^2}^{\text{per-point "error"}}$$

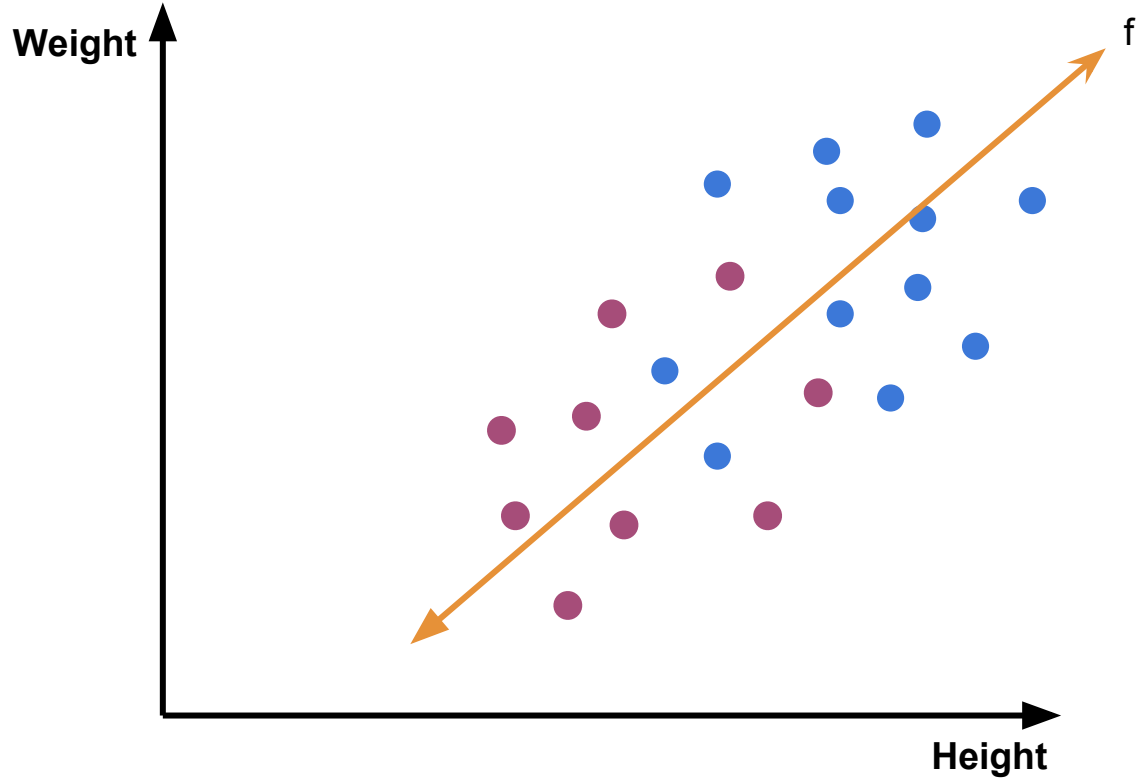
Moar Data!



WIZARD PARTY



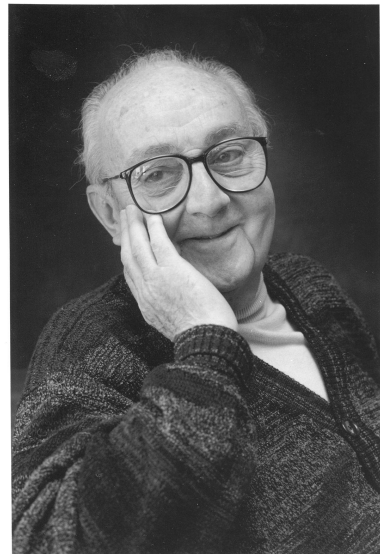
Optimize!



Bam!

Note:

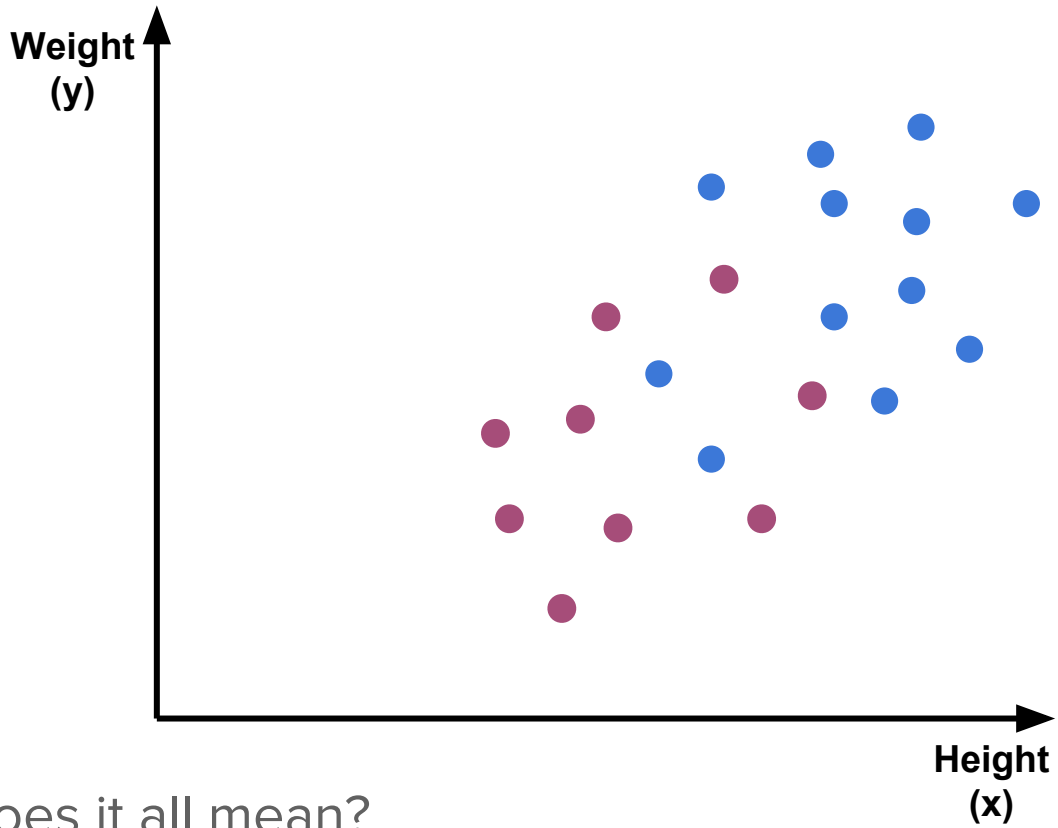
- Obviously, the real relationship between height and weight is very complex.
- A model is a simplification of the real world.
- ***"All models are wrong, but some are useful"***
- George Box



Example: Logistic Regression

- **Idea:** the features of a data point can tell us about the **class** of thing it is.
- Used at **Foursquare** for sentiment analysis (positive or negative sentiment?).
- Objective function:

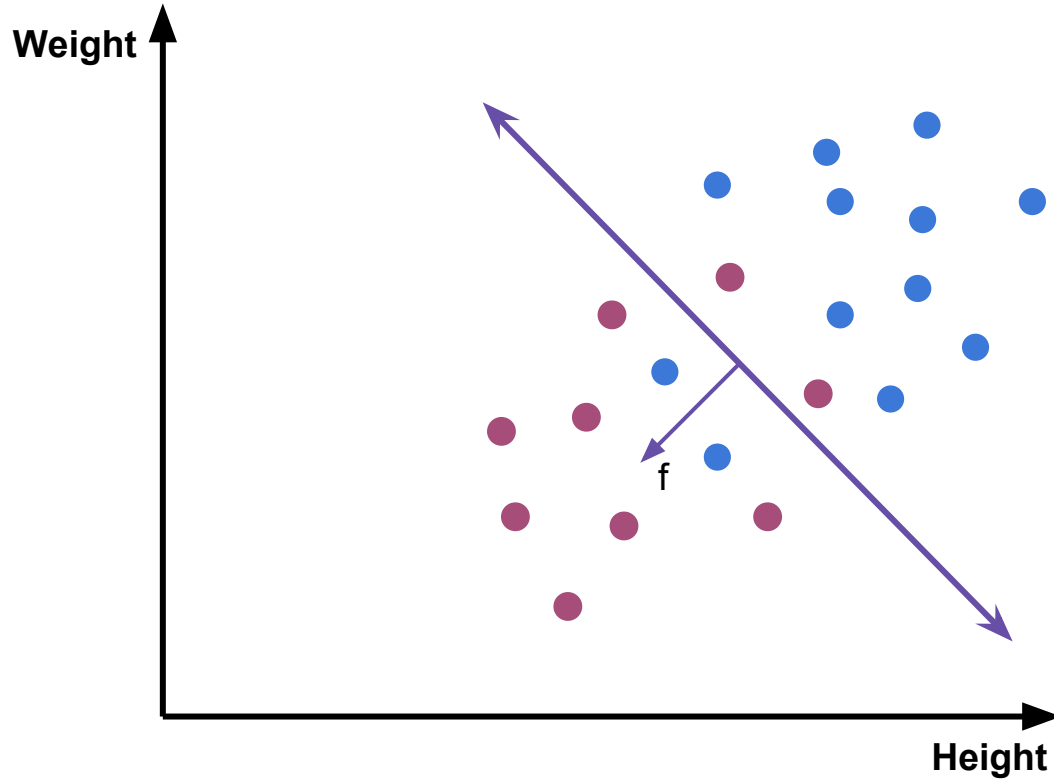
$$\mathcal{L} = \sum_{i=1}^n \log(\underbrace{f(x_i, y_i) \cdot c_i}_{\text{probability of right class}})$$



WIZARD PARTY



Optimize!



Wham!

Summary so far:

- Data can be thought of as points (vectors) in (some dimensional) space.
- A machine learning system consists of the following building blocks:
 - A model (a simplified view of some real-world relationship)
 - An objective function (a way to measure how good a model is, relative to data)
 - An algorithm for updating the model with regards to the objective.
- Neural networks, etc are just fancier versions of this.
- No magic anywhere! Except the dot product.

WIZARD PARTY

