



AI Bridge

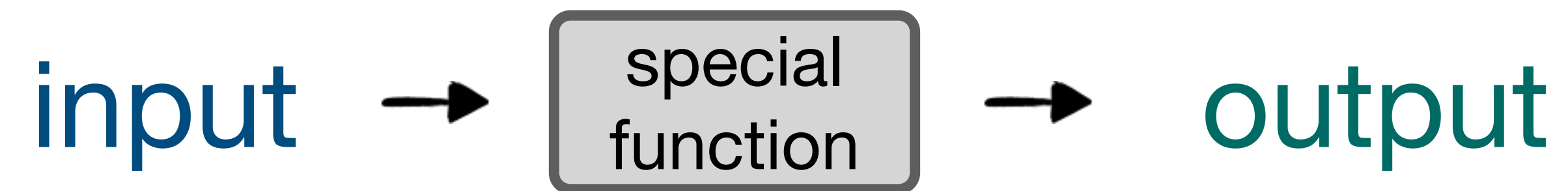
Lecture 4

Our first question:

what is Artificial Intelligence?

Generally...

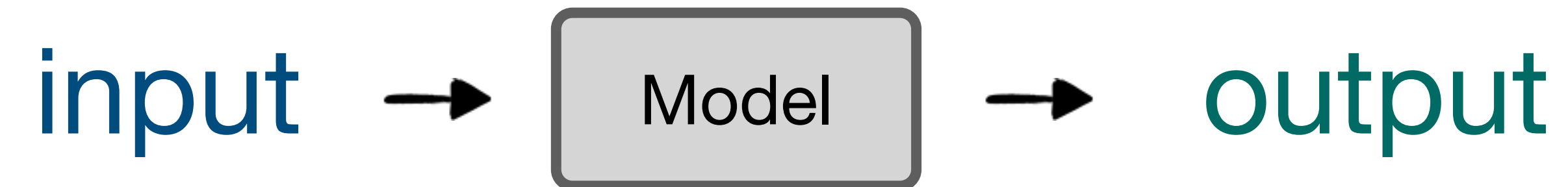
- An AI is a fancy function that maps **inputs** to **outputs**.



More generally, **given a thing**, AI can tell you a thing.

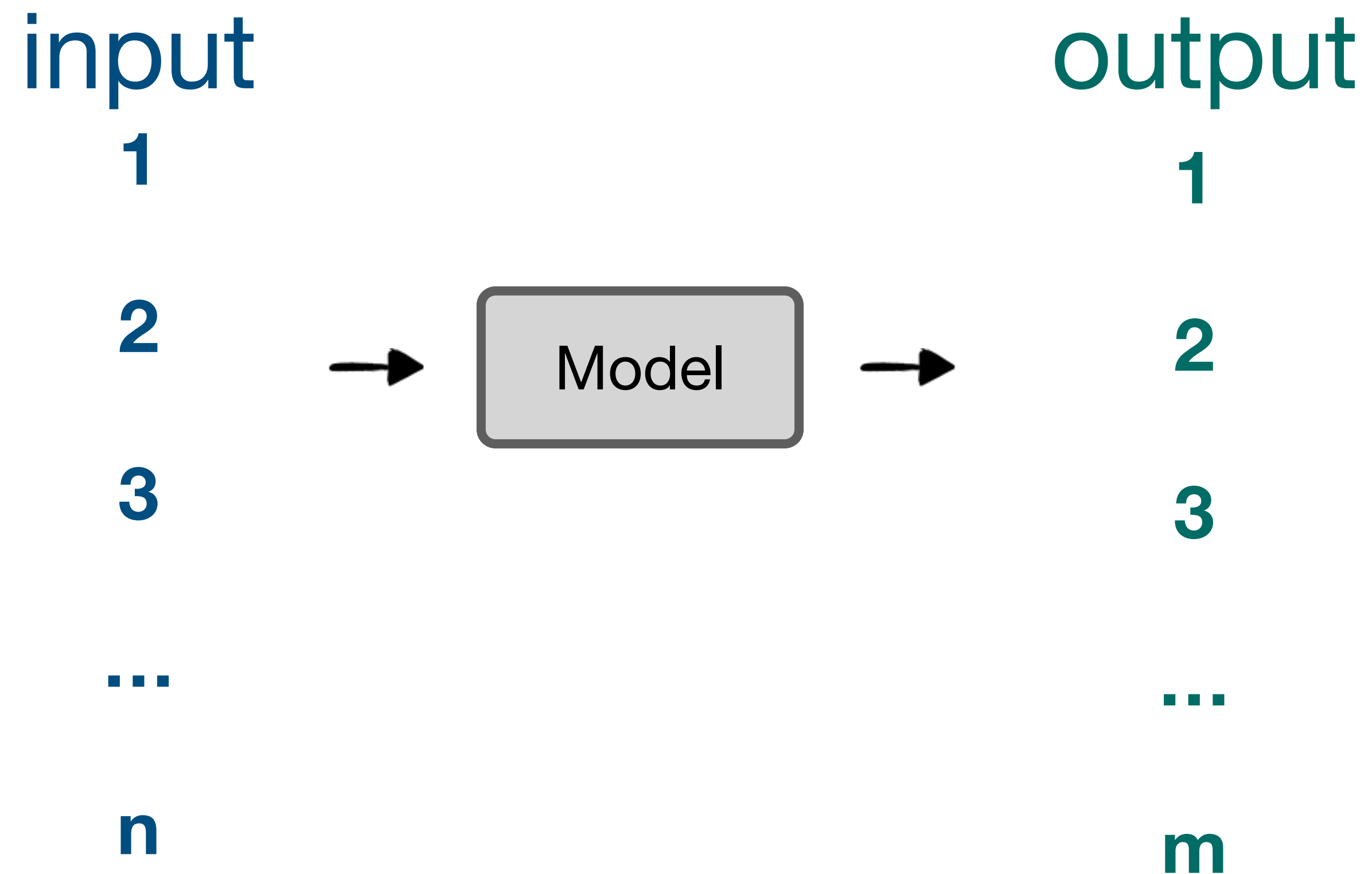
AI Models

- We call different types of AI functions **models**.

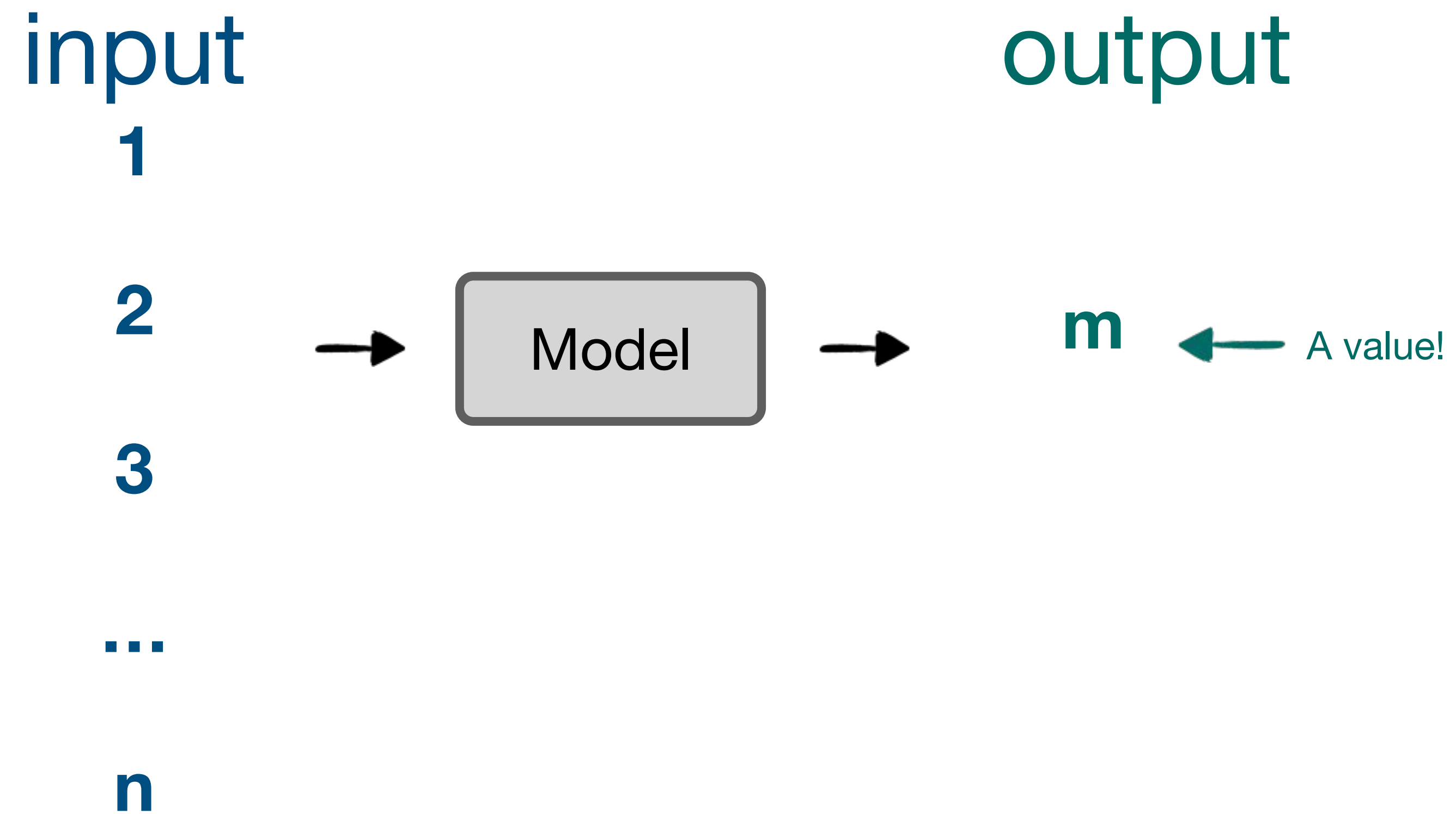


AI Models

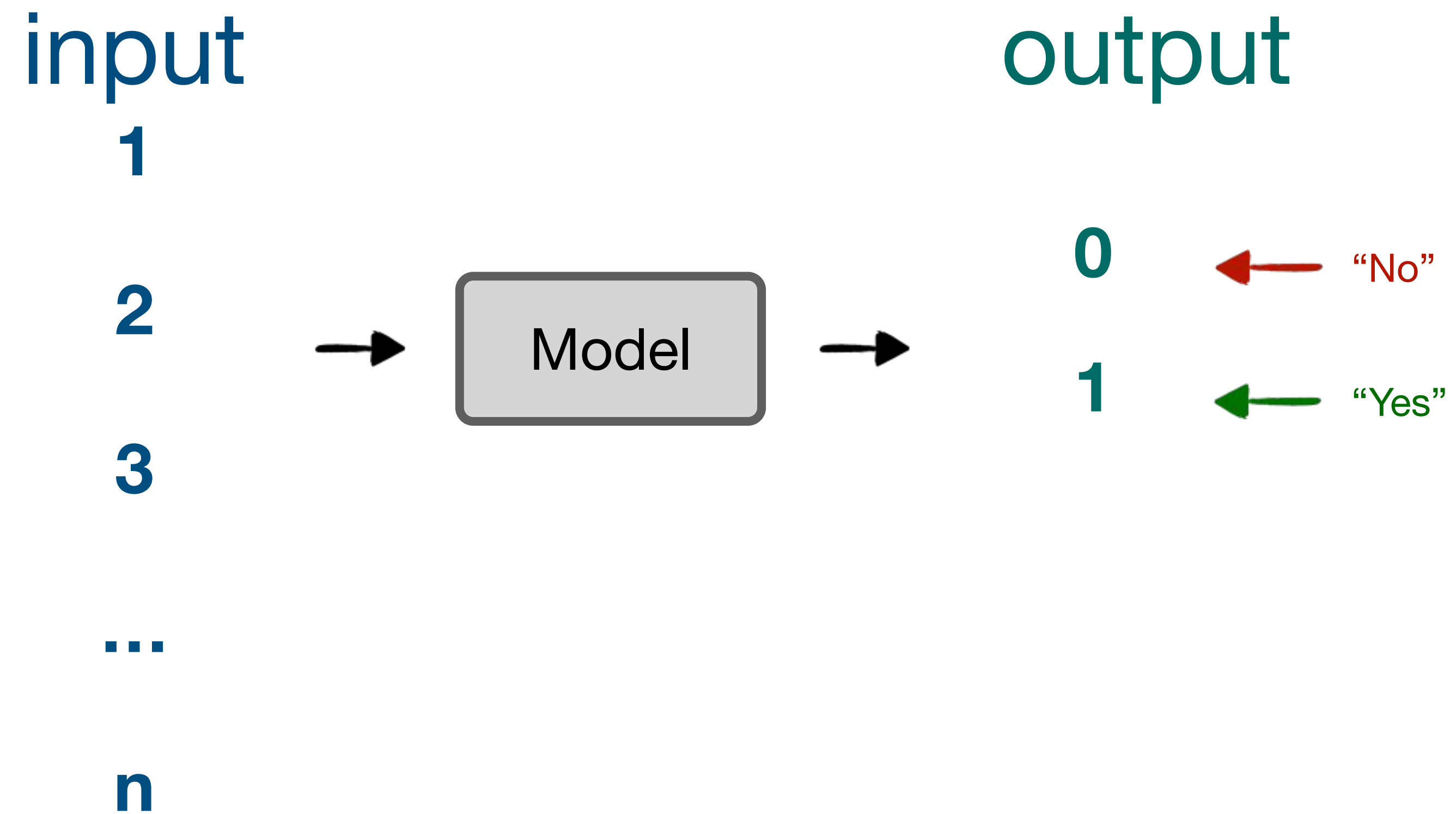
- What are the inputs and outputs?
- R_n to R_m functions



AI Model Examples



AI Model Examples



AI Model Examples

input

output

cat

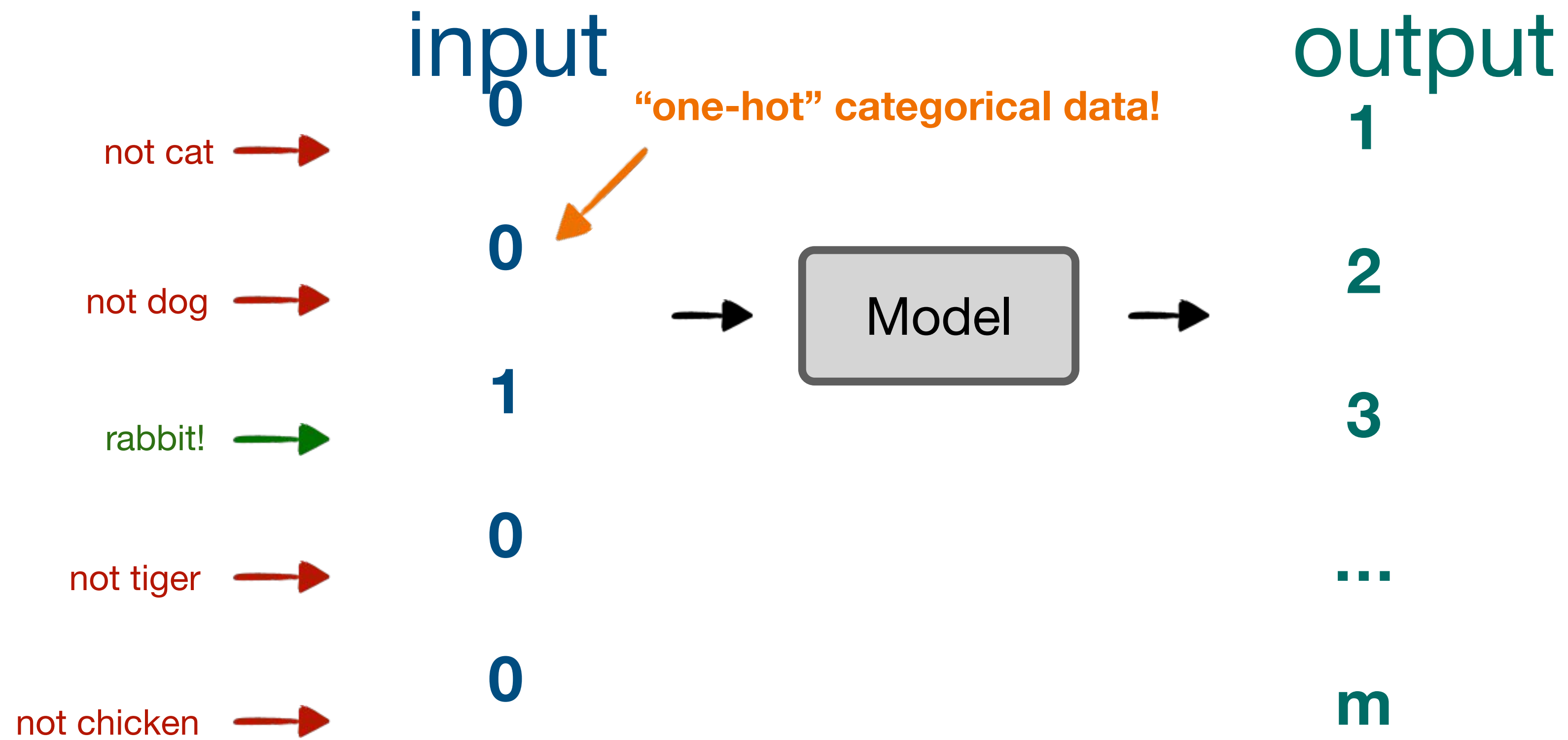
dog

rabbit!

tiger

chicken

AI Model Examples



input

output



- by changing the shapes of input and output, models can represent a lot of different problems

- Labeled data
- Direct feedback
- Predict outcome/future



- No labels
- No feedback
- "Find hidden structure"

- Decision process
- Reward system
- Learn series of actions

SUPERVISED LEARNING

Supervised Learning

- “**Supervised learning** (SL) is the machine learning task of learning a function that maps an input to an output based on example input-output pairs. It infers a function from labeled training data consisting of a set of training examples.” – Wikipedia
- Input-output pairs: Features and labels
- Training/learning and inference
- Most widely used ML techniques in real-world applications.

Supervised Learning



What types are there?

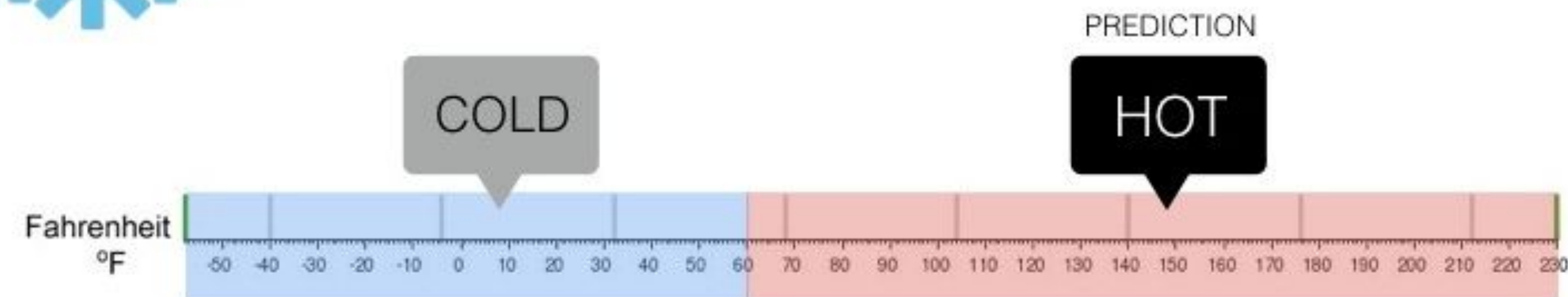
Classification

- Predicting a label/class/category
 - Ex: spam or not, cancer or not, cat or dog, red wine vs. white wine

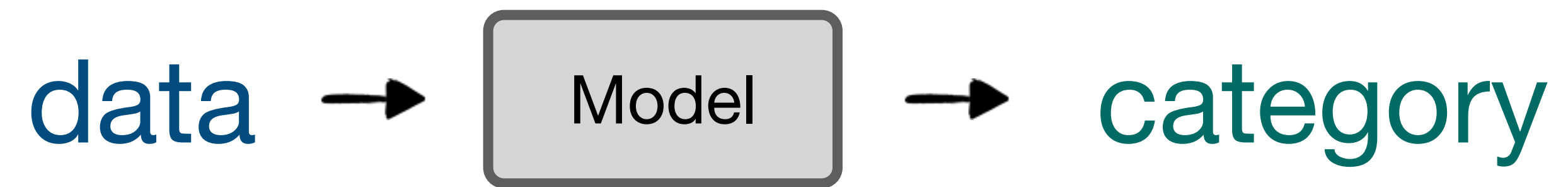


Classification

Will it be Cold or Hot tomorrow?



What does classification do?



Regression

- Predicting a (continuous) quantity
 - Ex: Survival rate, wine quality, yield prediction

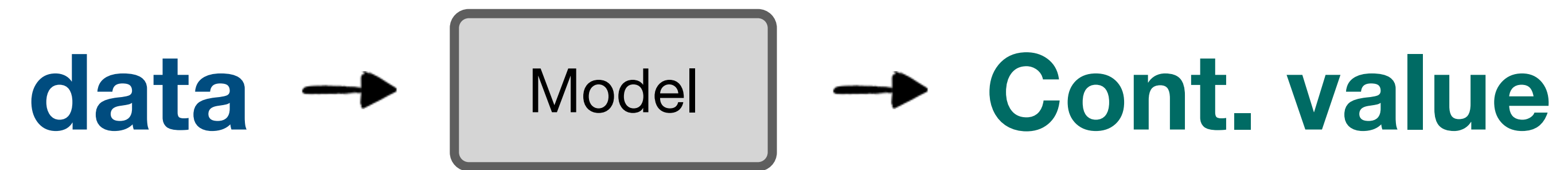


Regression

What is the temperature going to be tomorrow?



What does regression do?



Examples

- You're running a company, and you want to develop learning algorithms to address each of two problems.
 - Problem 1: You have a large inventory of identical items. You want to predict how many of these items will sell over the next 3 months.
 - Problem 2: You'd like software to examine individual customer accounts, and for each account decide if it has been hacked/compromised.
- Are they **classification or regression?**

Supervised Algorithms Practice

- **You're running a company, and you want to develop learning algorithms to address each of two problems.**
 - Problem 1: You have a large inventory of identical items. You want to predict how many of these items will sell over the next 3 months.
- **Can we formulate it as a classification problem?**

Supervised Algorithms Practice

- **You're running a company, and you want to develop learning algorithms to address each of two problems.**
 - Problem 1: You have a large inventory of identical items. You want to predict how many of these items will sell over the next 3 months.
- **Can we formulate it as a classification problem?**

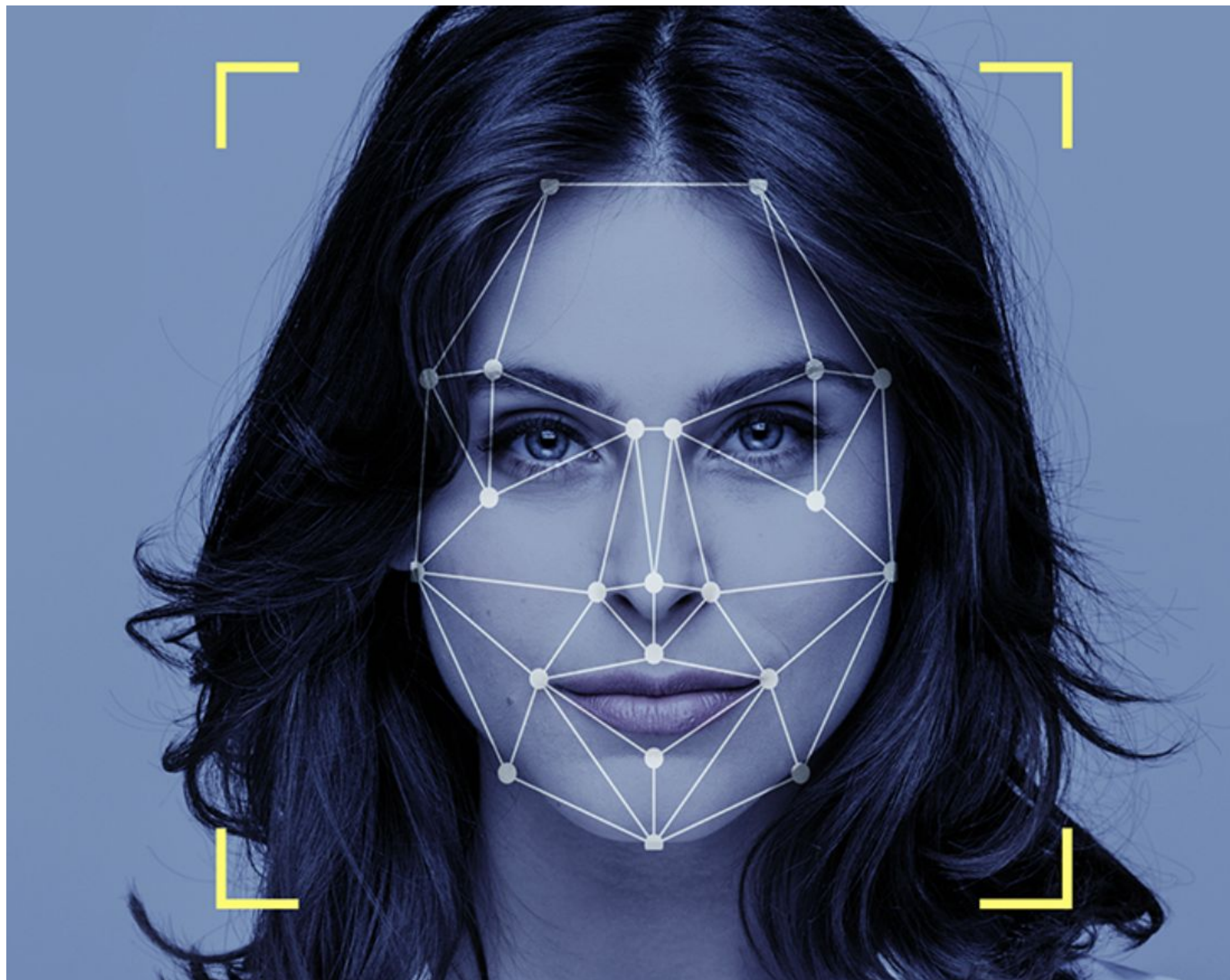
We could. Sometimes, we can reformulate.

Let's start with regression instead!

Speech Recognition



Face Recognition



Sentiment Analysis

Sentiment Analysis

The diagram illustrates Sentiment Analysis with three examples, each represented by a card with an emoji, a text snippet, and a sentiment label.

- Positive:** A smiling face emoji (😊) is shown above the text "My experience so far has been fantastic!". The word "fantastic!" is highlighted in a light green box, and a green button labeled "POSITIVE" is at the bottom.
- Neutral:** A neutral face emoji (😐) is shown above the text "The product is ok I guess". The words "ok I guess" are highlighted in a light yellow box, and a yellow button labeled "NEUTRAL" is at the bottom.
- Negative:** An angry face emoji (😡) is shown above the text "Your support team is useless". The words "useless" are highlighted in a light red box, and a red button labeled "NEGATIVE" is at the bottom.

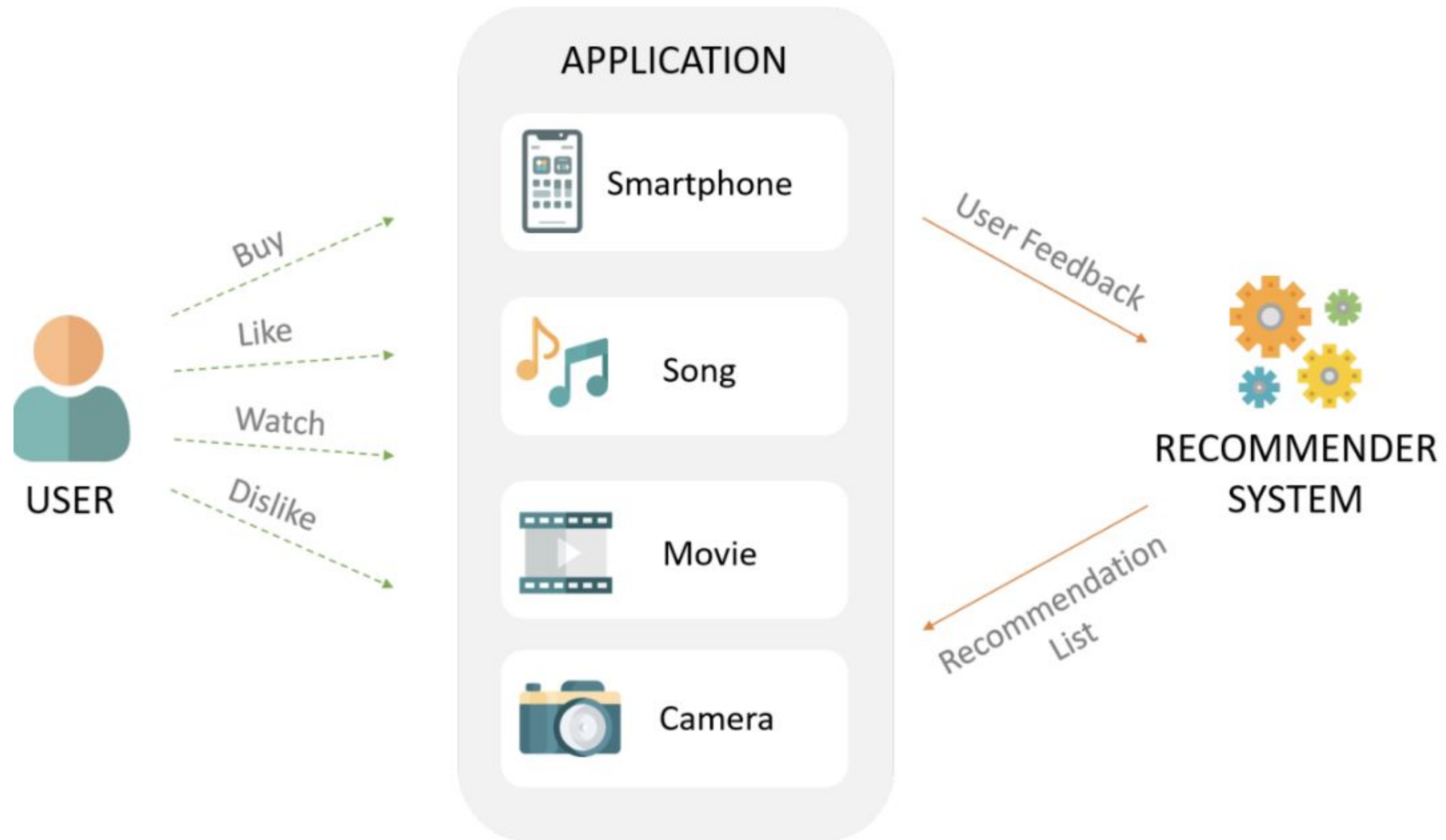
Spam Filter



Fraud Detection



Recommendation



Stock Price Prediction



Food Quality/Safety Prediction



Our dataset: wine quality

What does our dataset look like?



The header of the UCI Machine Learning Repository website. It features the UCI logo with a sloth, the text 'Machine Learning Repository' and 'Center for Machine Learning and Intelligent Systems'. Navigation links include 'About', 'Citation Policy', 'Donate a Data Set', and 'Contact'. A search bar is present with a 'Search' button and radio buttons for 'Repository' (selected) and 'Web'. A 'View ALL Data Sets' link is also visible.

Wine Quality Data Set

Download: [Data Folder](#), [Data Set Description](#)

Abstract: Two datasets are included, related to red and white vinho verde wine samples, from the north of Portugal. The goal is to model wine quality based on physicochemical tests (see [Cortez et al., 2009], [[Web Link](#)]).



Data Set Characteristics:	Multivariate	Number of Instances:	4898	Area:	Business
Attribute Characteristics:	Real	Number of Attributes:	12	Date Donated	2009-10-07
Associated Tasks:	Classification, Regression	Missing Values?	N/A	Number of Web Hits:	1891084

Source:

Paulo Cortez, University of Minho, Guimarães, Portugal, <http://www3.dsi.uminho.pt/pcortez>
A. Cerdeira, F. Almeida, T. Matos and J. Reis, Viticulture Commission of the Vinho Verde Region(CVRVV), Porto, Portugal
@2009

Wow! 12 attributes!
(and quality, which can be counted as a 13th)

Structuring our dataset

- Fixed acidity
- Volatile acidity ← a feature of input data
- Citric acid
- Residual sugar
- Chlorides
- Free sulfur dioxide
- Total sulfur dioxide
- Density
- pH
- Sulphates
- Alcohol
- White/Red

■ a **sample** is a collection of **features**

So if we want to predict wine quality...

- Fixed acidity
- Volatile acidity
- Citric acid
- Residual sugar
- Chlorides
- Free sulfur dioxide
- Total sulfur dioxide
- Density
- pH
- Sulphates
- Alcohol
- White/Red



Quality (0-10)

...we'll need a model!

- Fixed acidity
- Volatile acidity
- Citric acid
- Residual sugar
- Chlorides
- Free sulfur dioxide
- Total sulfur dioxide
- Density
- pH
- Sulphates
- Alcohol
- White/Red



LINEAR REGRESSION

Linear Regression

- **What does linear regression represent?**

The features

vs.

The label

Fixed acidity

Volatile acidity

Citric acid

Residual sugar

Chlorides

Free sulfur dioxide

Total sulfur dioxide

Density

pH

Sulphates

Alcohol

White/Red

Quality (0-10)

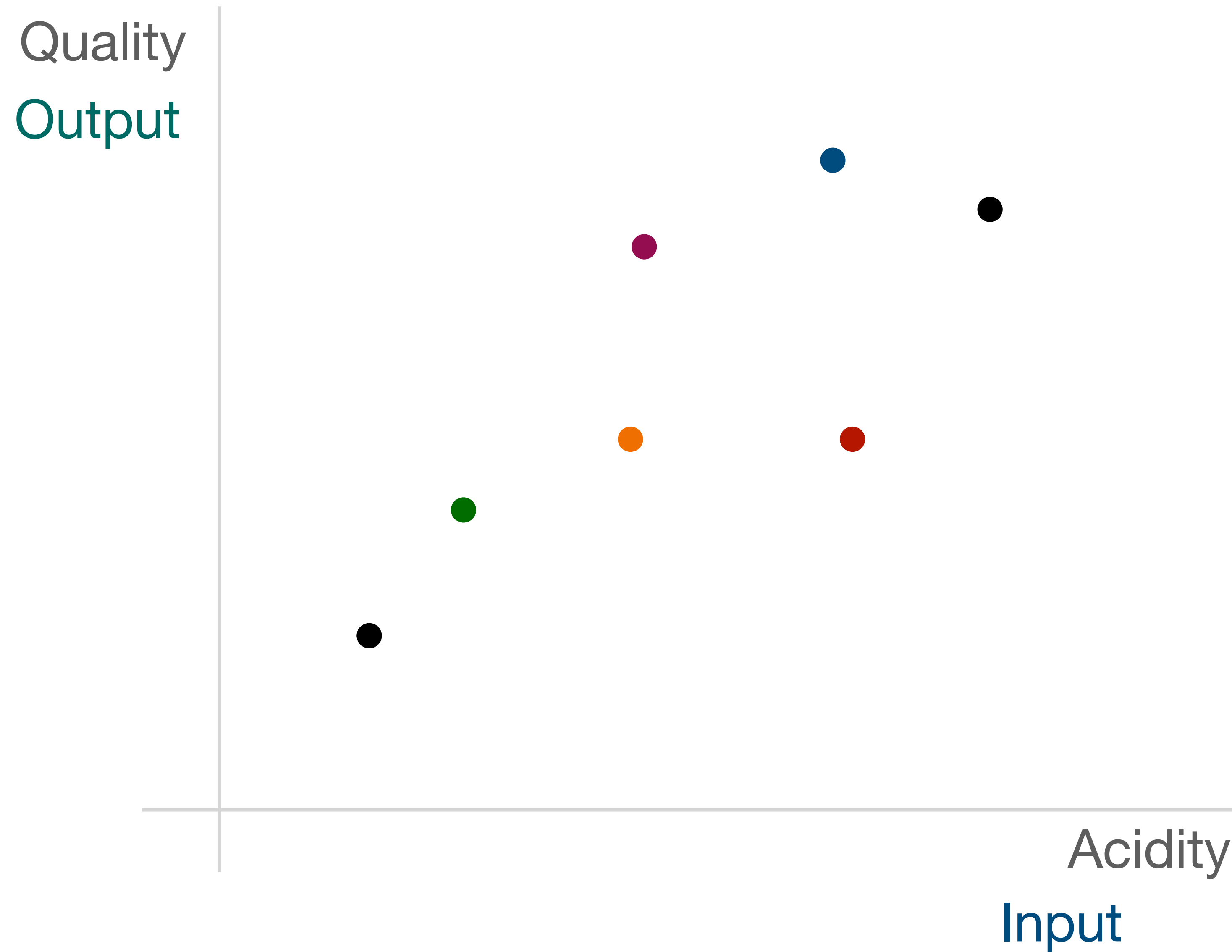
Let's start with one variable.

Linear Regression

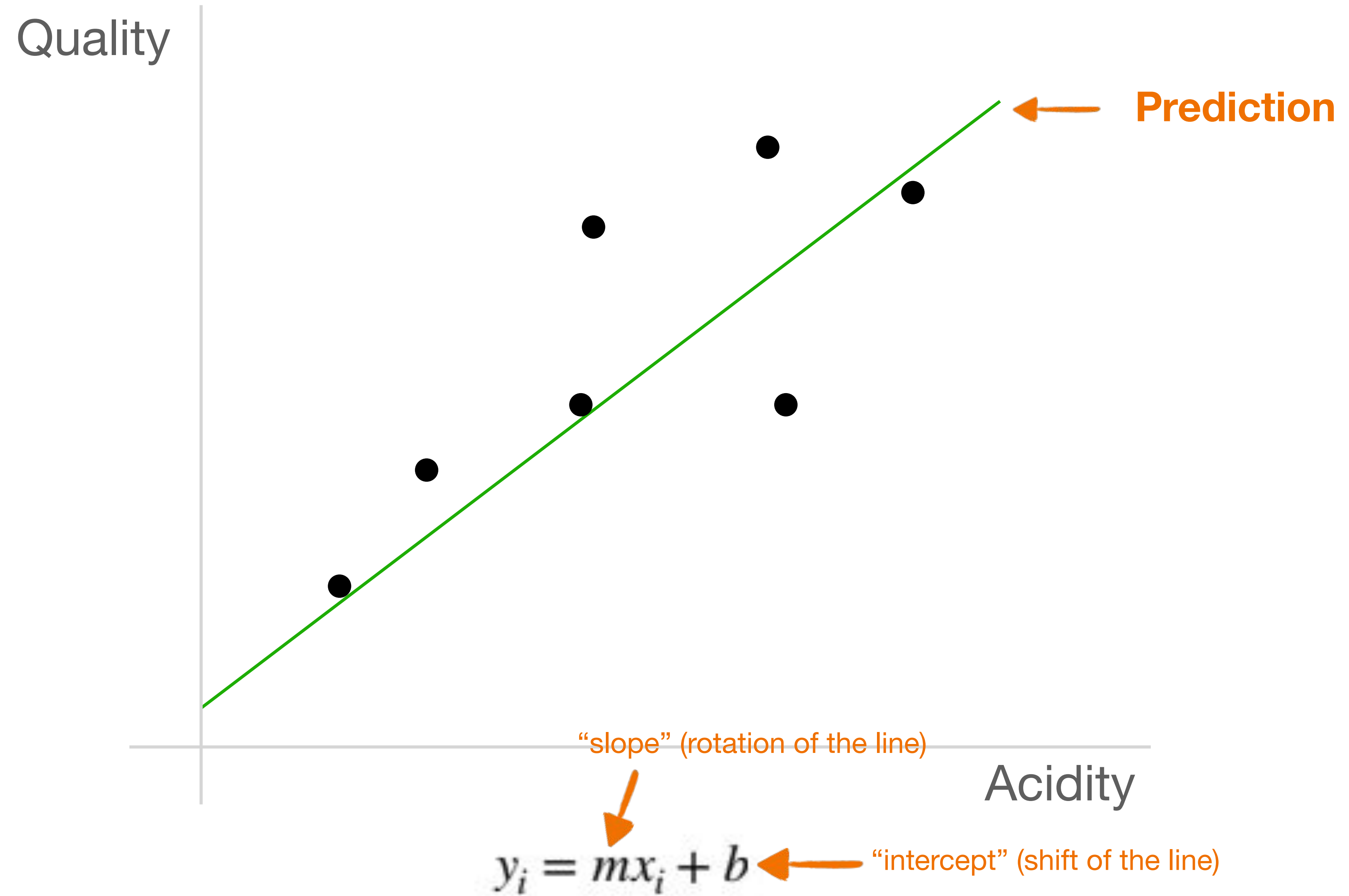
- We'll just consider **one feature** of our **sample**.



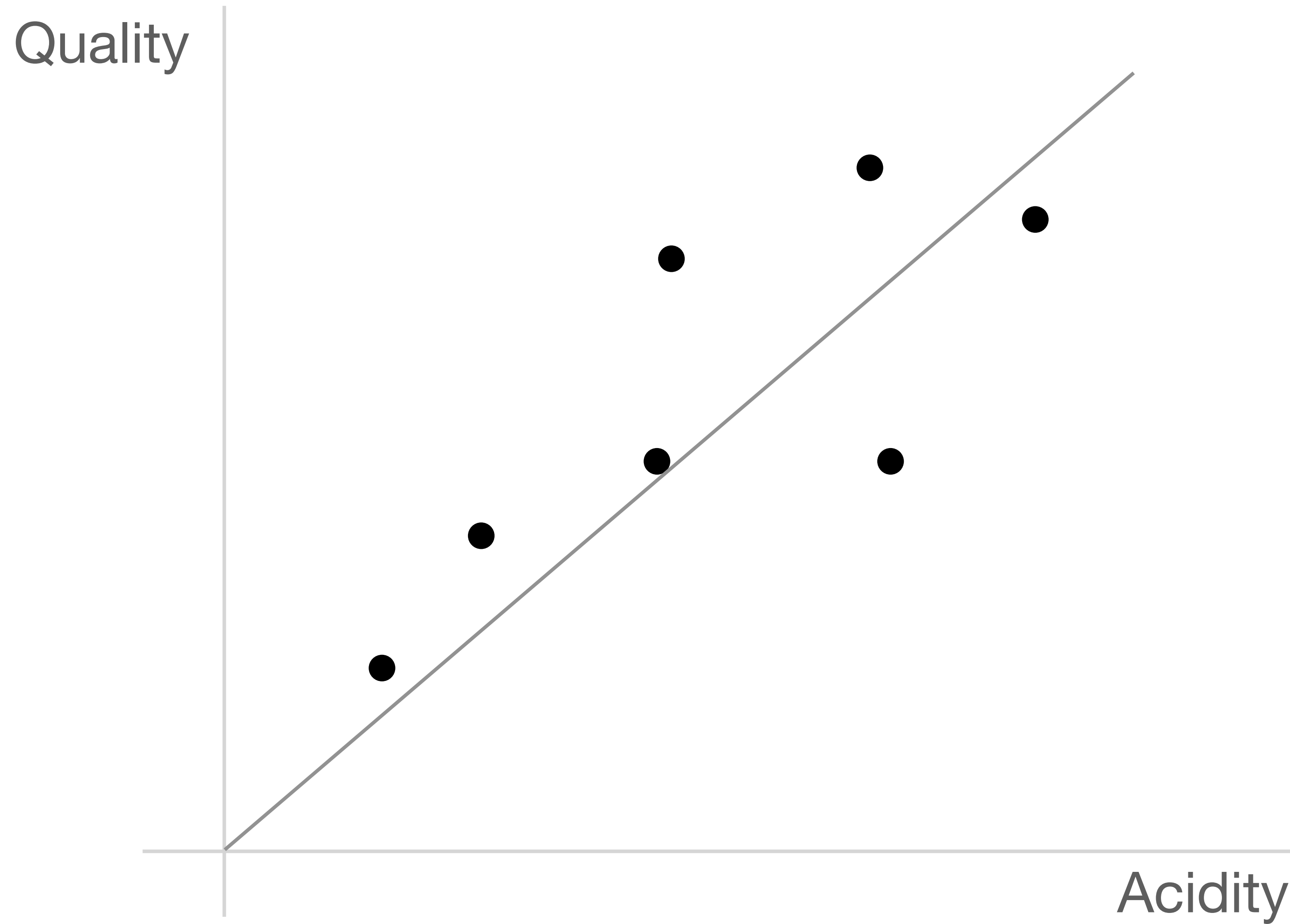
Linear Regression



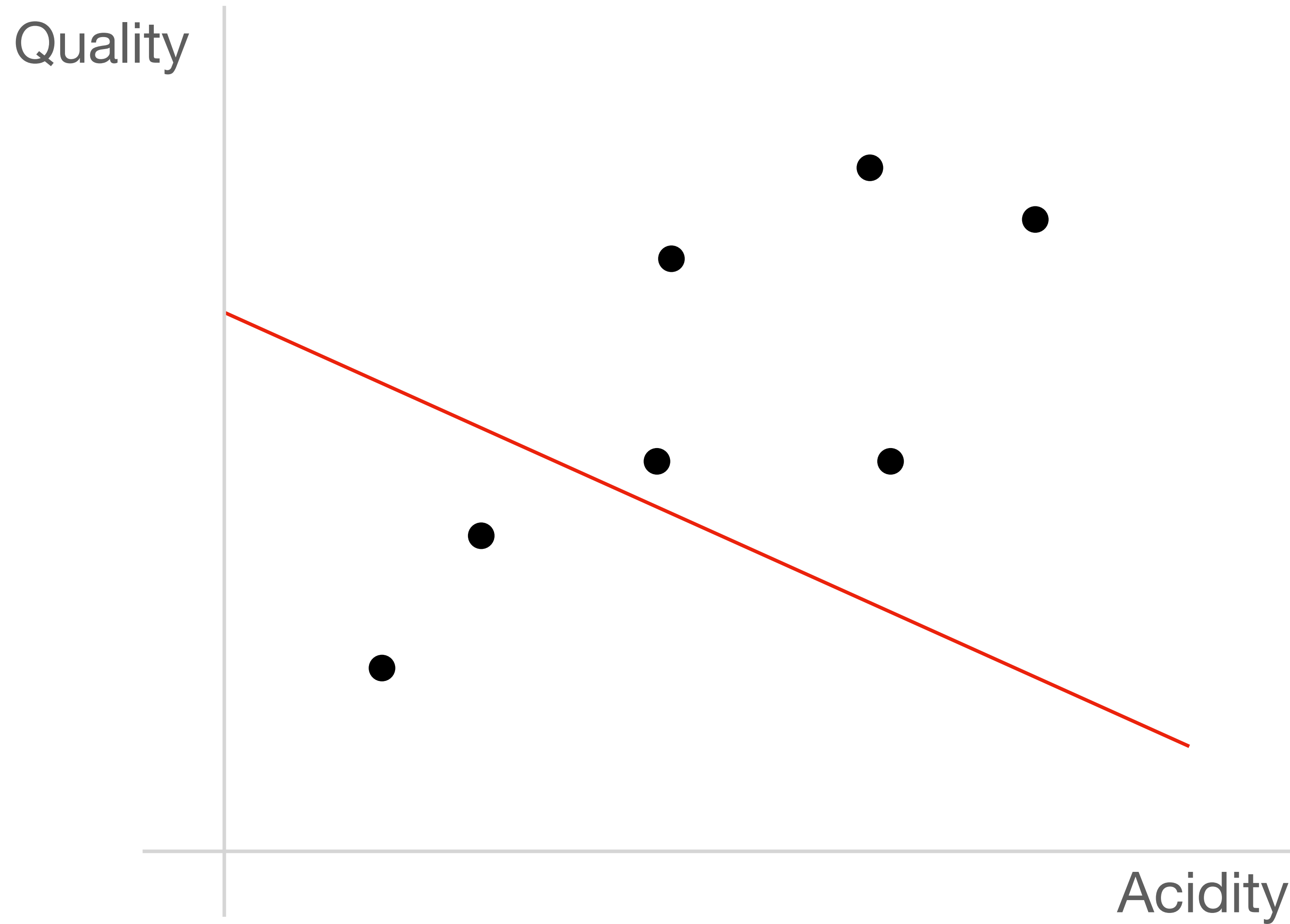
Linear Regression



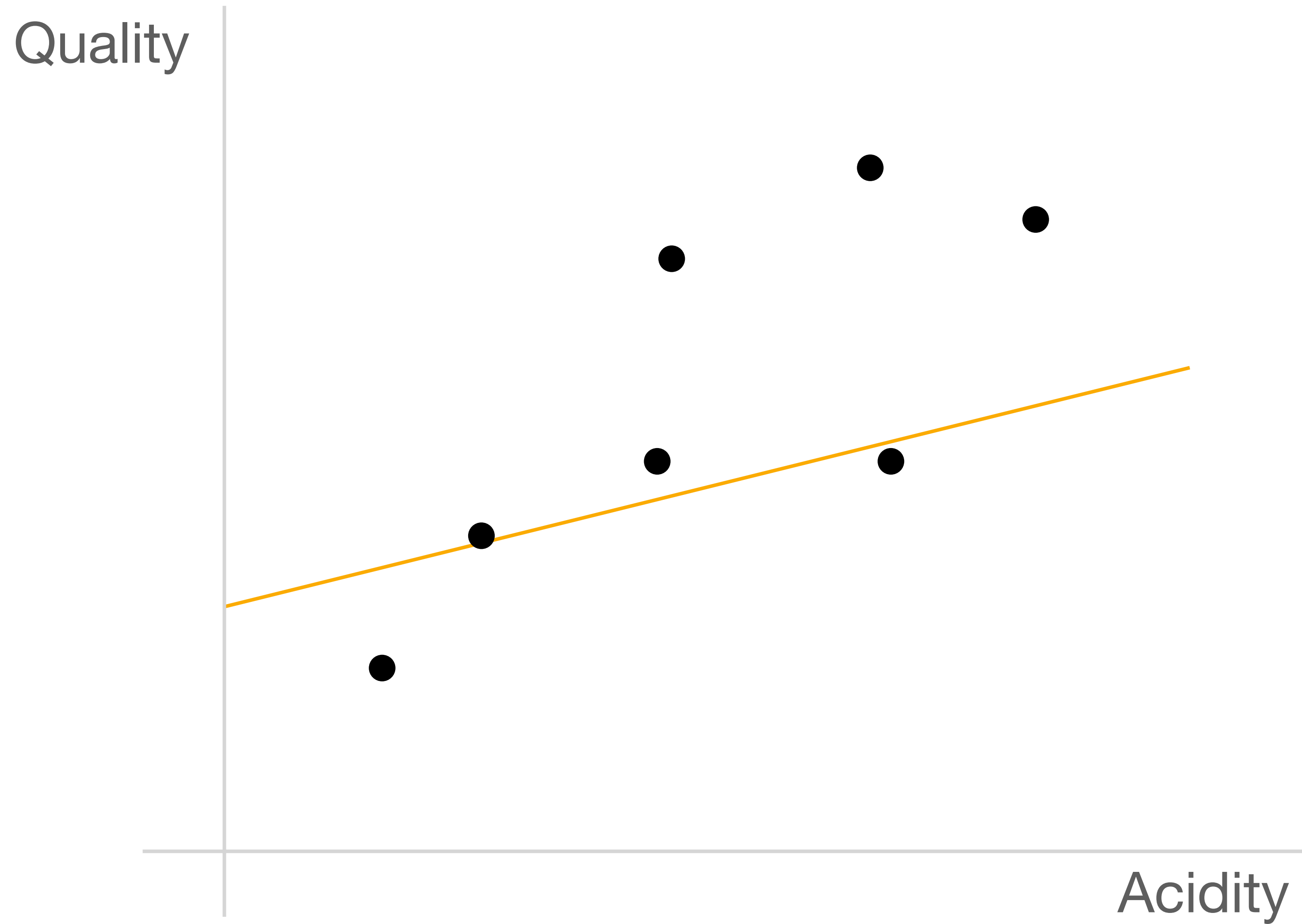
Linear Regression



Linear Regression



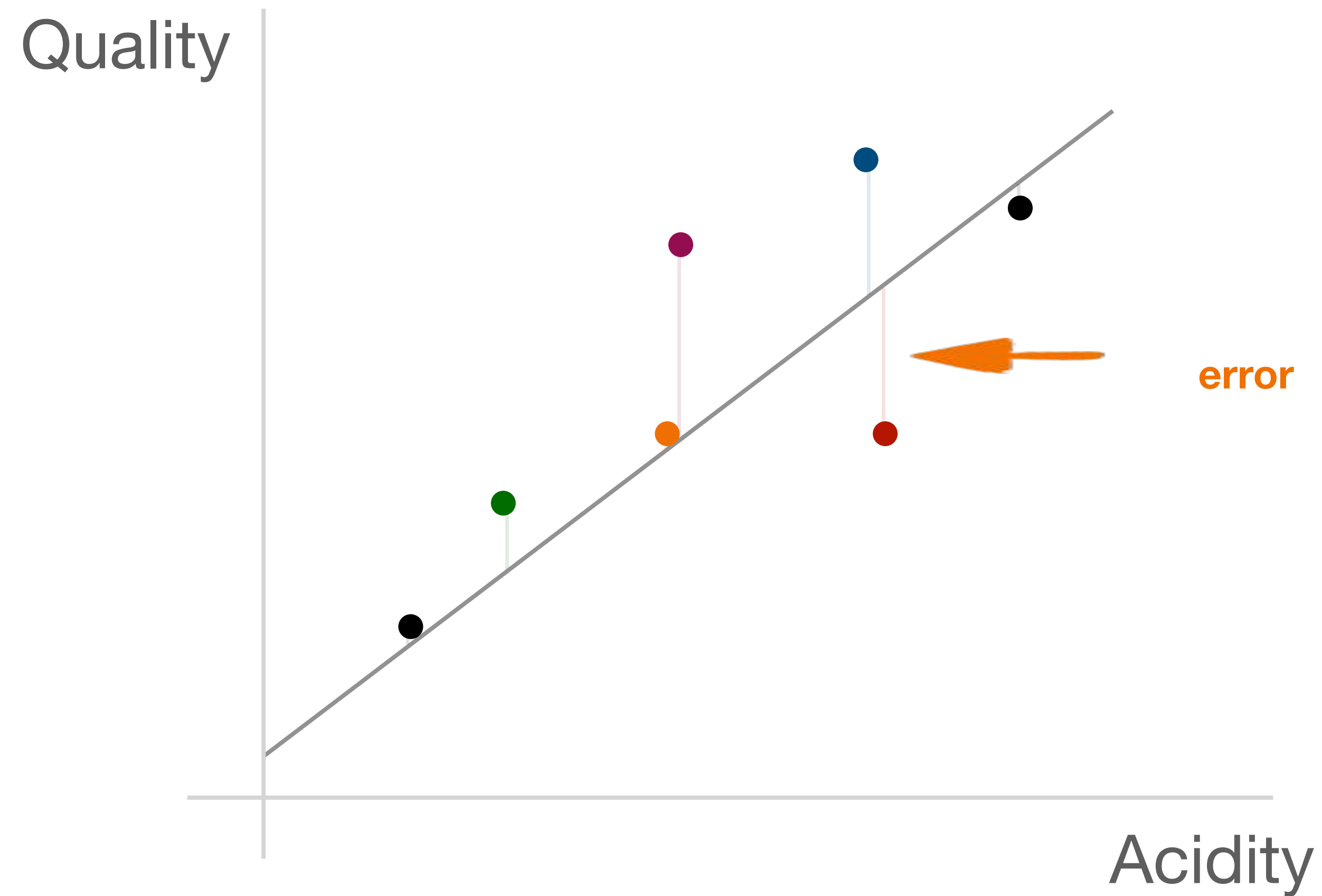
Linear Regression



How do we evaluate these
models objectively?

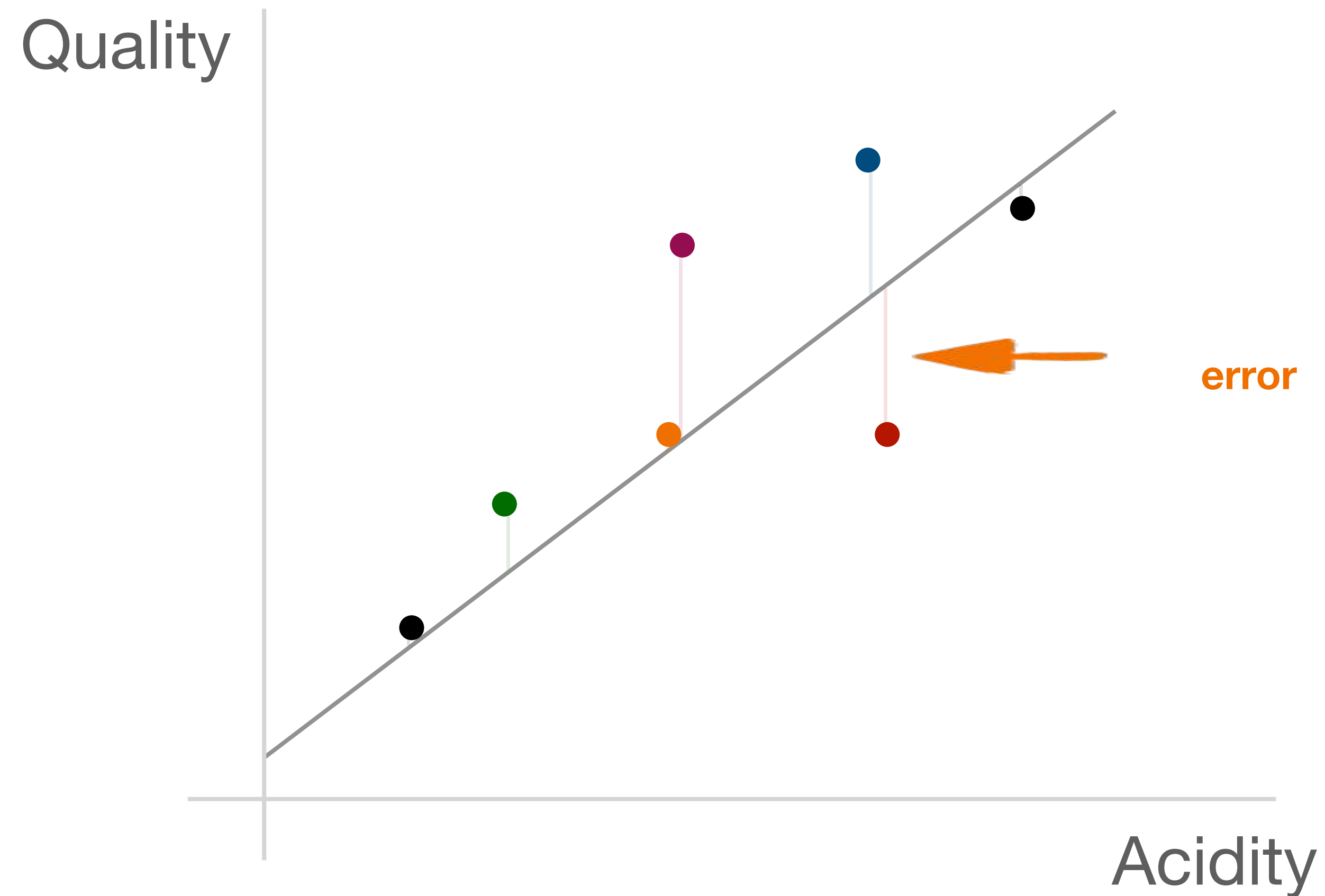
Calculating Error

- **error** is a measure of the “incorrectness” of a line



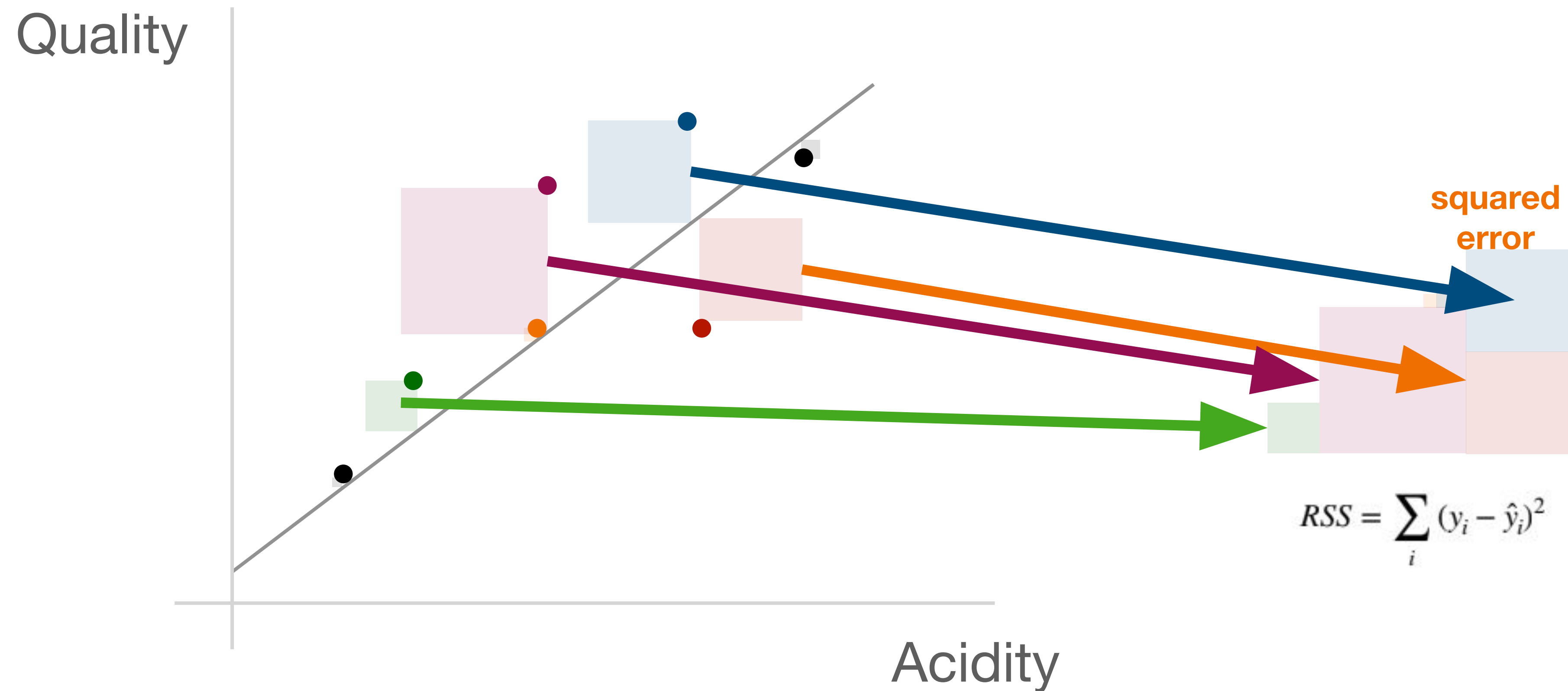
Calculating Error

- **simple error:** difference between the predicted value and the actual value



Calculating Error

- **sum-of-squares error**: sum of the squared difference between predicted and actual values



Calculating Error

- How do we minimize error? Cost function

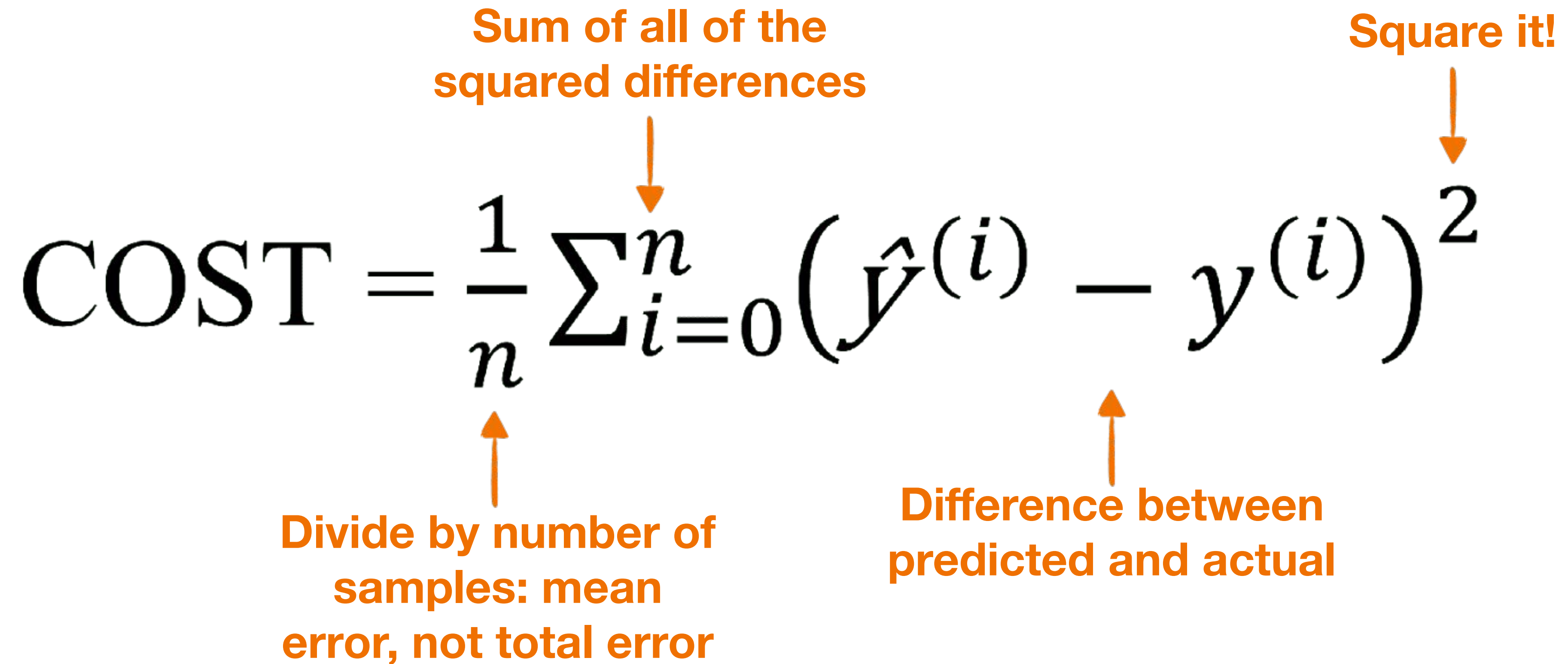
$$\text{COST} = \frac{1}{n} \sum_{i=0}^n (\hat{y}^{(i)} - y^{(i)})^2$$

Sum of all of the squared differences

Square it!

Divide by number of samples: mean error, not total error

Difference between predicted and actual

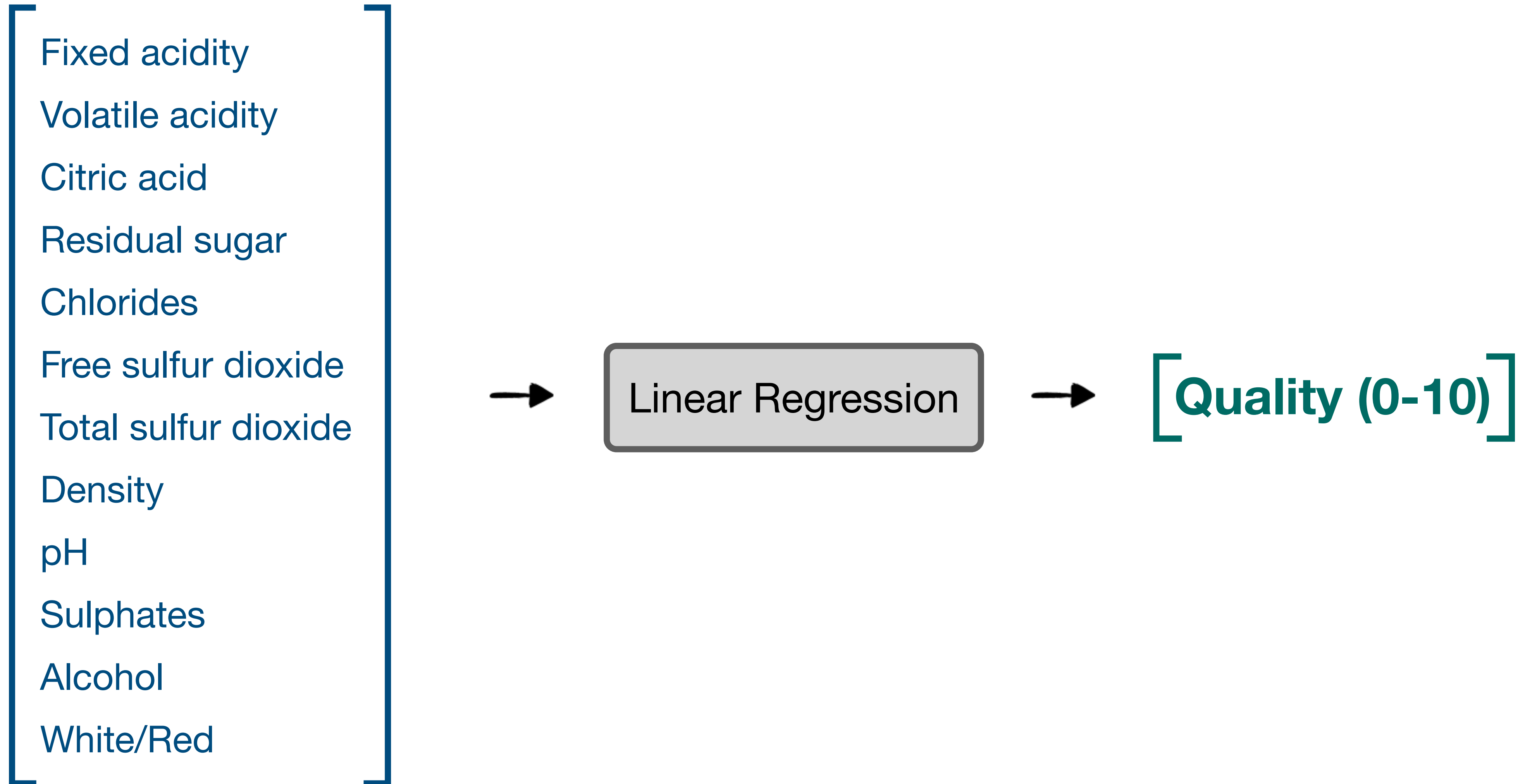


Higher-dimensional linear regression

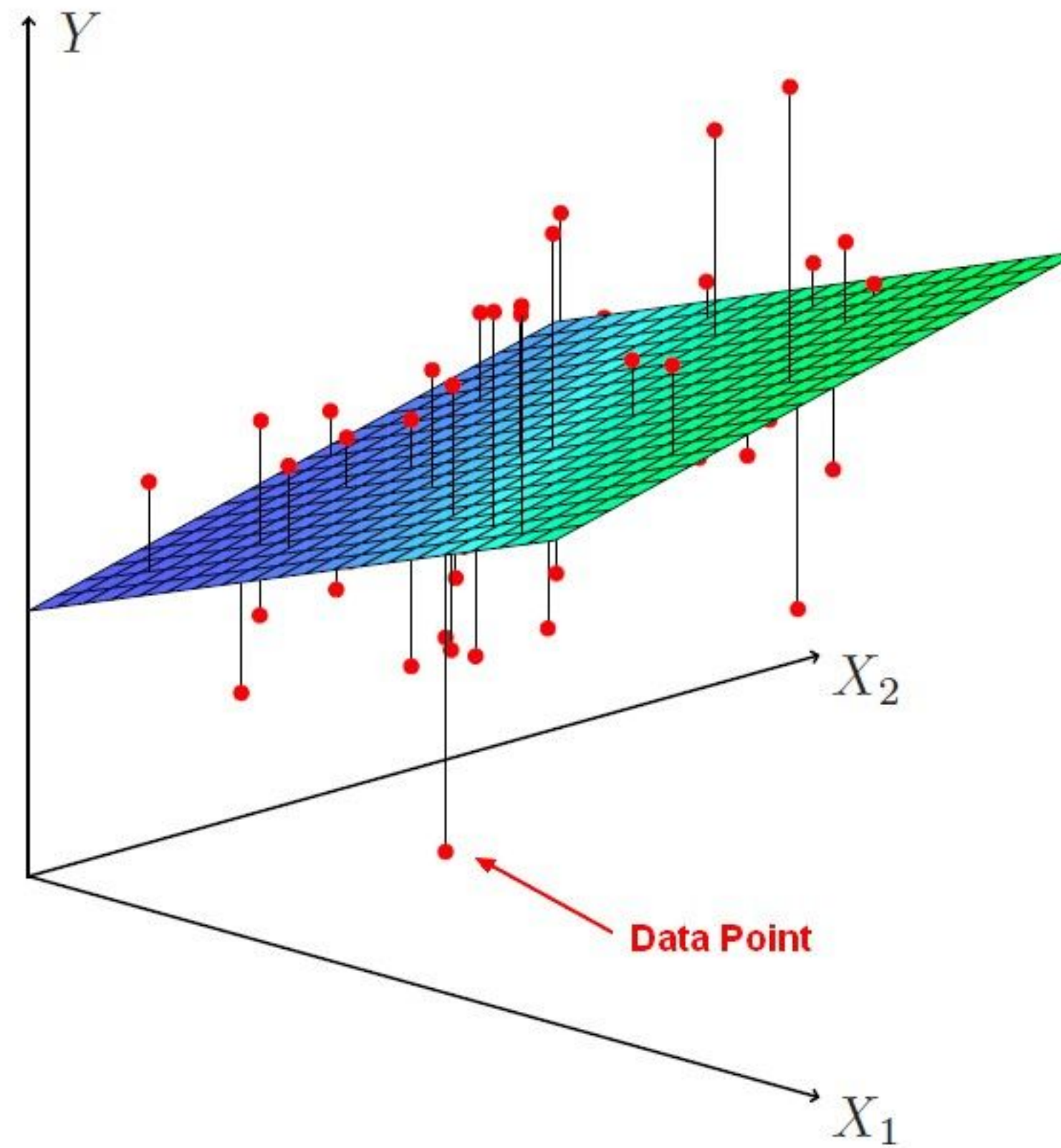
So what about the
main dataset? And
its 12 features?



Higher-dimensional linear regression



Higher-dimensional linear regression



$$\hat{y} = w_0 + w_1x_1 + w_2x_2 + w_3x_3 + \dots + w_nx_n$$

Hyperplanes!

More dimensions, similar
math.

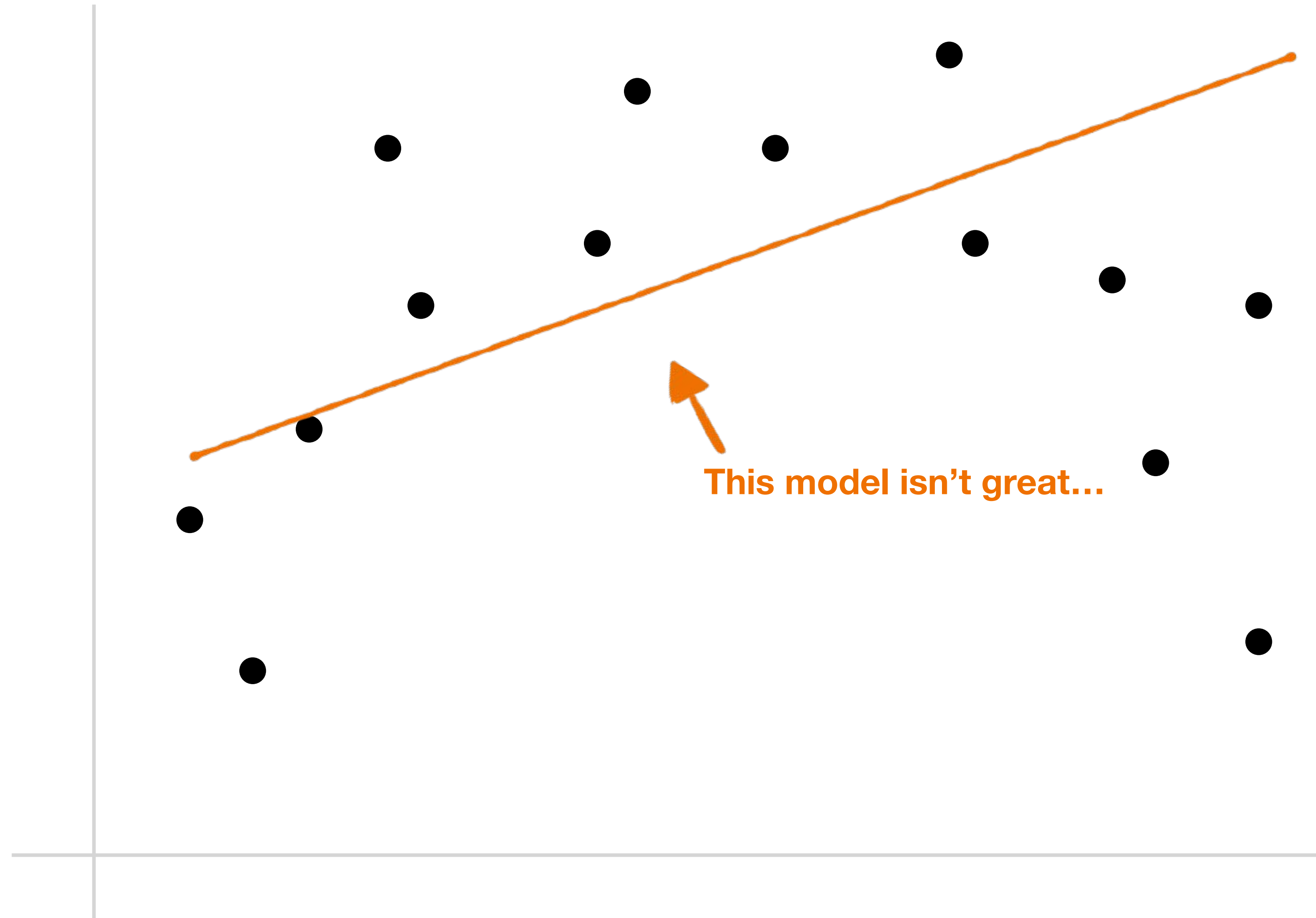
One key assumption we make

- **dataset linearity**

What if our data
doesn't have a
linear
representation?



The assumption fails...



■ model is unable to capture relationship between variables

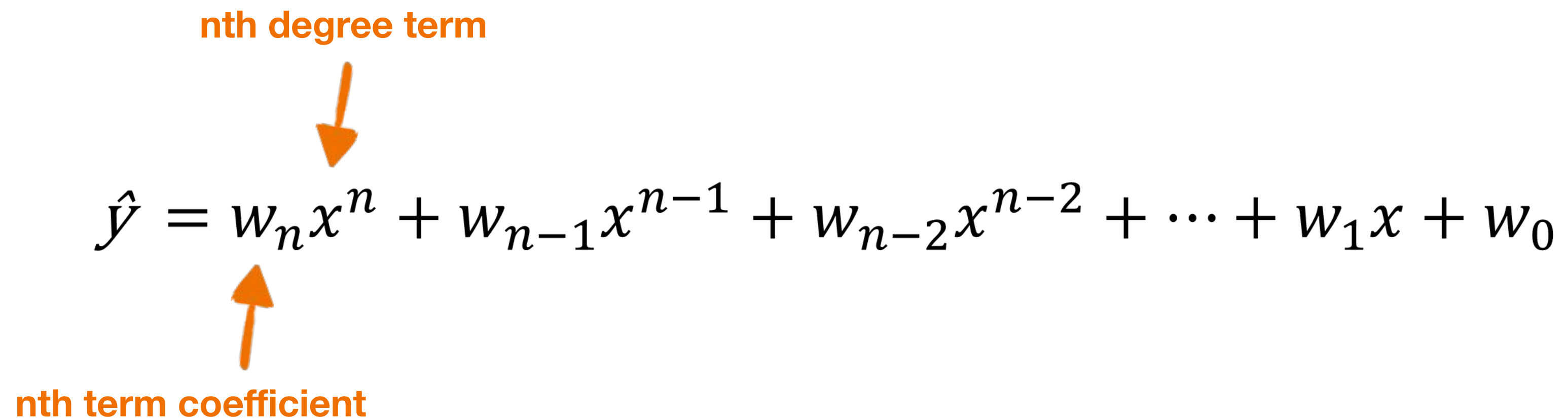
We need to make our model
more powerful!

POLYNOMIAL REGRESSION

Polynomial Regression

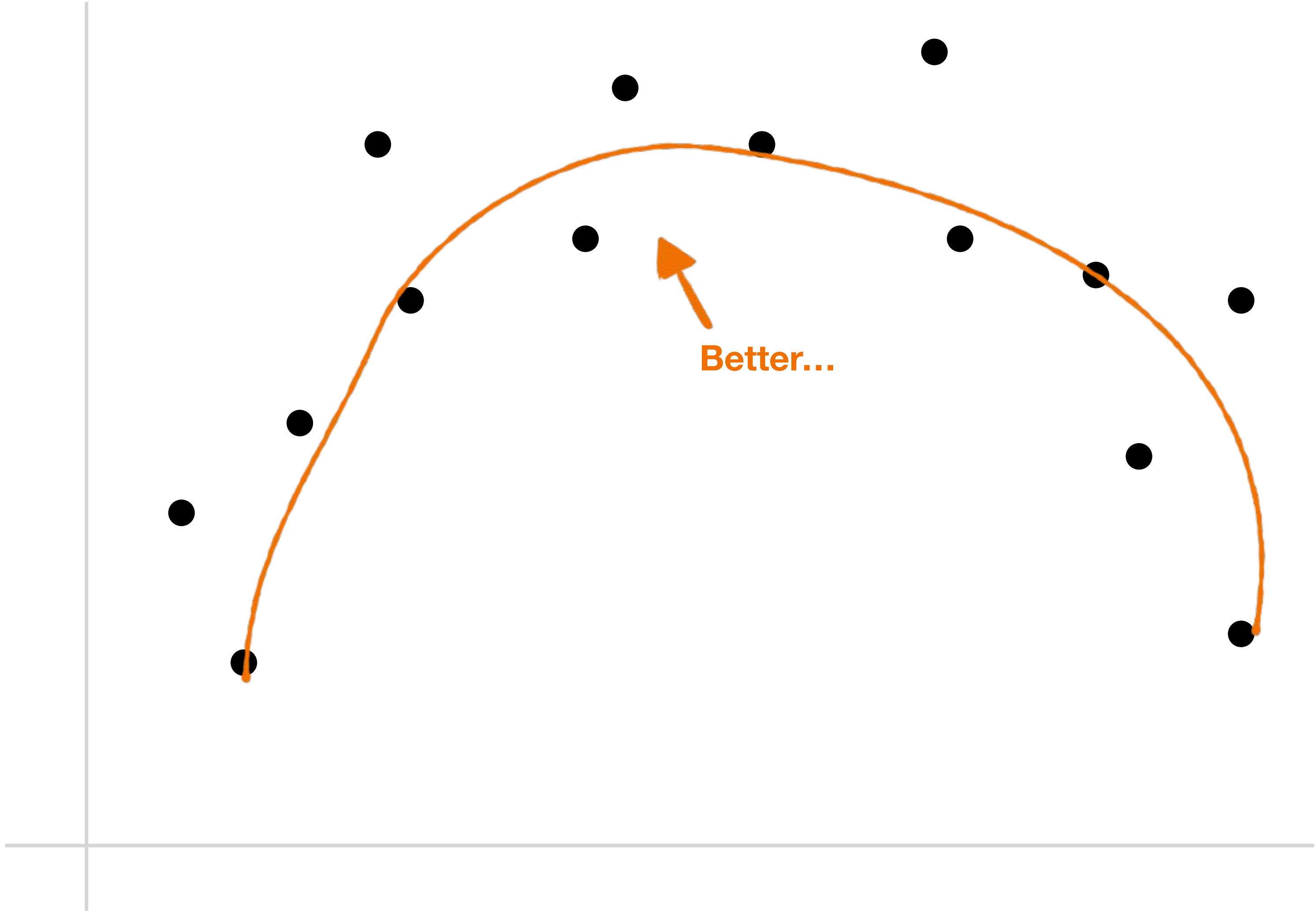
- introducing **higher-dimensional terms** to add **curvature**

nth degree term

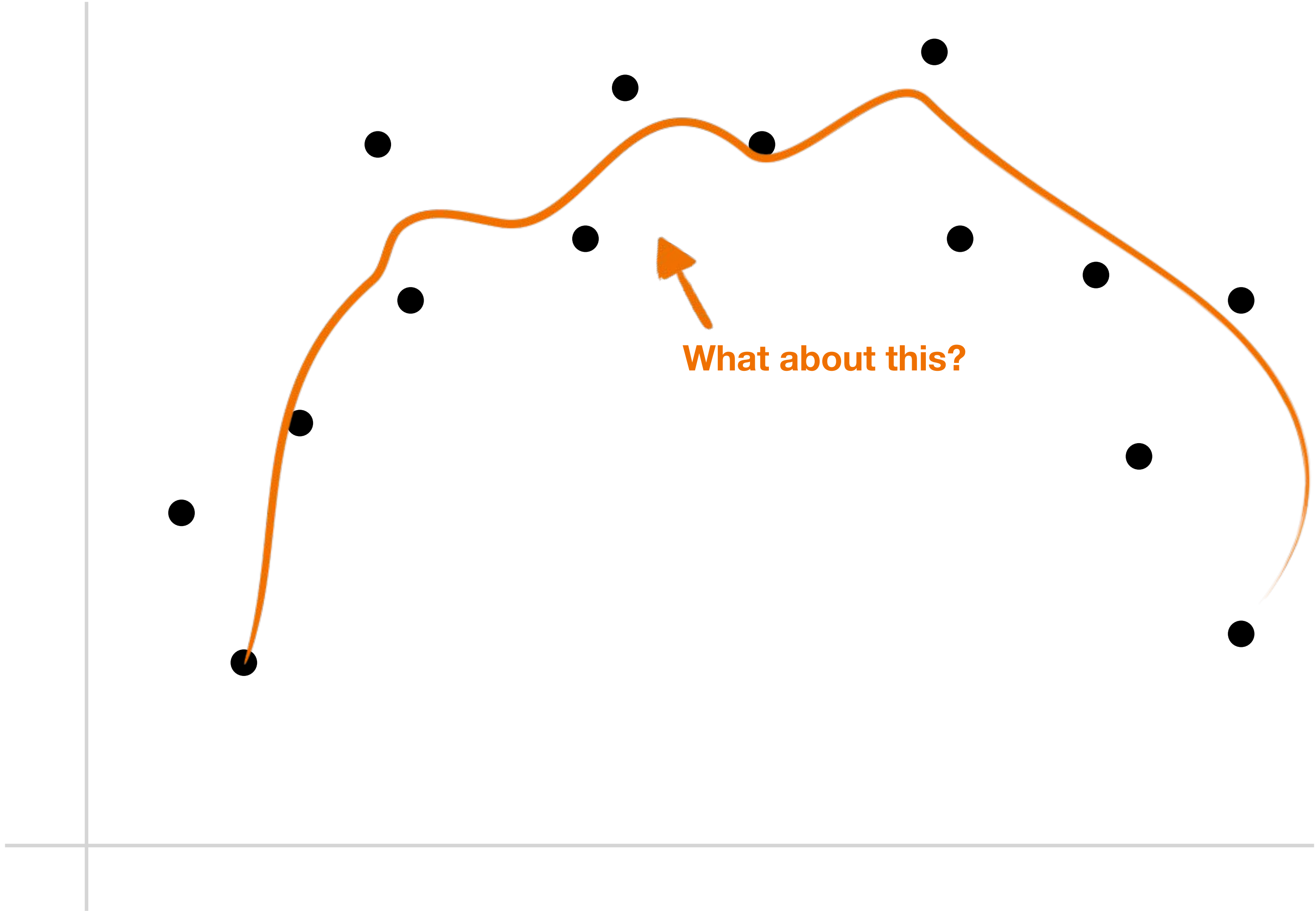

$$\hat{y} = w_n x^n + w_{n-1} x^{n-1} + w_{n-2} x^{n-2} + \dots + w_1 x + w_0$$

nth term coefficient

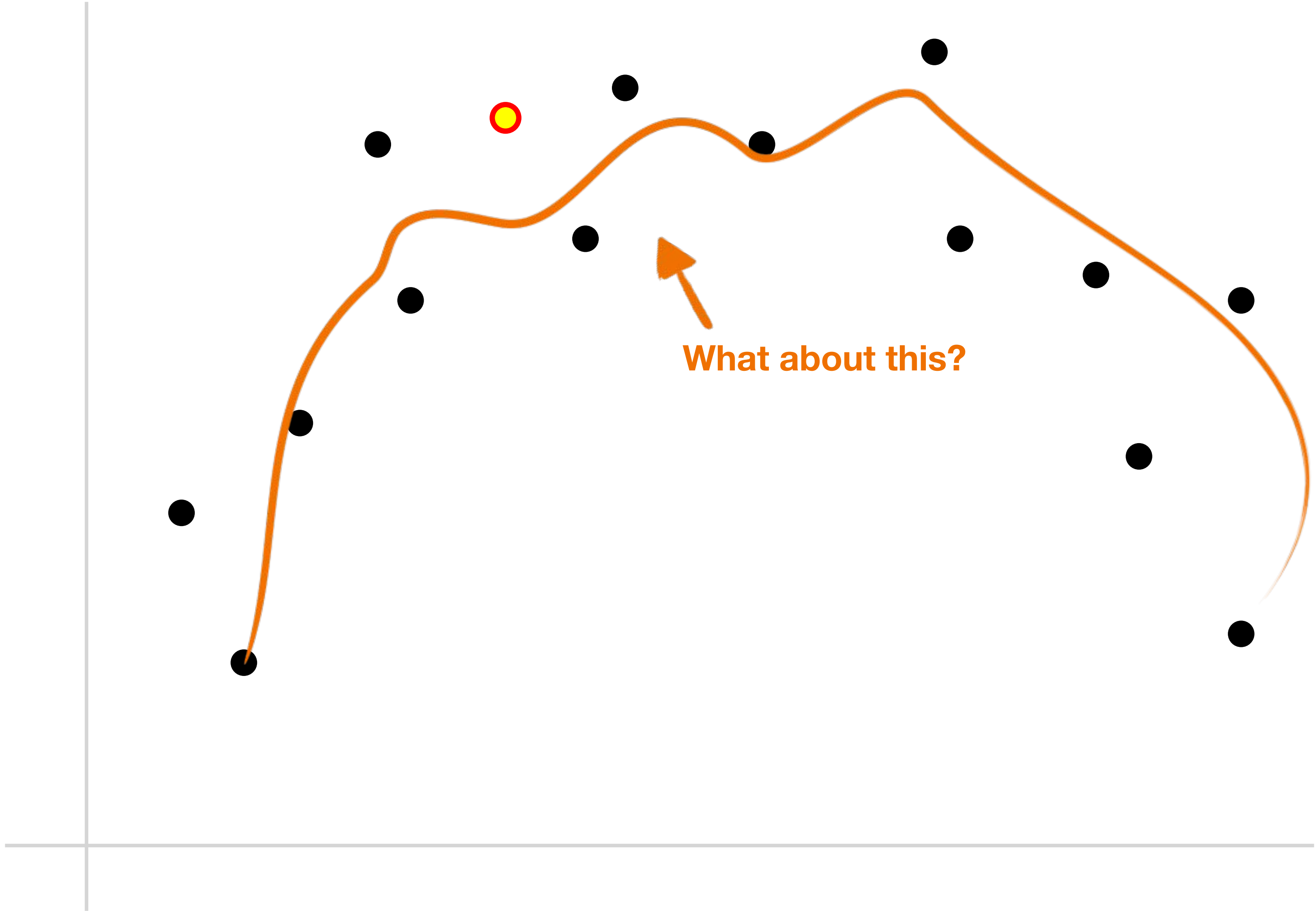
Polynomial Regression



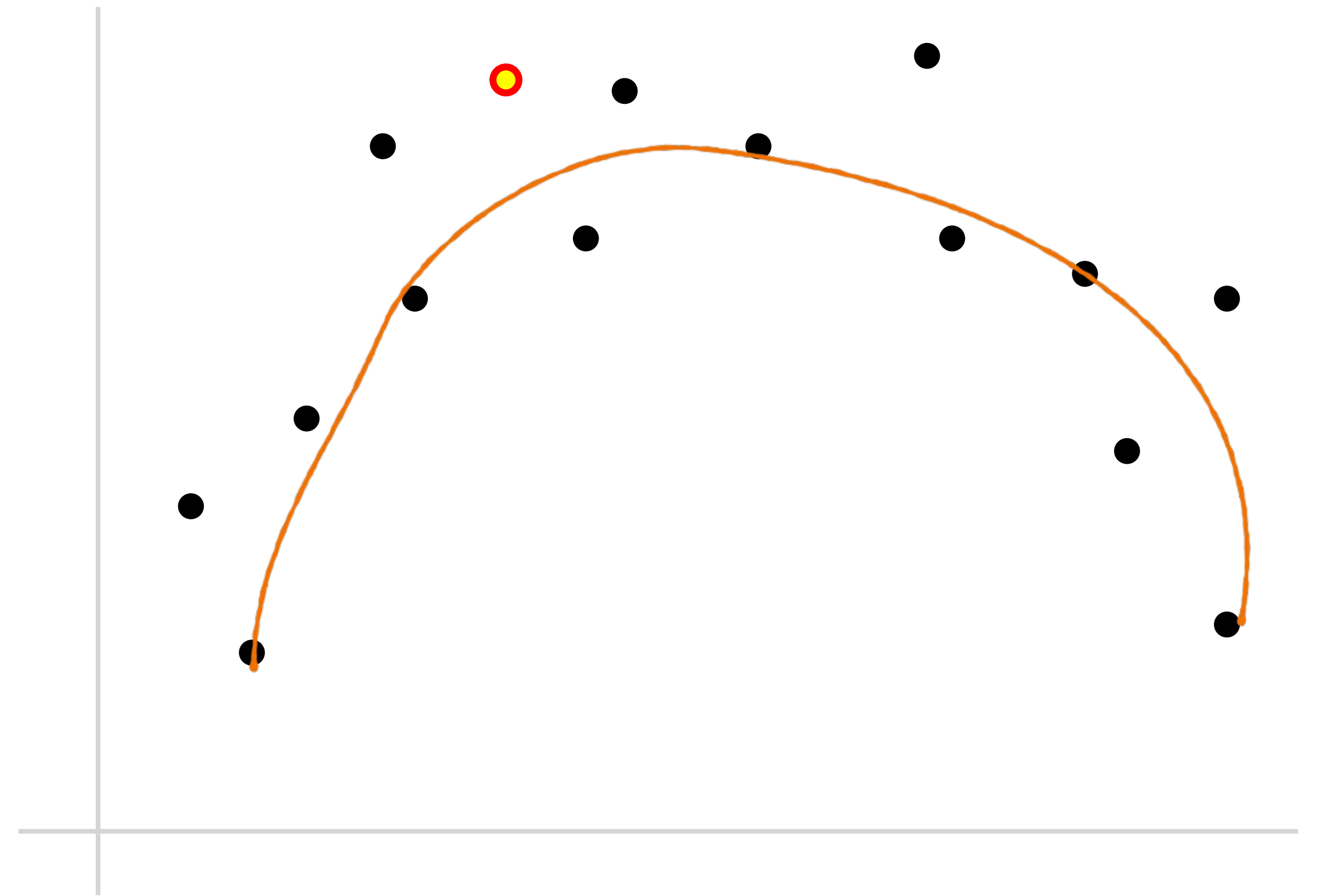
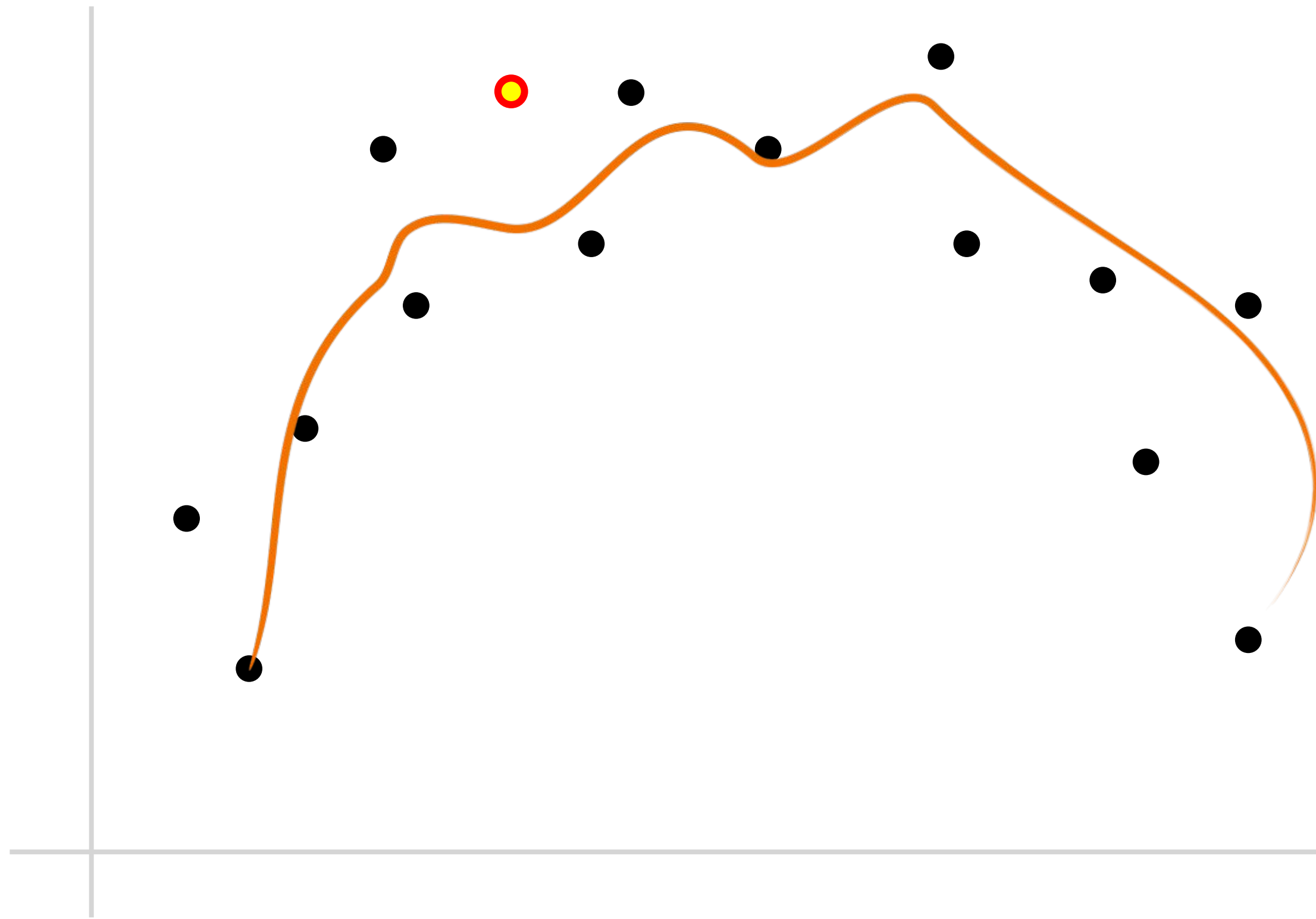
Polynomial Regression



Polynomial Regression

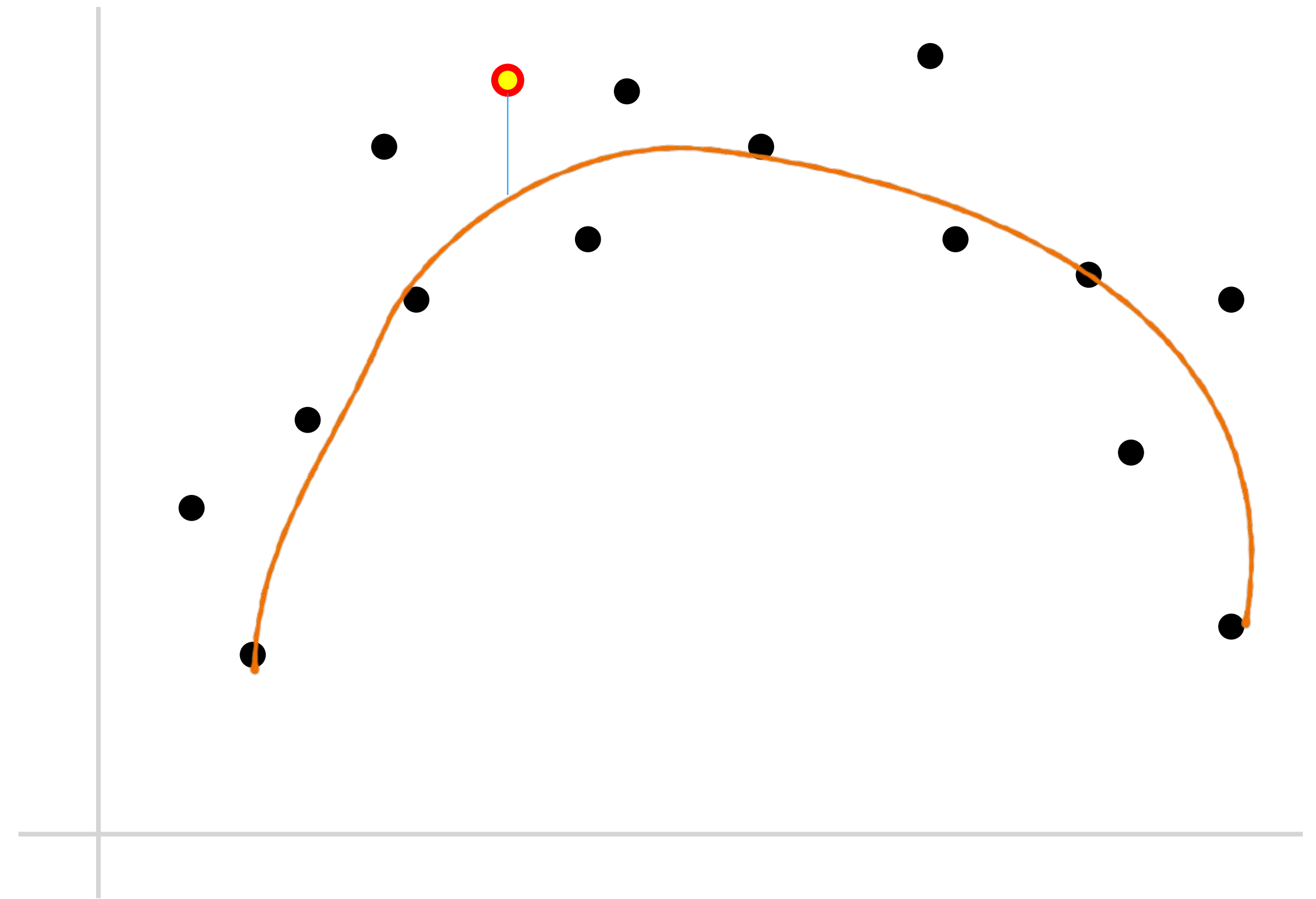
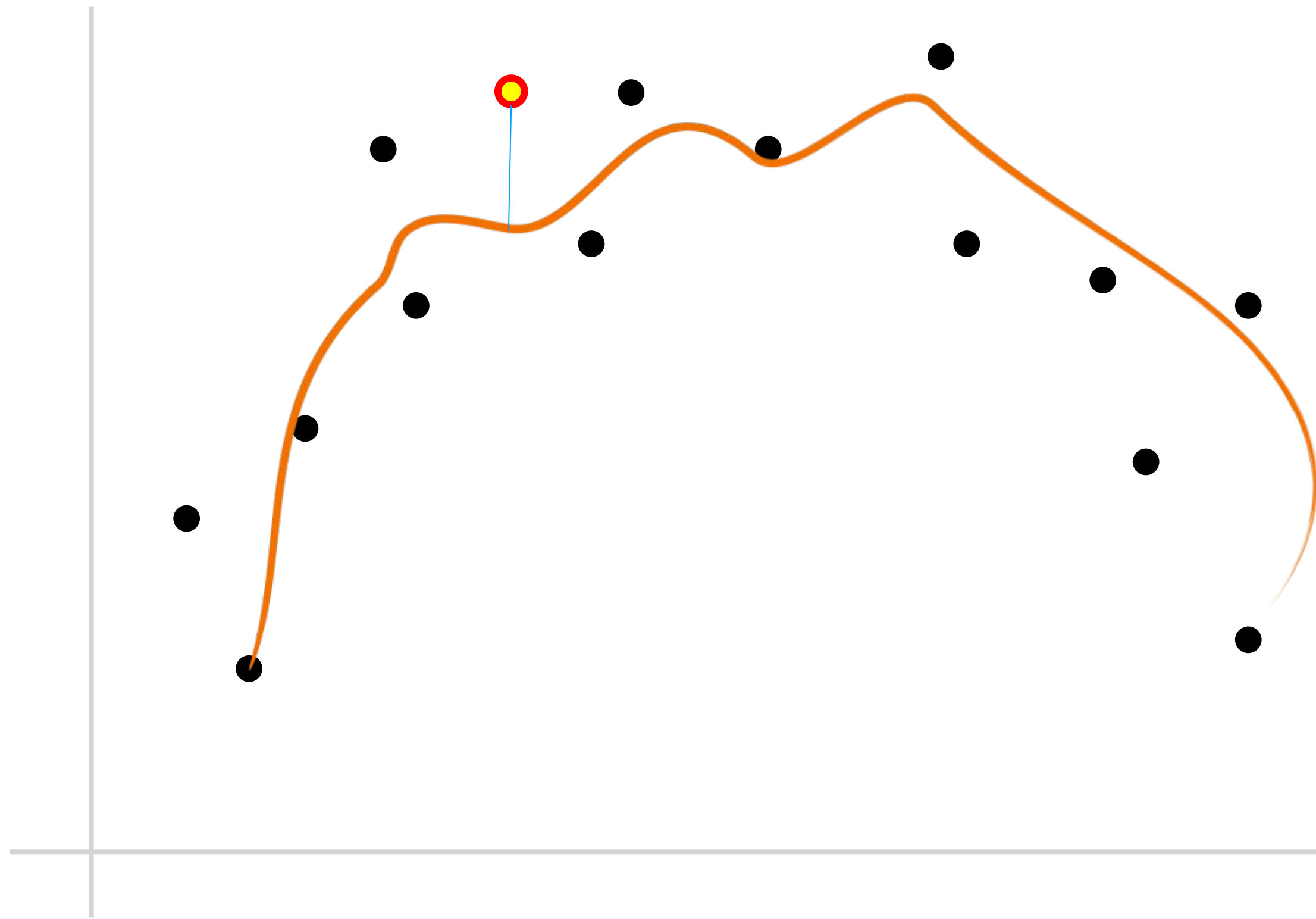


Polynomial regression - Which one is better?



Polynomial Regression

A significant difference!



Polynomial Regression



Calculating Error

- The **same** cost function!

$$\text{COST} = \frac{1}{n} \sum_{i=0}^n (\hat{y}^{(i)} - y^{(i)})^2$$

Sum of all of the squared differences

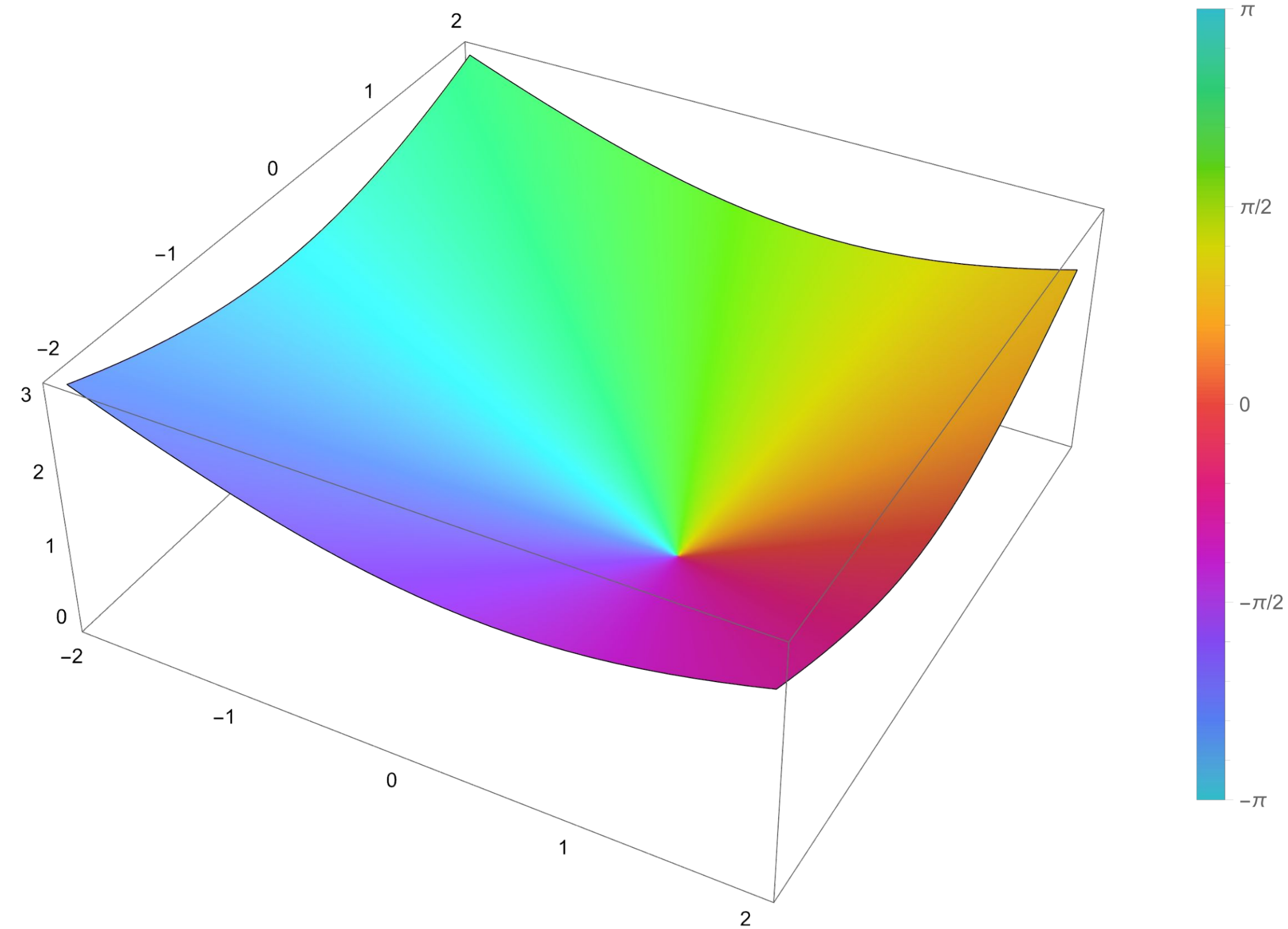
Square it!

Divide by number of samples: mean error, not total error

Difference between predicted and actual

Higher-dimensional polynomial regression

- **More powerful** (and complex) models!



Higher-dimensional polynomial regression

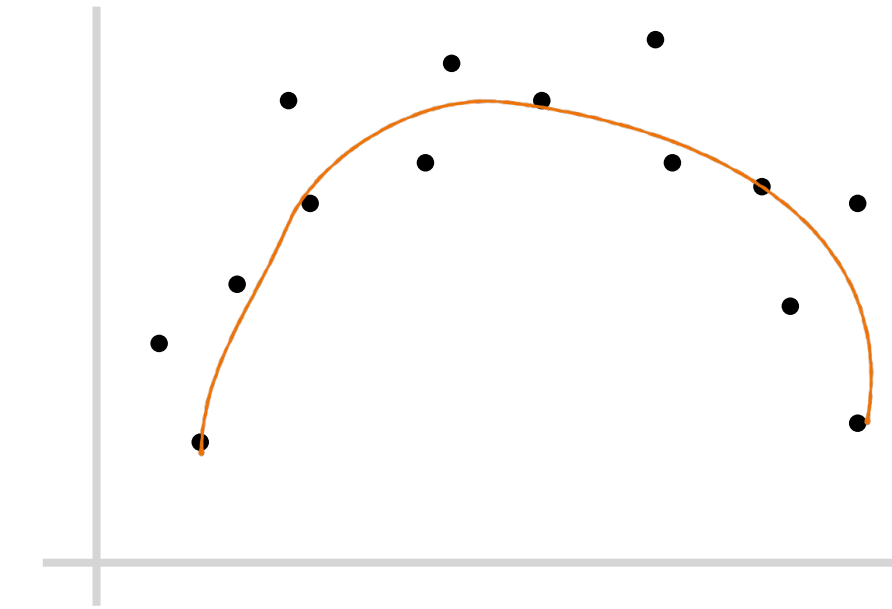
- **More powerful** (and complex) models!

$$\hat{y} = w_1x_1 + w_2x_2^2 + w_3x_3^3 + \cdots + w_nx_n^n + \cdots + w_ix_1x_2 + \cdots$$

Higher-dimensional polynomial regression

- Don't worry, the math stays the same:

$$\hat{y} = w_n x^n + w_{n-1} x^{n-1} + w_{n-2} x^{n-2} + \dots + w_1 x + w_0$$



$$\text{COST} = \frac{1}{n} \sum_{i=0}^n (\hat{y}^{(i)} - y^{(i)})^2$$

What about
classification?



LOGISTIC REGRESSION

Logistic Regression

- Fixed acidity
- Volatile acidity
- Citric acid
- Residual sugar
- Chlorides
- Free sulfur dioxide
- Total sulfur dioxide
- Density
- pH
- Sulphates
- Alcohol



0
1

← a class

• categorical label outputs are named “**classes**”

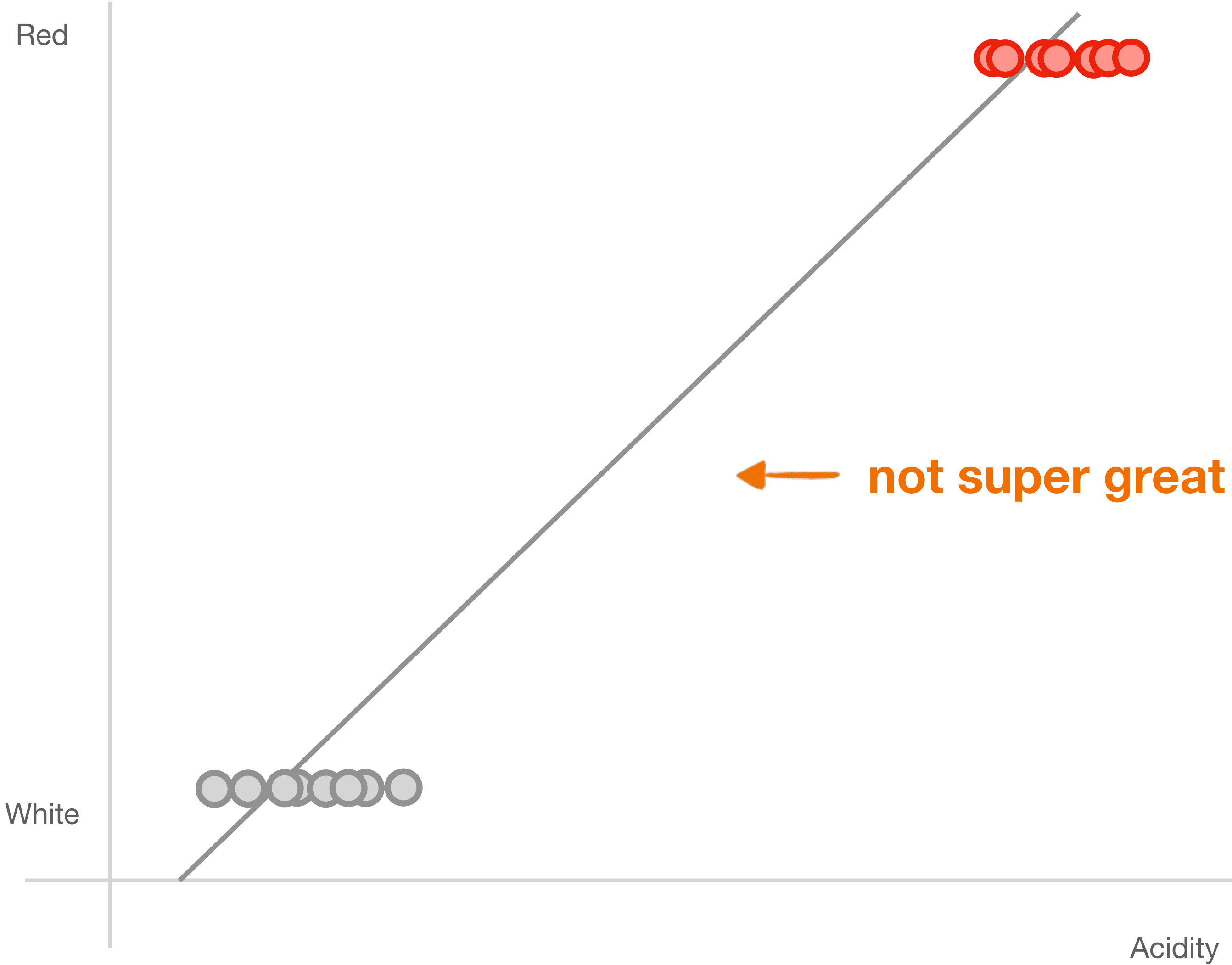
Logistic Regression

White = 0

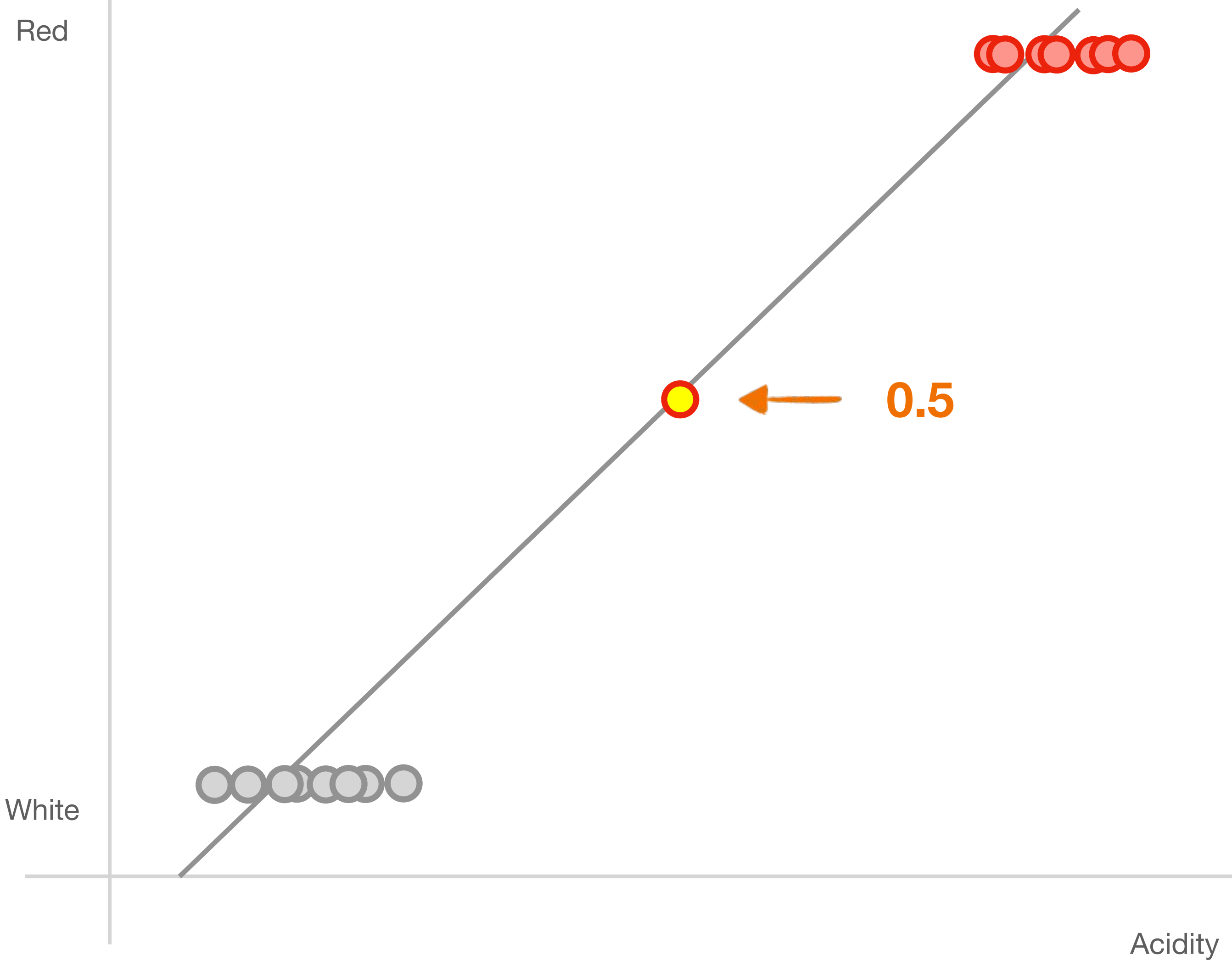
Red = 1

Why can't we just use linear or polynomial regression?

Logistic Regression

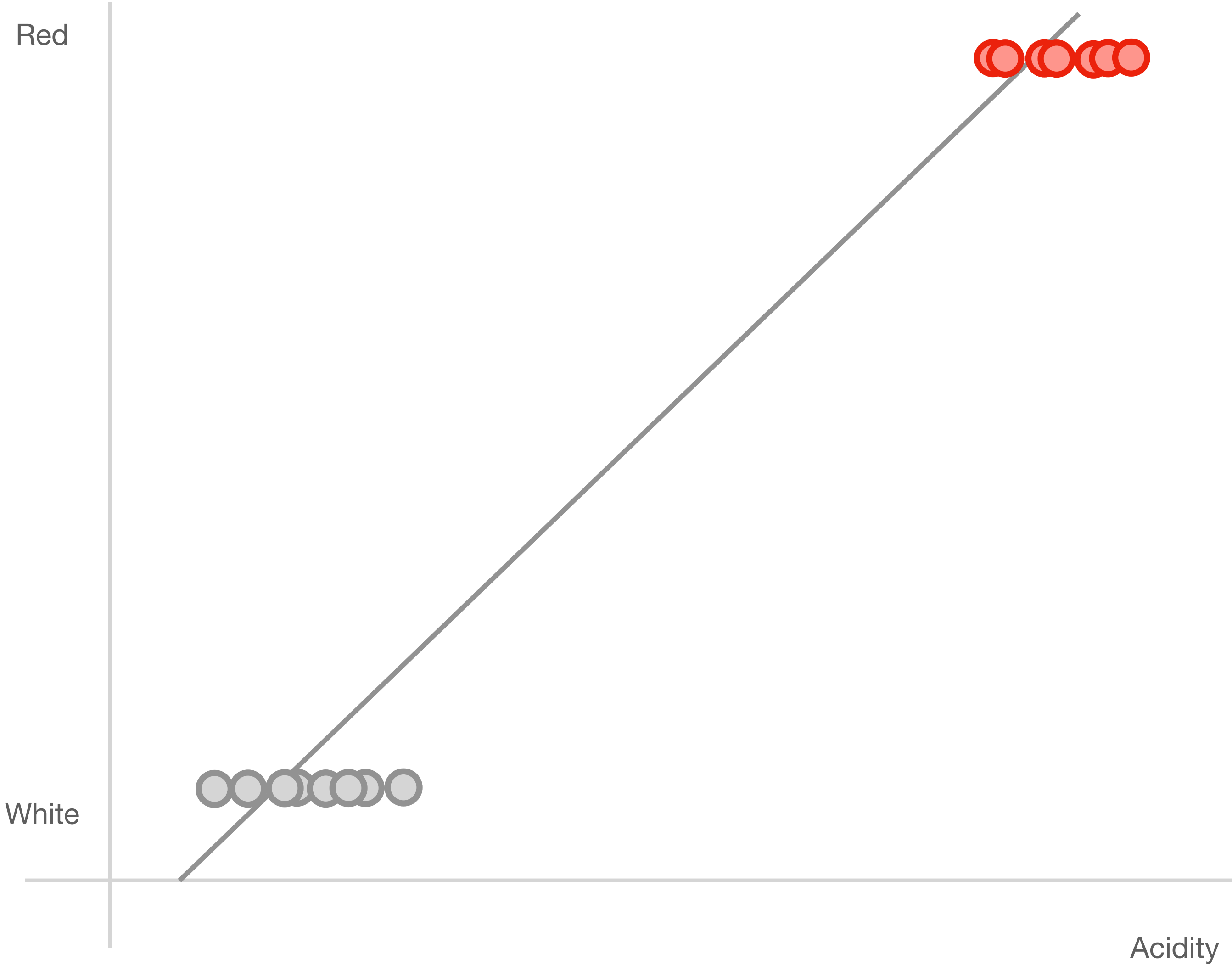


Logistic Regression

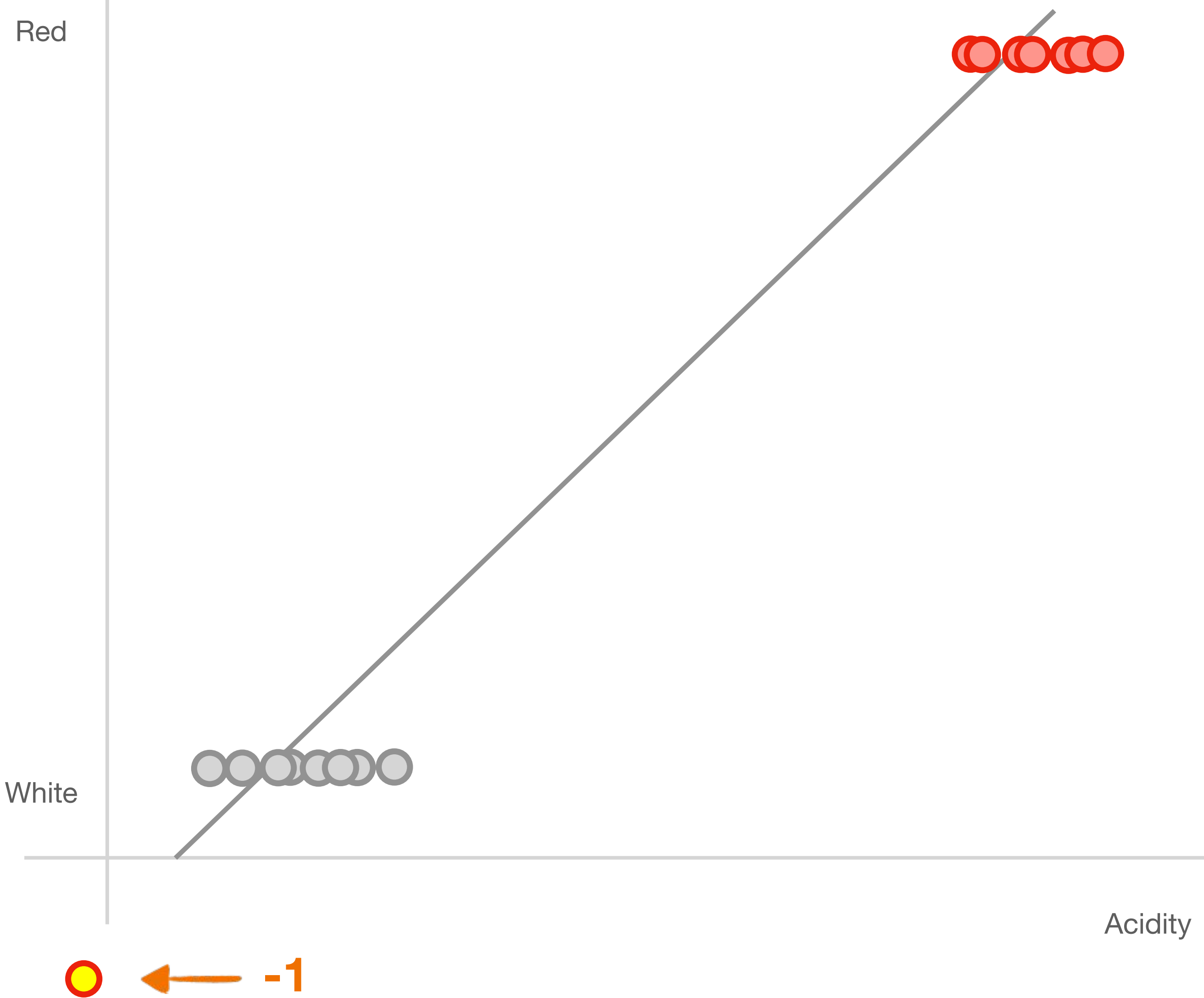


Logistic Regression

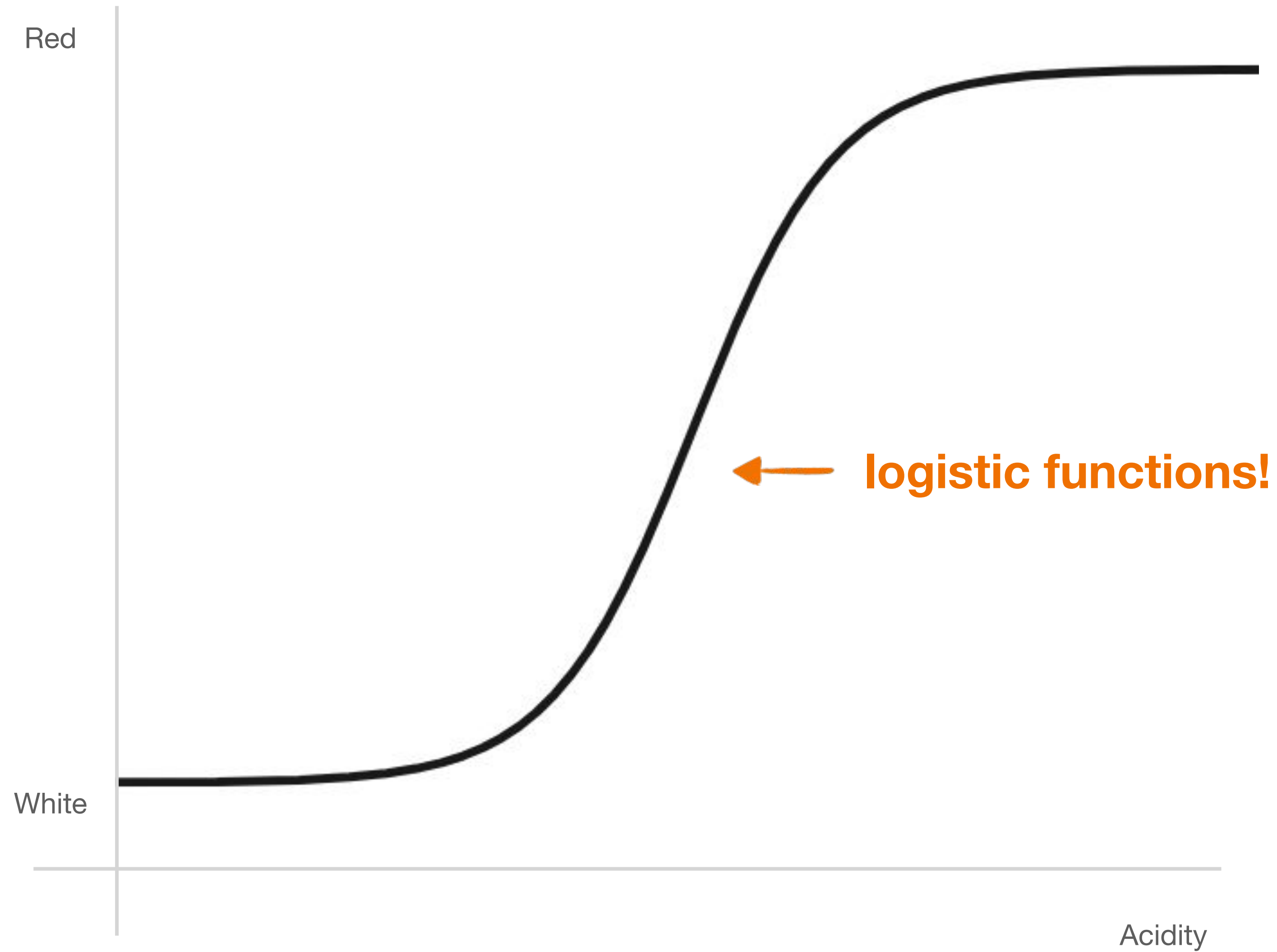
○ ← 2



Logistic Regression



Logistic Regression



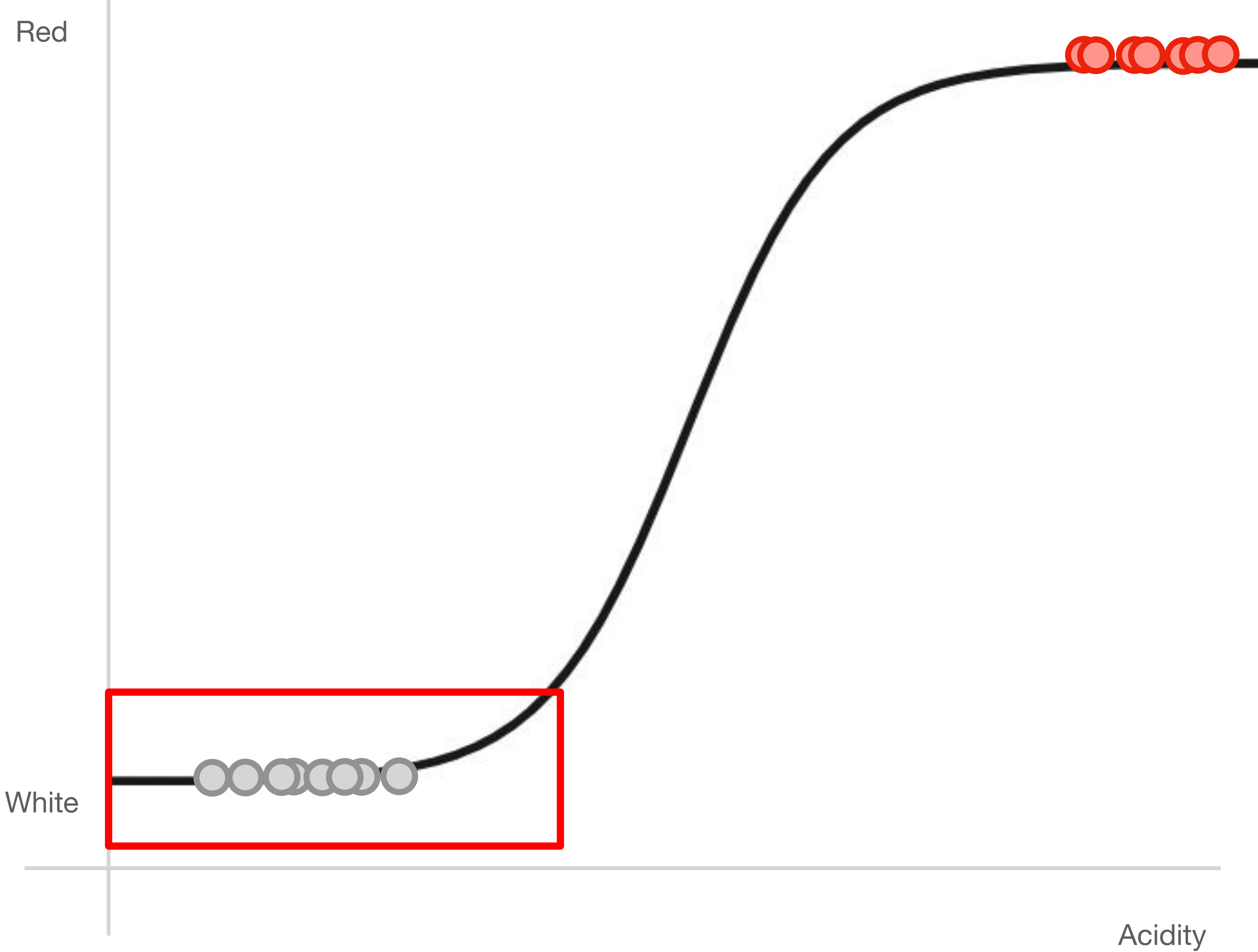
Logistic Regression

A logistic function:

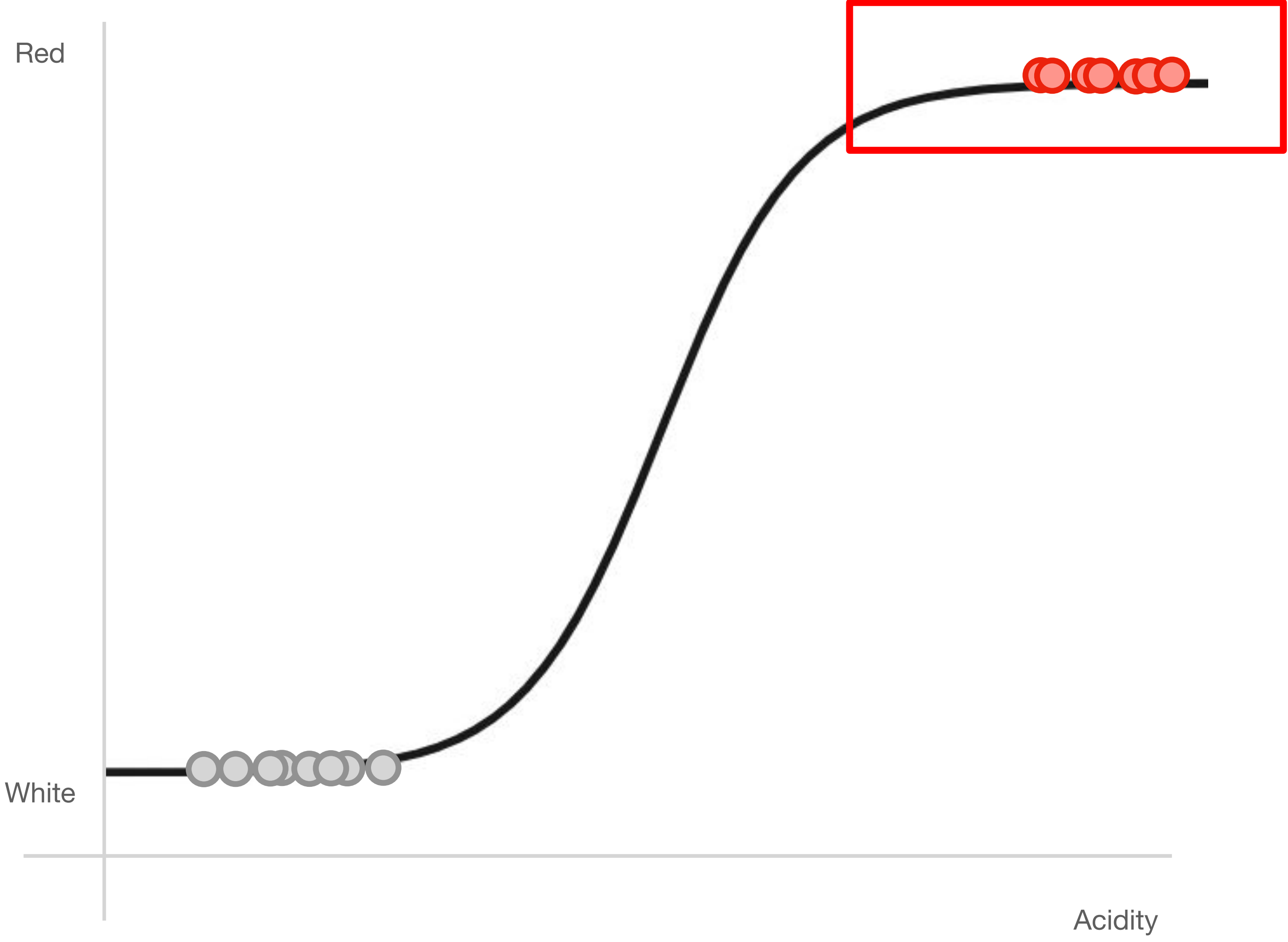
$$\hat{y} = \frac{1}{1 + e^{-w_0 + w_1 x_1 + \dots + w_n x_n}}$$

↑
our input features
are here

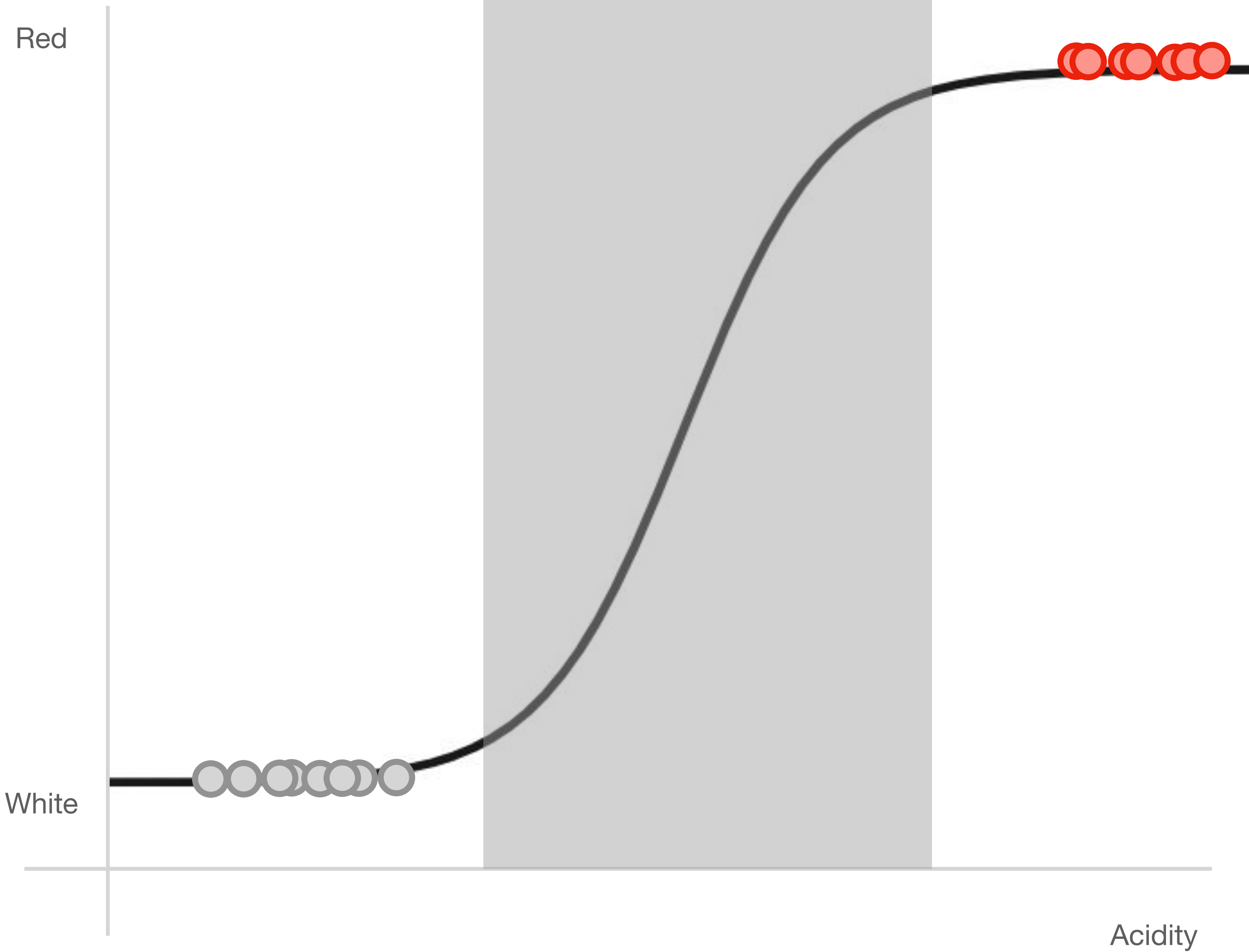
Logistic Regression



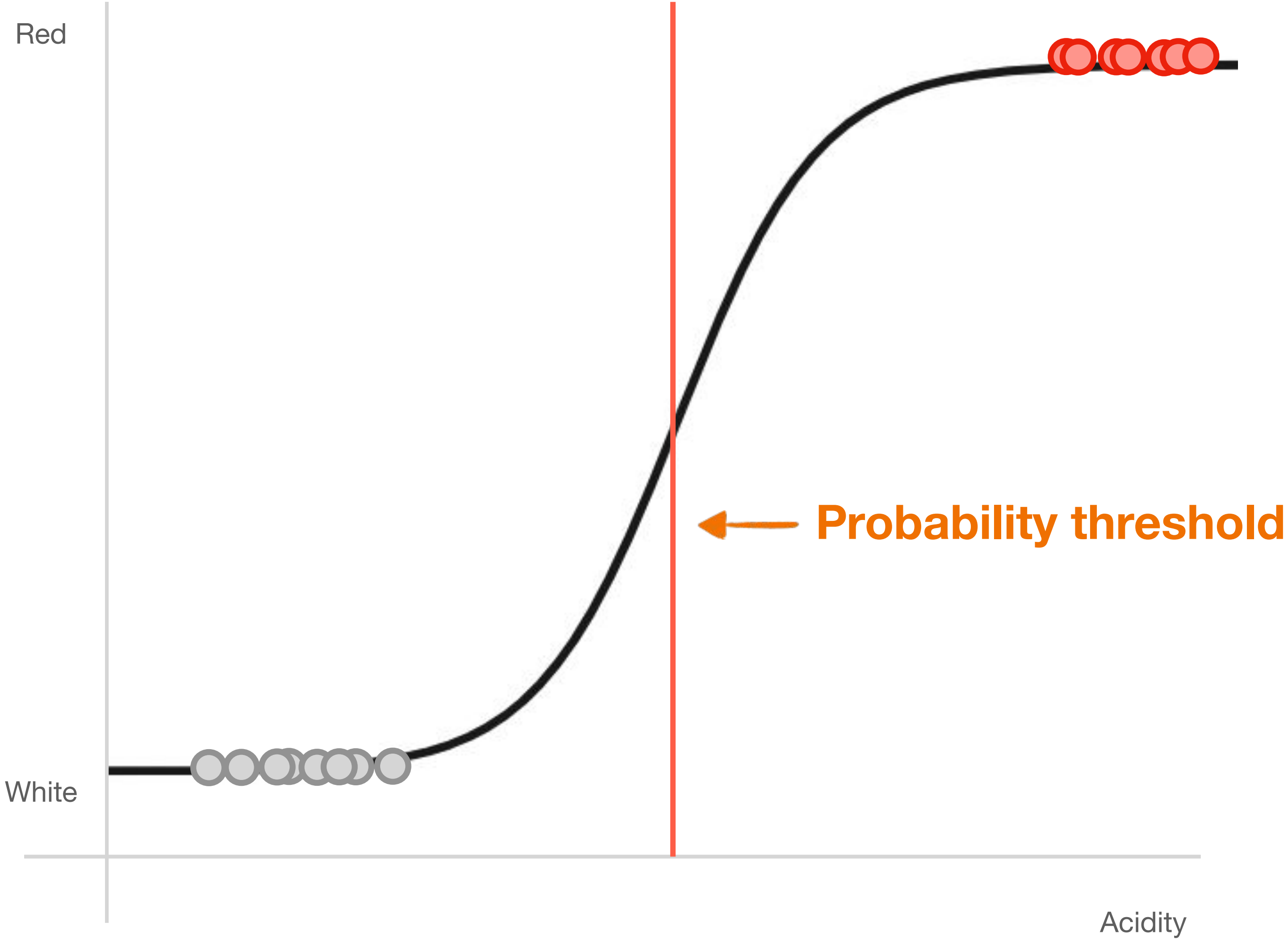
Logistic Regression



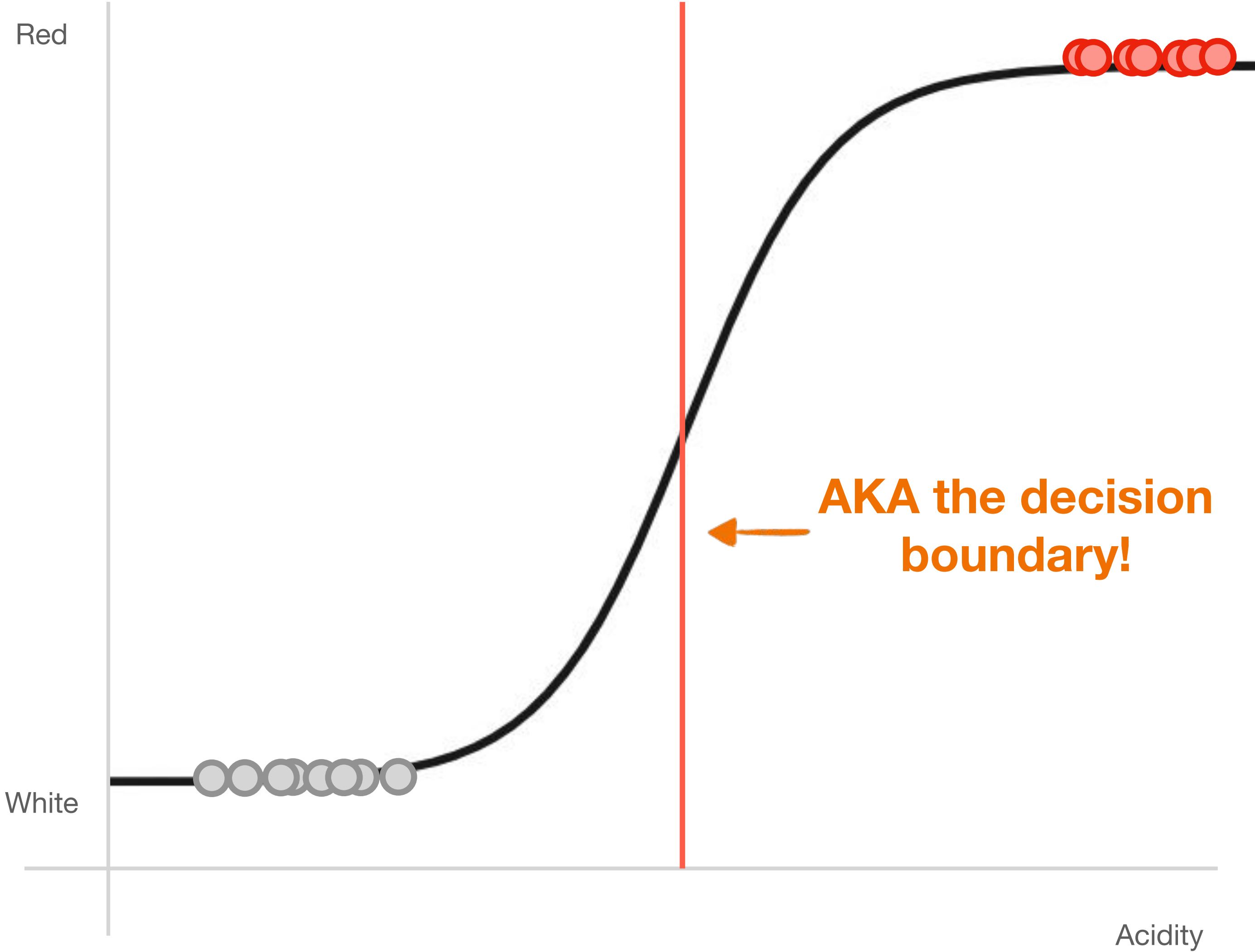
Logistic Regression



Logistic Regression



Logistic Regression



Logistic Regression

A new cost function...

$$\text{COST} = -\frac{1}{n} \sum_{i=0}^n (y^{(i)} \log(\hat{y}^{(i)}) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)}))$$

Logistic Regression

A new cost function...

$$\text{COST} = -\frac{1}{n} \sum_{i=0}^n (y^{(i)} \log(\hat{y}^{(i)})) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)})$$

Logistic Regression

A new cost function...

$$\text{COST} = -\frac{1}{n} \sum_{i=0}^n (y^{(i)} \log(\hat{y}^{(i)}) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)}))$$

Logistic Regression

A new cost function...

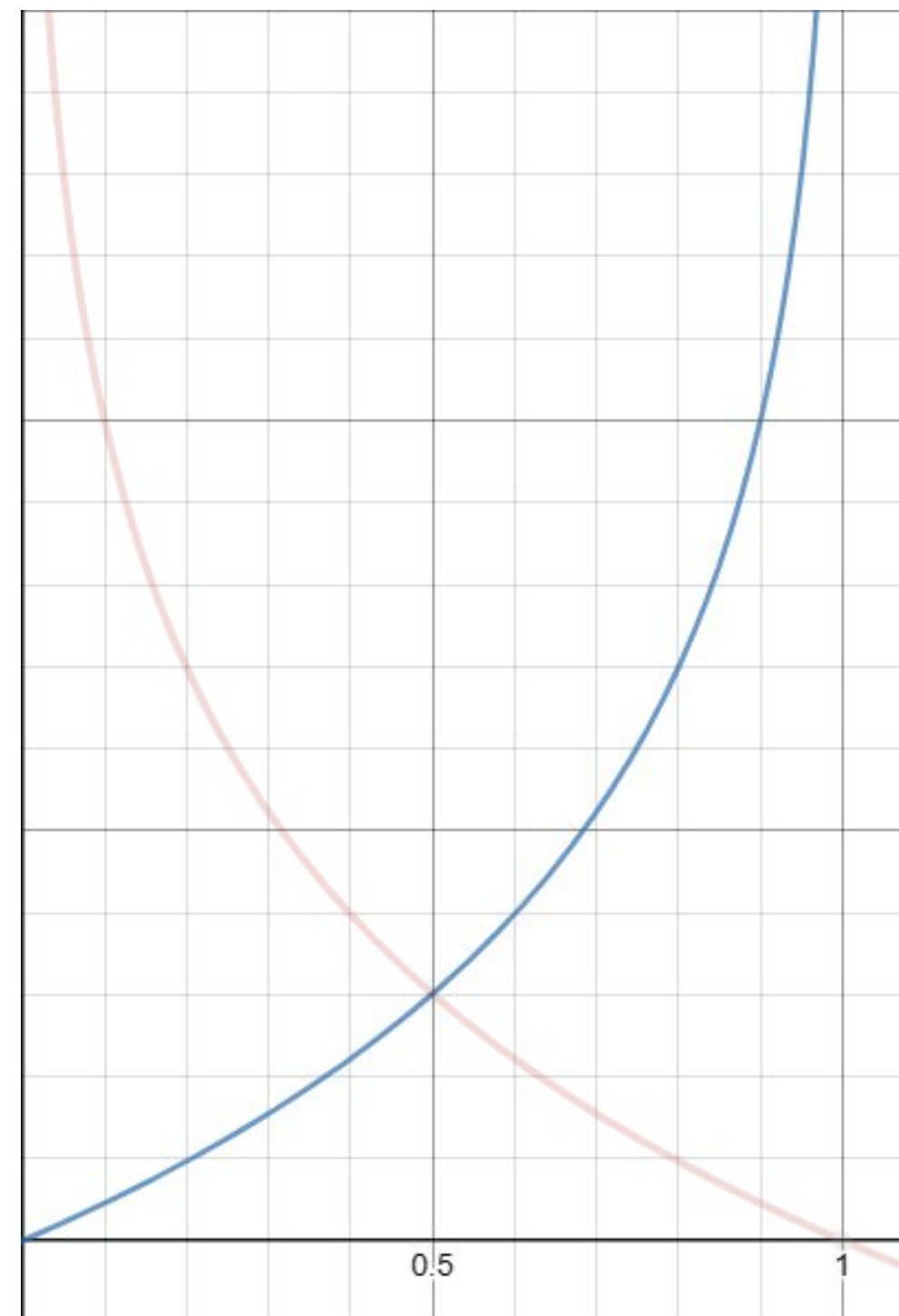
$$\text{COST} = -\frac{1}{n} \sum_{i=0}^n \left(y^{(i)} \log(\hat{y}^{(i)}) \right) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)})$$



Logistic Regression

A new cost function...

$$\text{COST} = -\frac{1}{n} \sum_{i=0}^n (y^{(i)} \log(\hat{y}^{(i)}) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)}))$$



But what if we have
more than two classes
for output?



Multi-class classification

White Red Champagne

One vs. One and One vs. All Classification

One vs. One Classification

One vs. One Classification

White

Red

One vs. One Classification

White	White
Red	Champagne

One vs. One Classification

White

White

Red

Red

Champagne

Champagne

One vs. One Classification

more pairs
voted for white!



White

White

Red

Red

Champagne

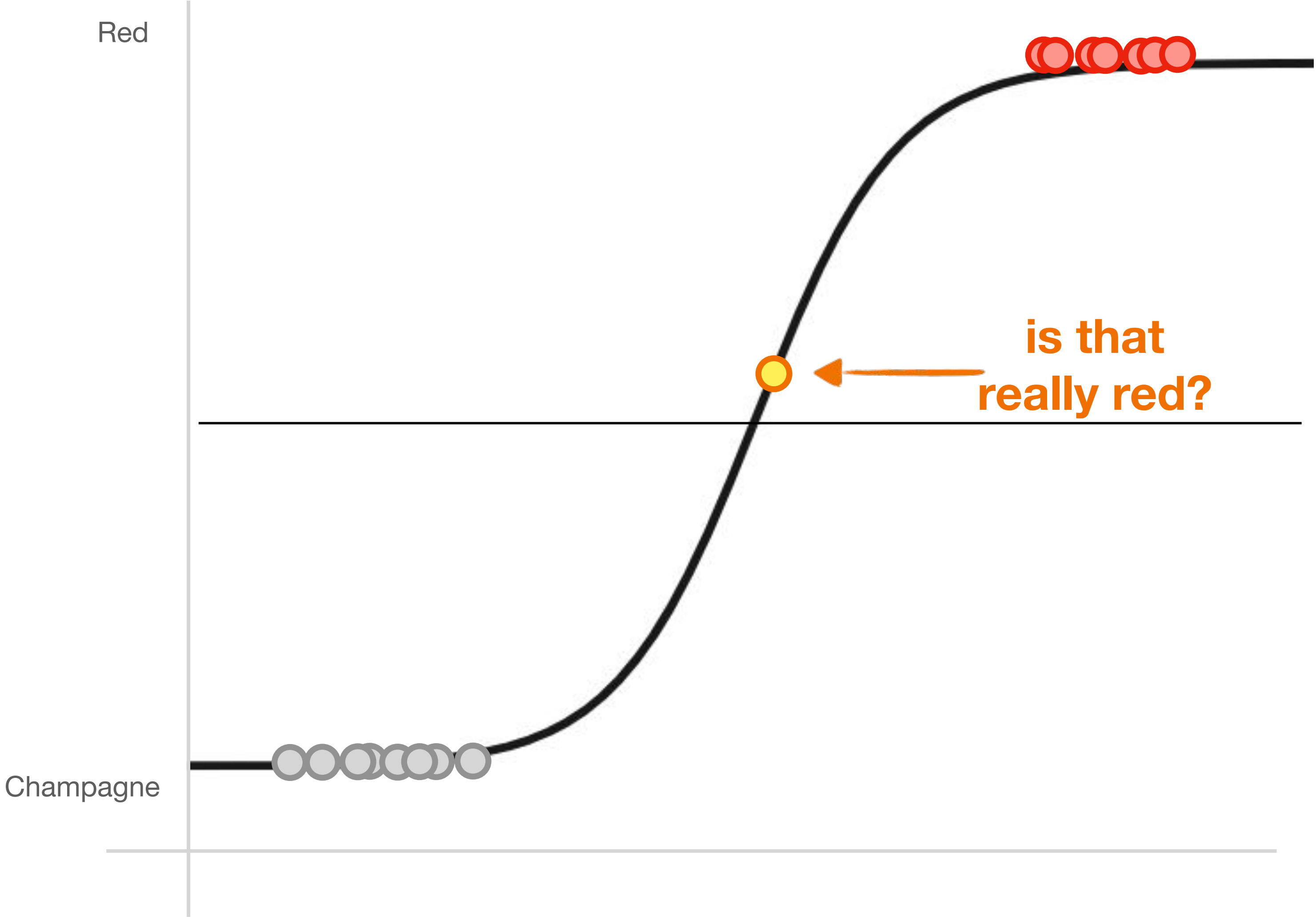
Champagne



**It's
white!**

- **One-vs-one** multiclass classification uses the most “voted for” class among paired models

Problems with One vs. One Classification



One vs. All Classification

White

Not white

Red

Not red

Champagne

Not champagne

One vs. All Classification

White

Not white

Red

Not red

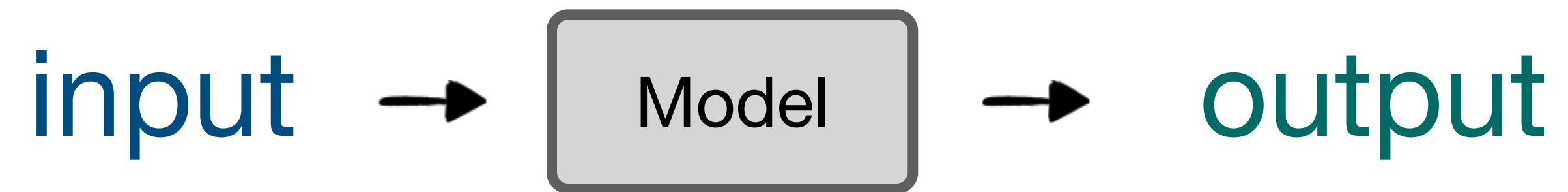
Champagne

Not champagne

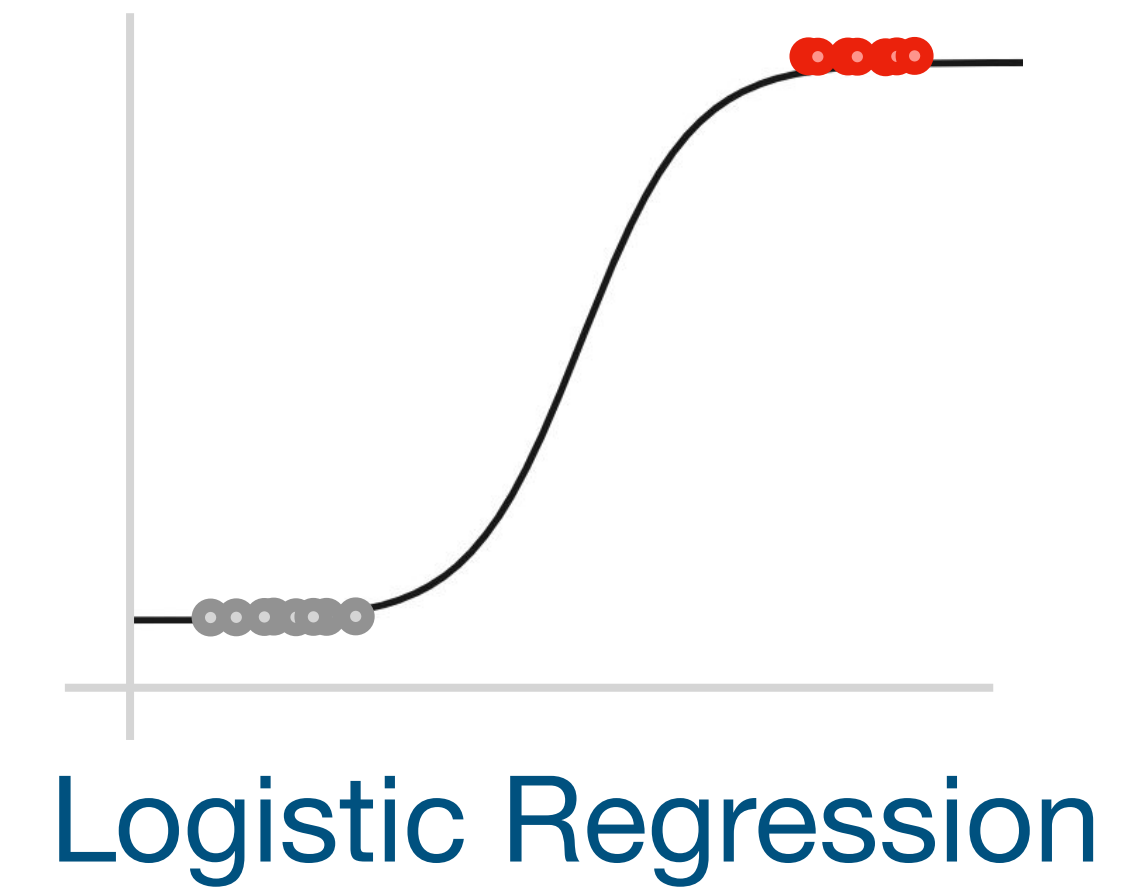
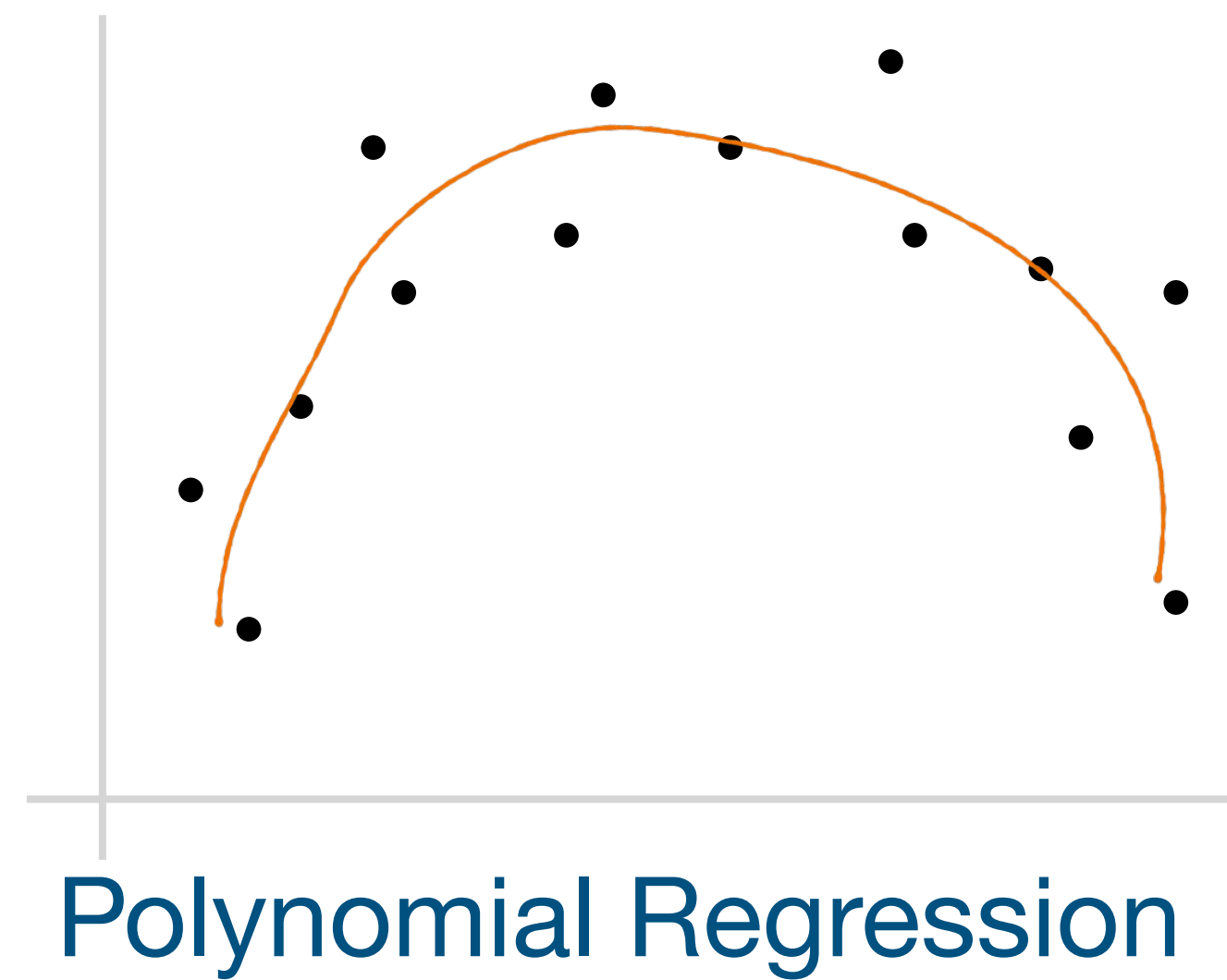
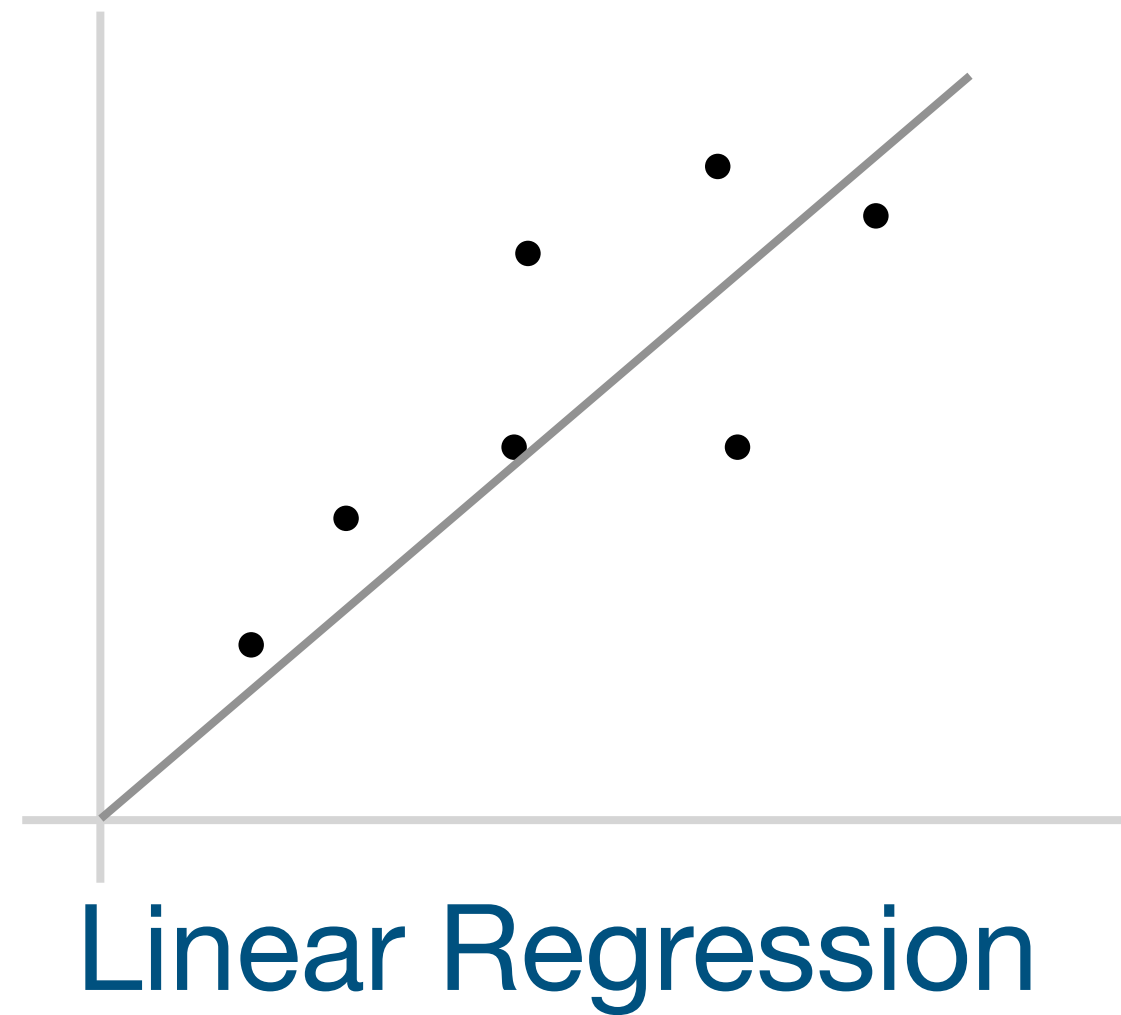
 **pick the answer with
highest probability**

A quick summary...

what *is* Artificial Intelligence?



Supervised Learning



Lab time!