





# Announcements

- HW2 due on **Friday September 26th at 11:59 PM**
- Midterm **Thursday October 23<sup>rd</sup> at 7:30 PM**

- Class roadmap:

ML Introduction
ML Unsupervised I
ML Unsupervised II
ML: Linear Regression

Machine Learning

# Outline

- What is machine learning?
- Supervised Learning
  - Classification
  - Regression
- Unsupervised Learning
  - Clustering
  - Self-Supervised Learning
- Reinforcement Learning



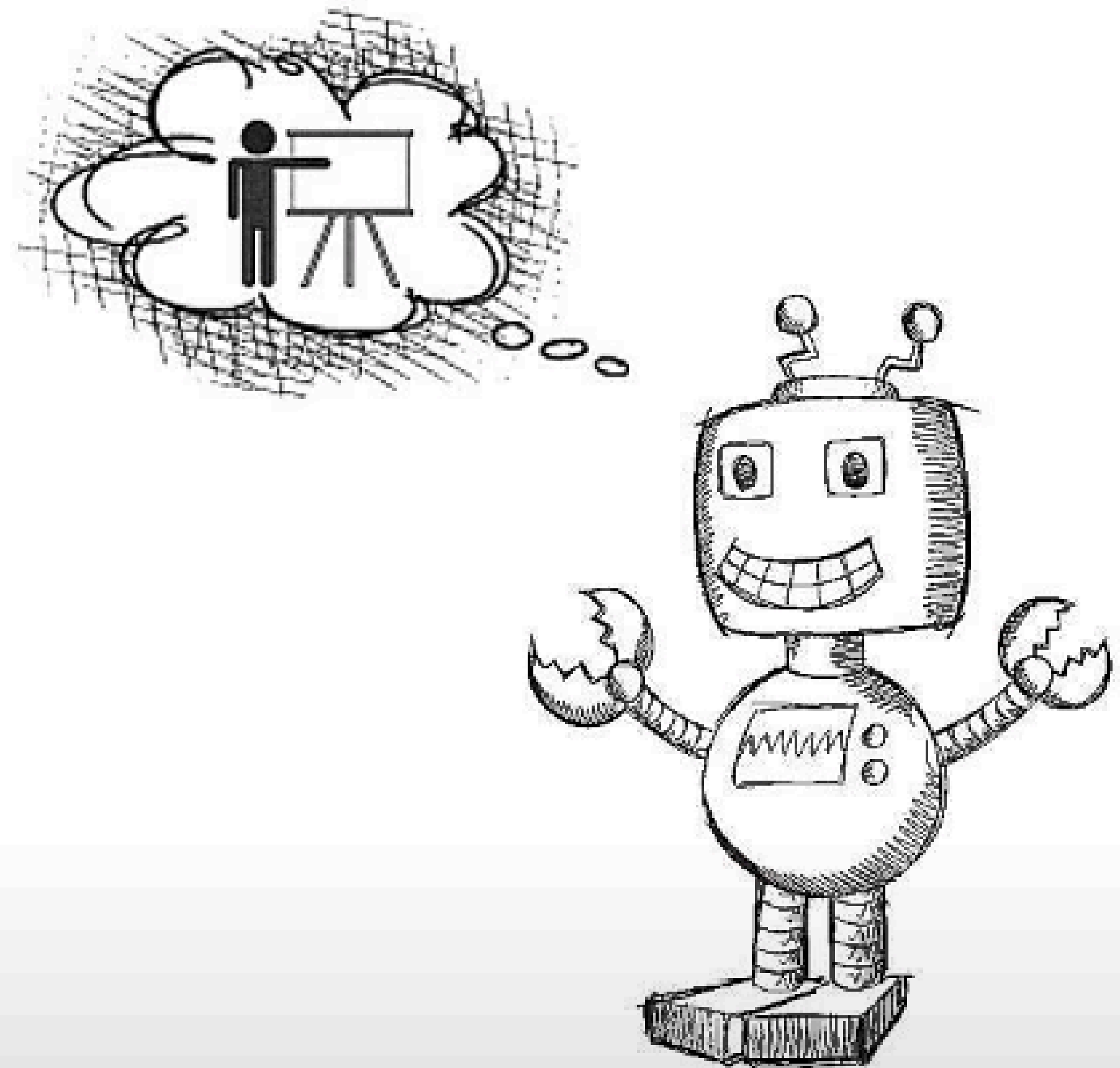


# Part I: What is machine learning?





**HUMANS LEARN FROM  
PAST EXPERIENCES**



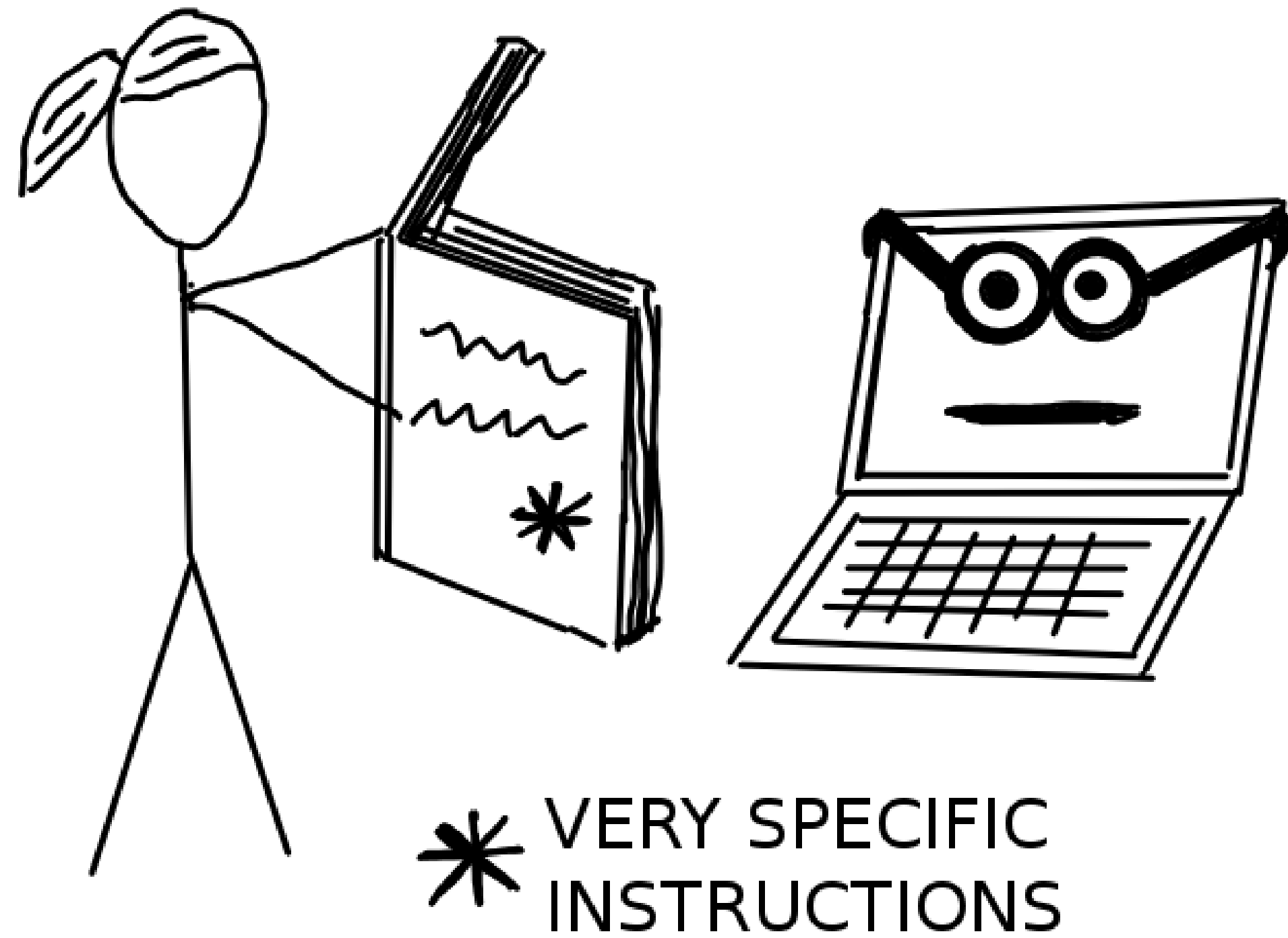
**MACHINES FOLLOW INSTRUCTIONS  
GIVEN BY HUMANS**

# What is **machine learning**?

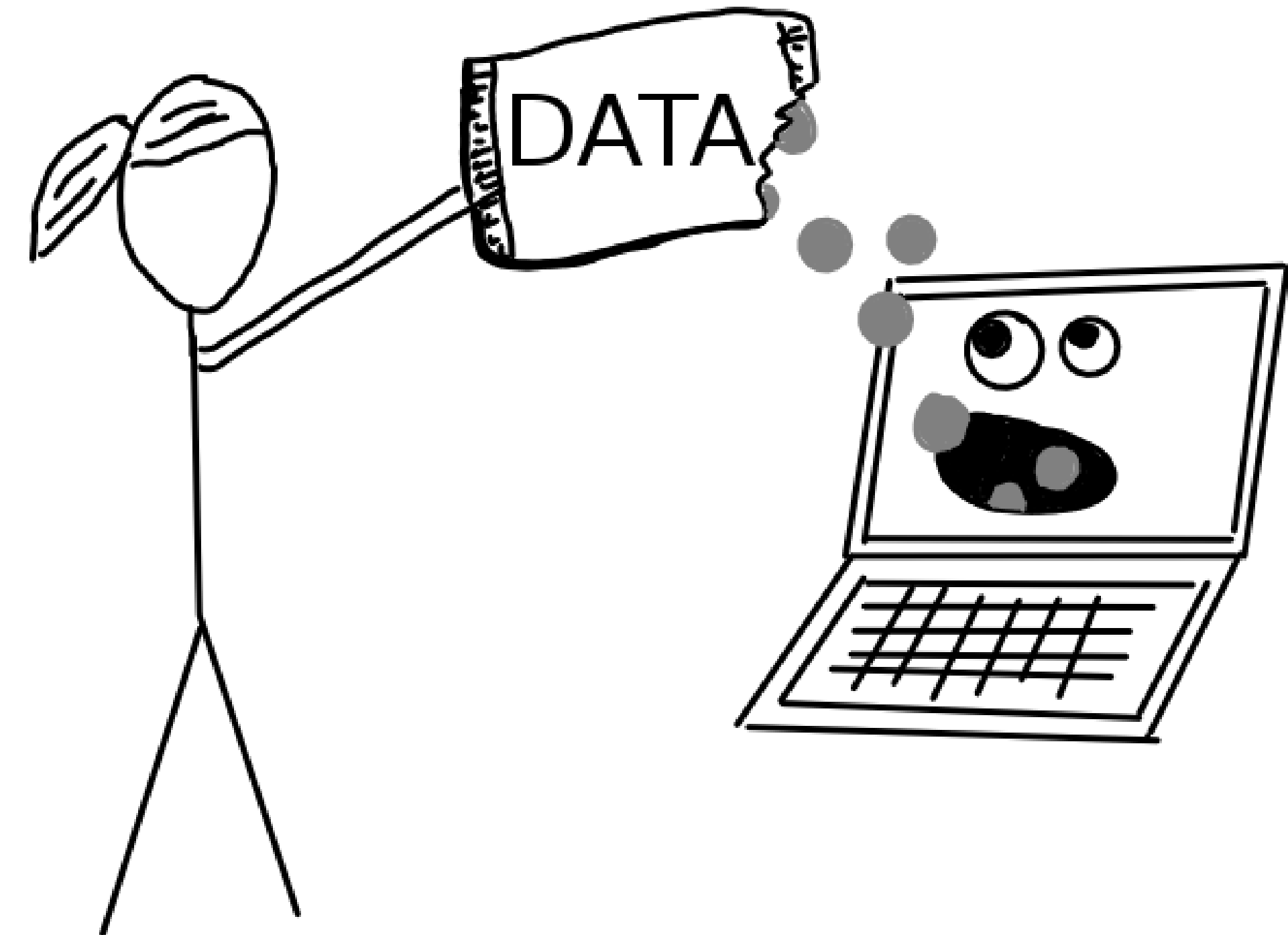
- Arthur Samuel (1959): Machine learning is the field of study that gives the computer the ability to learn **without being explicitly programmed**.



## Without Machine Learning



## With Machine Learning



# What is **machine learning**?

- Arthur Samuel (1959): Machine learning is the field of study that gives the computer the ability to learn **without being explicitly programmed**.
- Tom Mitchell (1997): A computer program is said to learn from **experience E** with respect to some class of **tasks T** and **performance measure P**, if its performance at tasks in T as measured by P, improves with experience E.





# Taxonomy of ML



Supervised  
Learning

Unsupervised  
Learning

Reinforcement  
Learning





## Part II: Supervised Learning



# Example 1: Predict whether a user likes a song or not



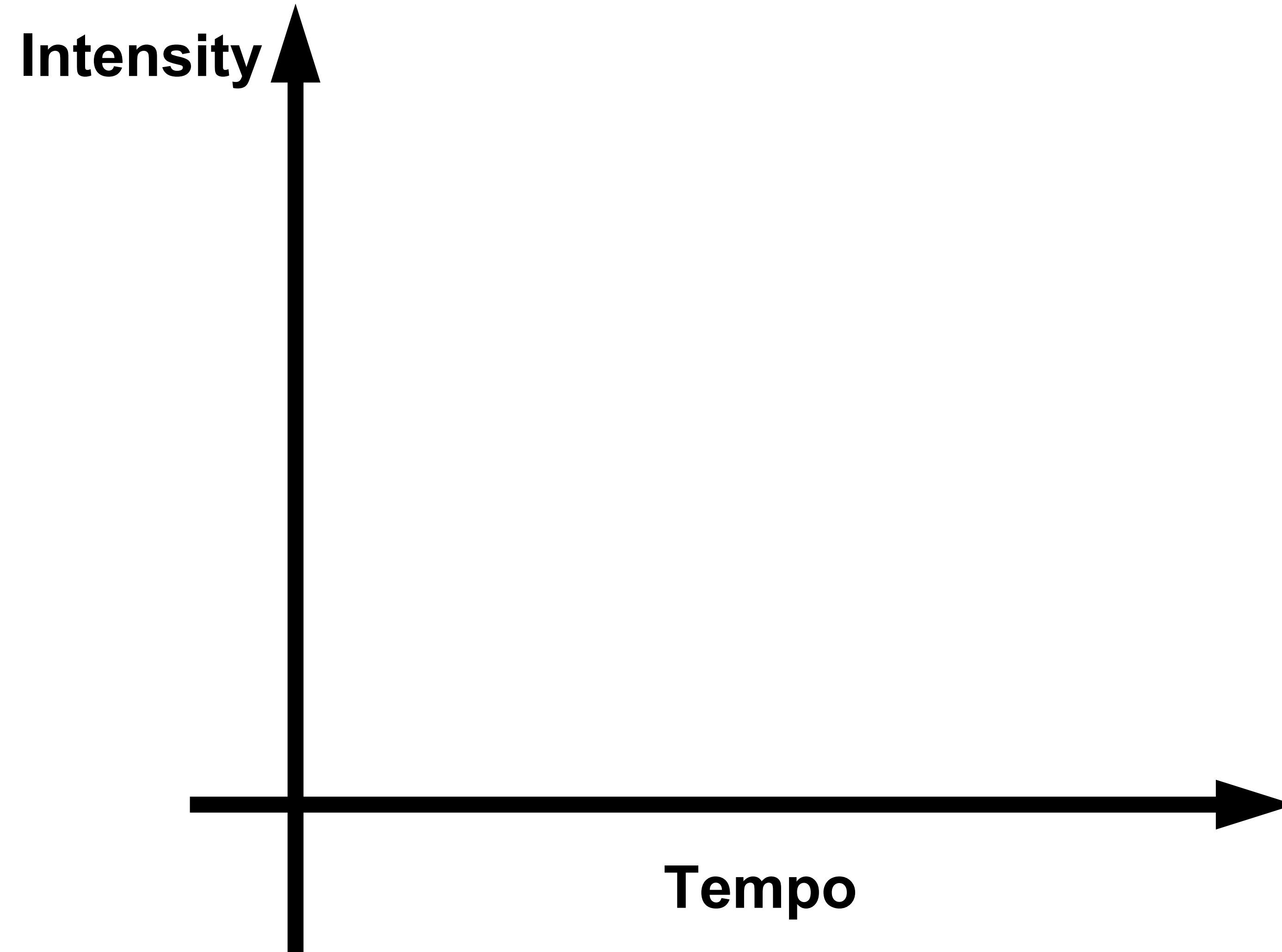
model



# Example 1: Predict whether a user likes a song or not



User Sharon





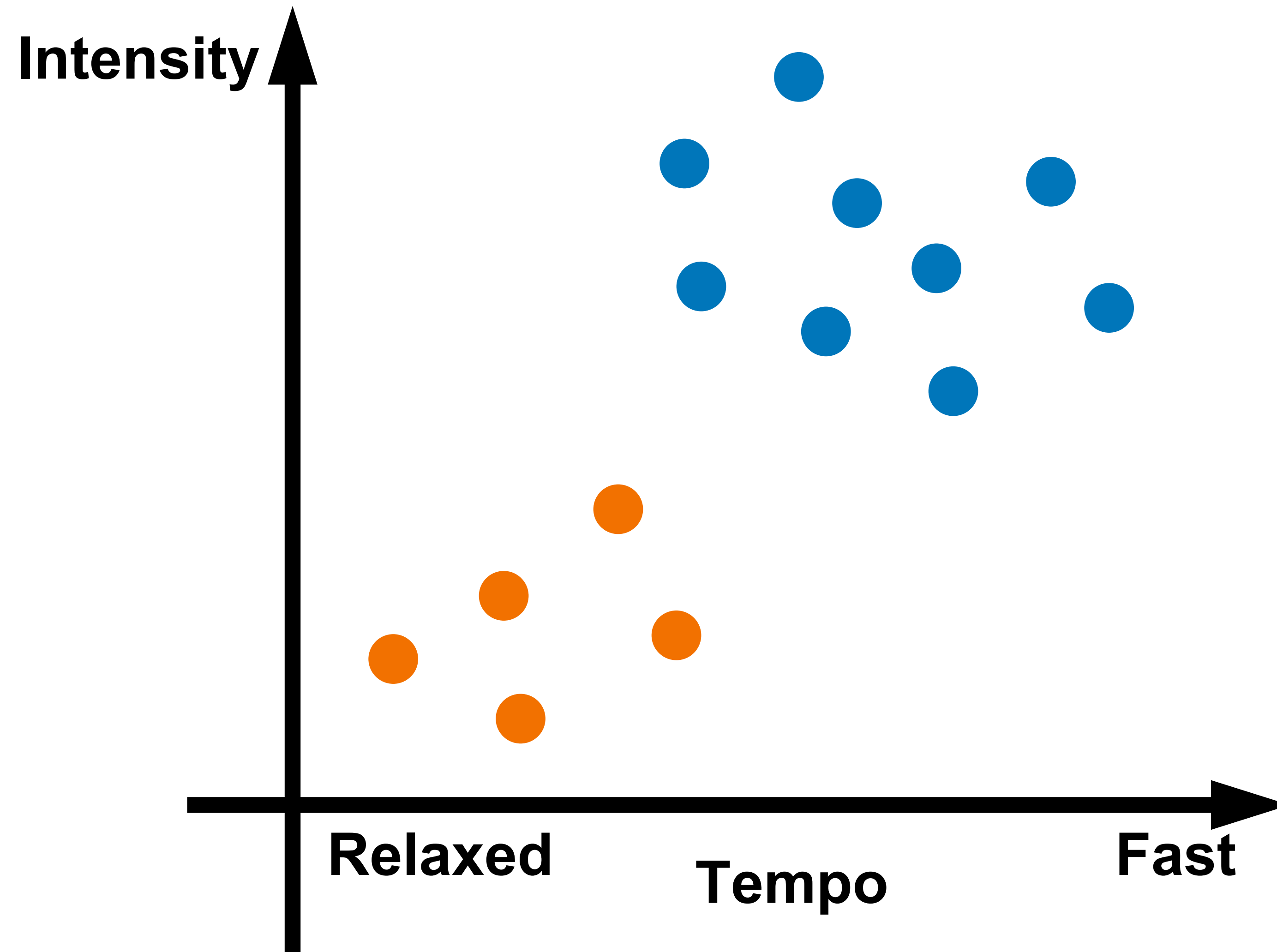
# Example 1: Predict whether a user likes a song or not



User Sharon

● Dislike

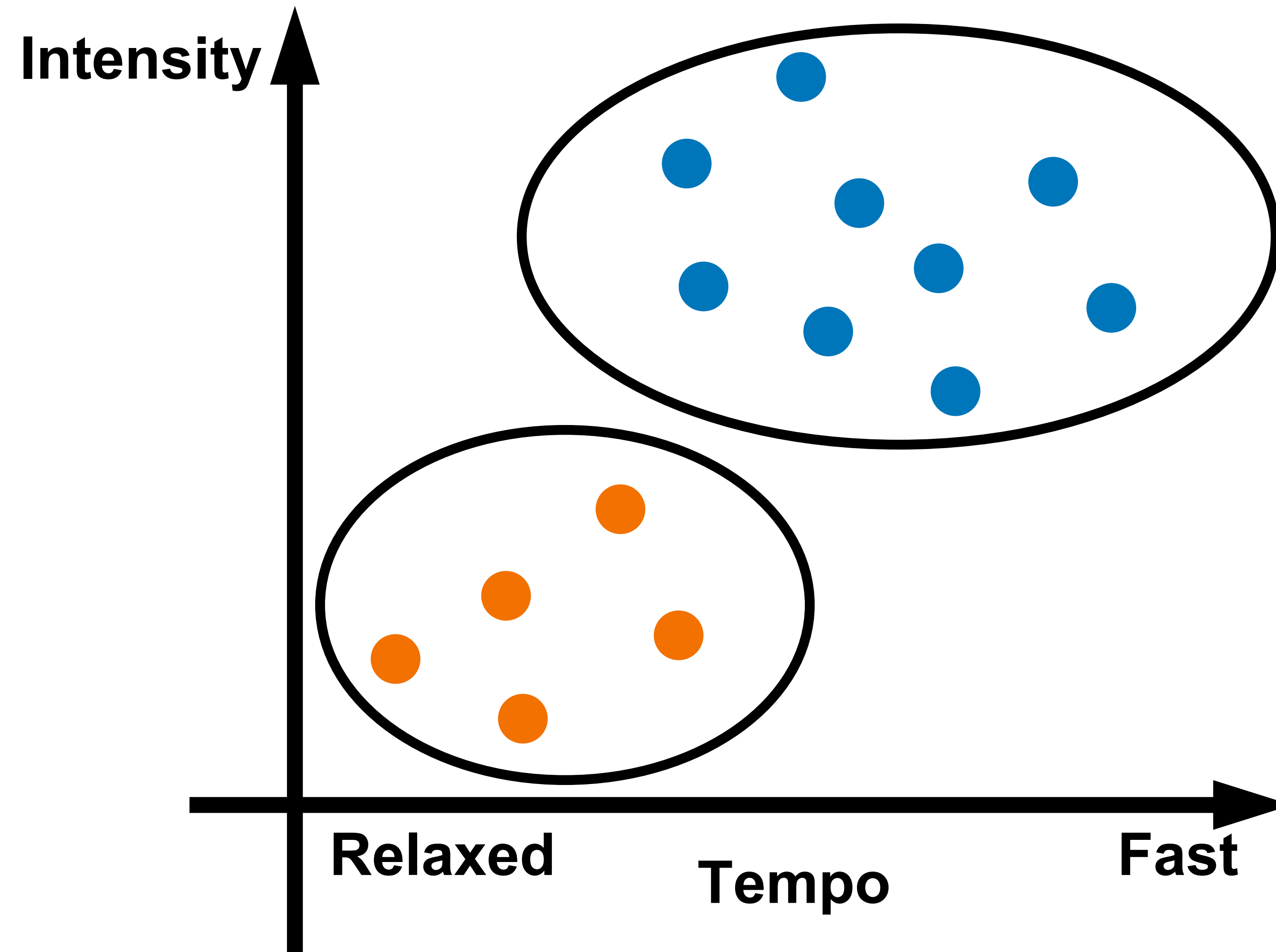
● Like



# Example 1: Predict whether a user likes a song or not



User Sharon





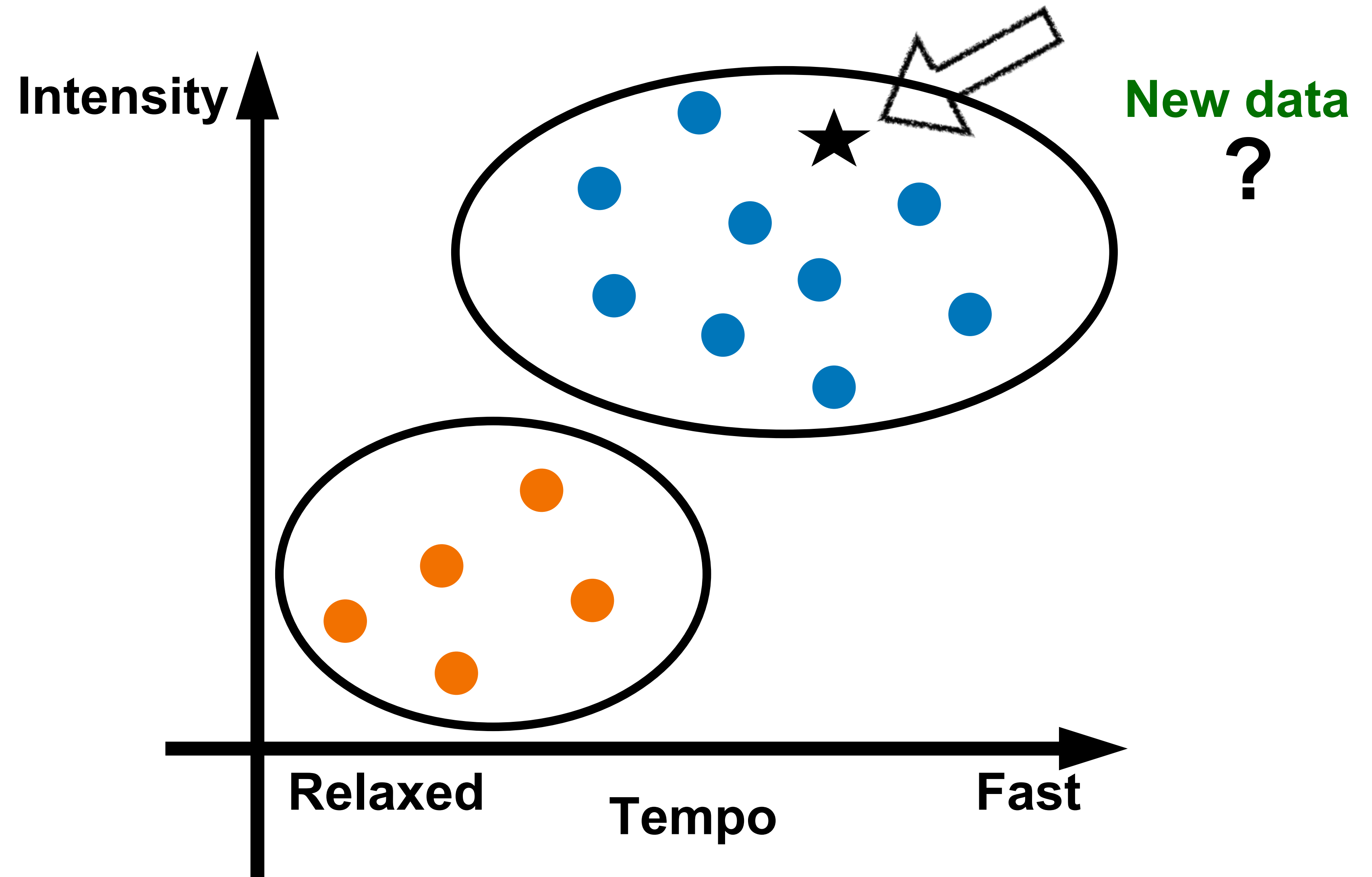
# Example 1: Predict whether a user likes a song or not



User Sharon

● Dislike

● Like



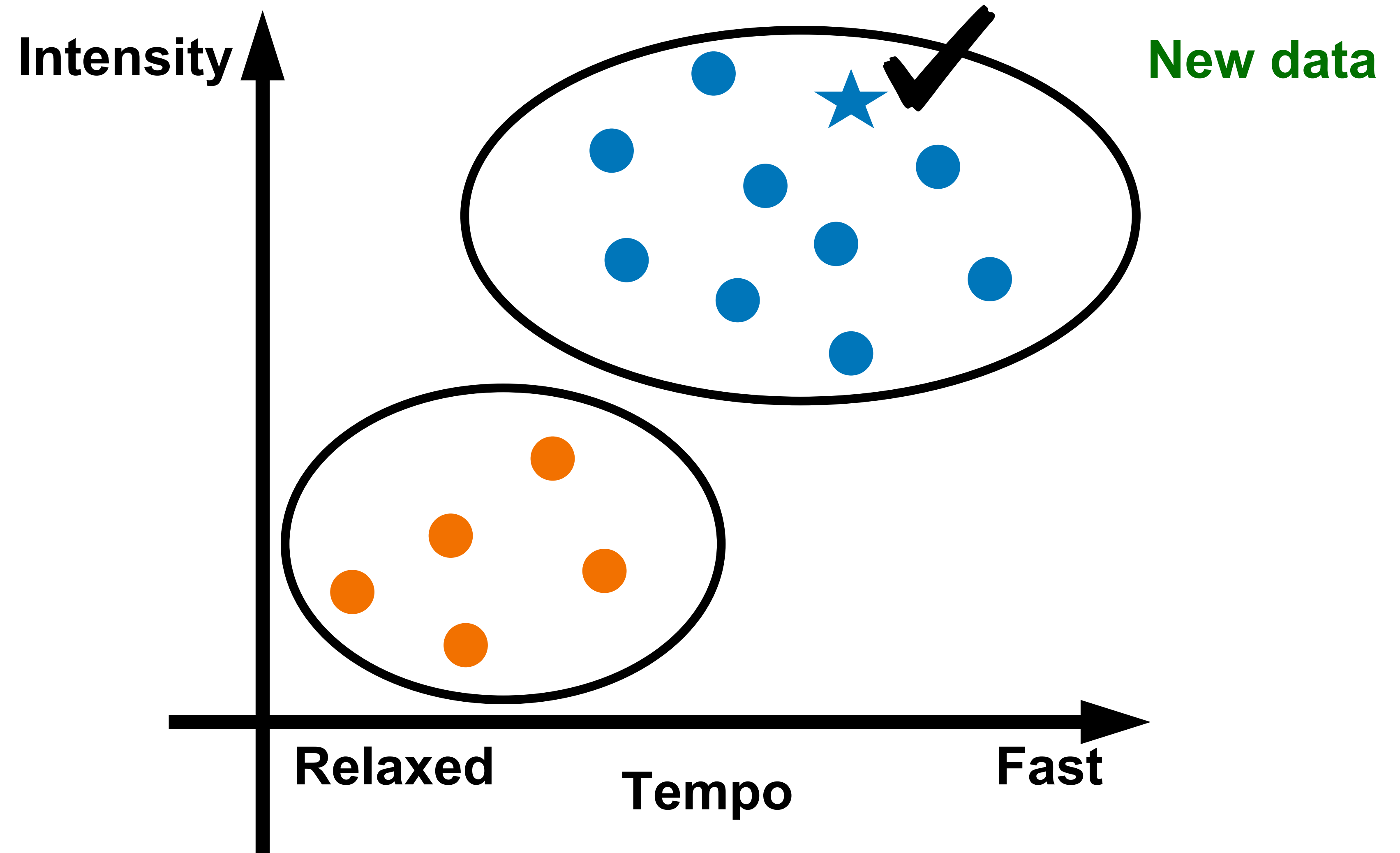
# Example 1: Predict whether a user likes a song or not



# User Sharon

 Dislike

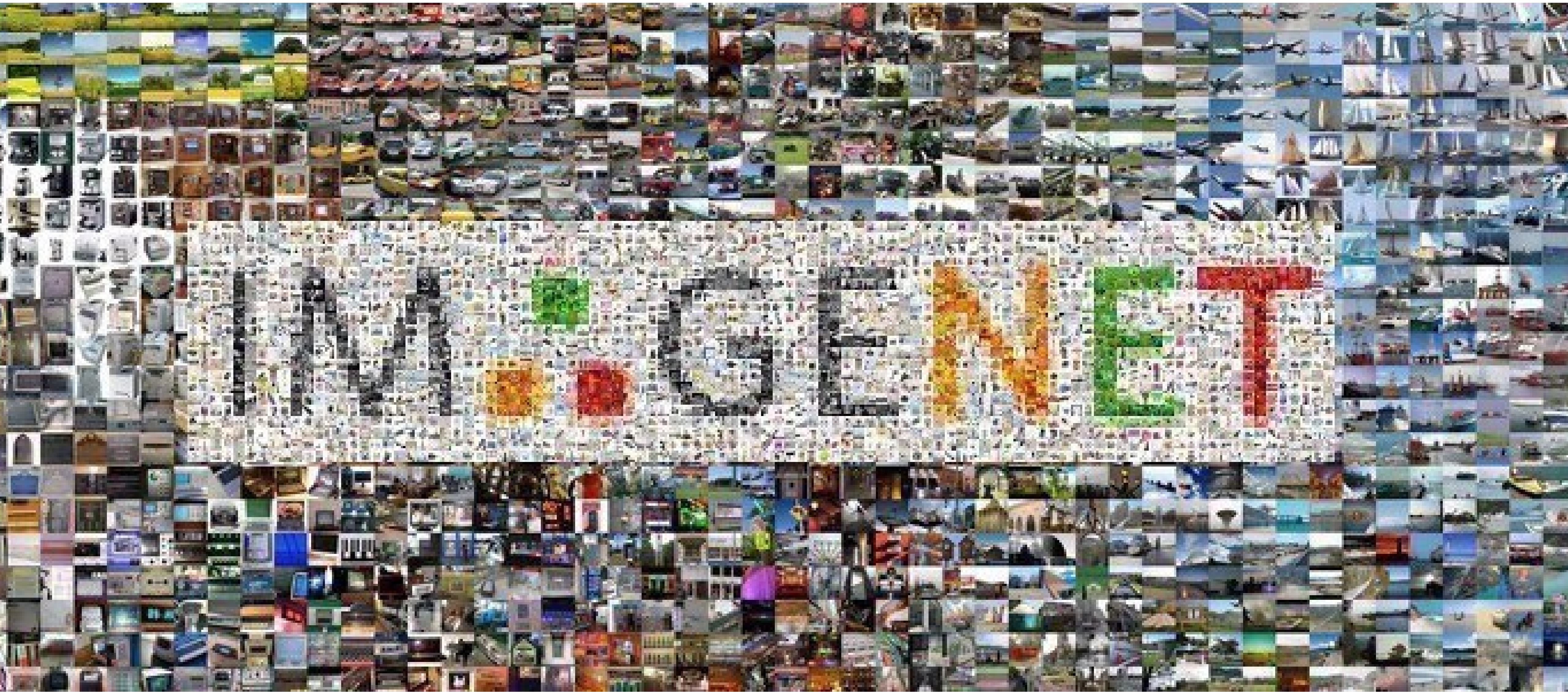
 Like





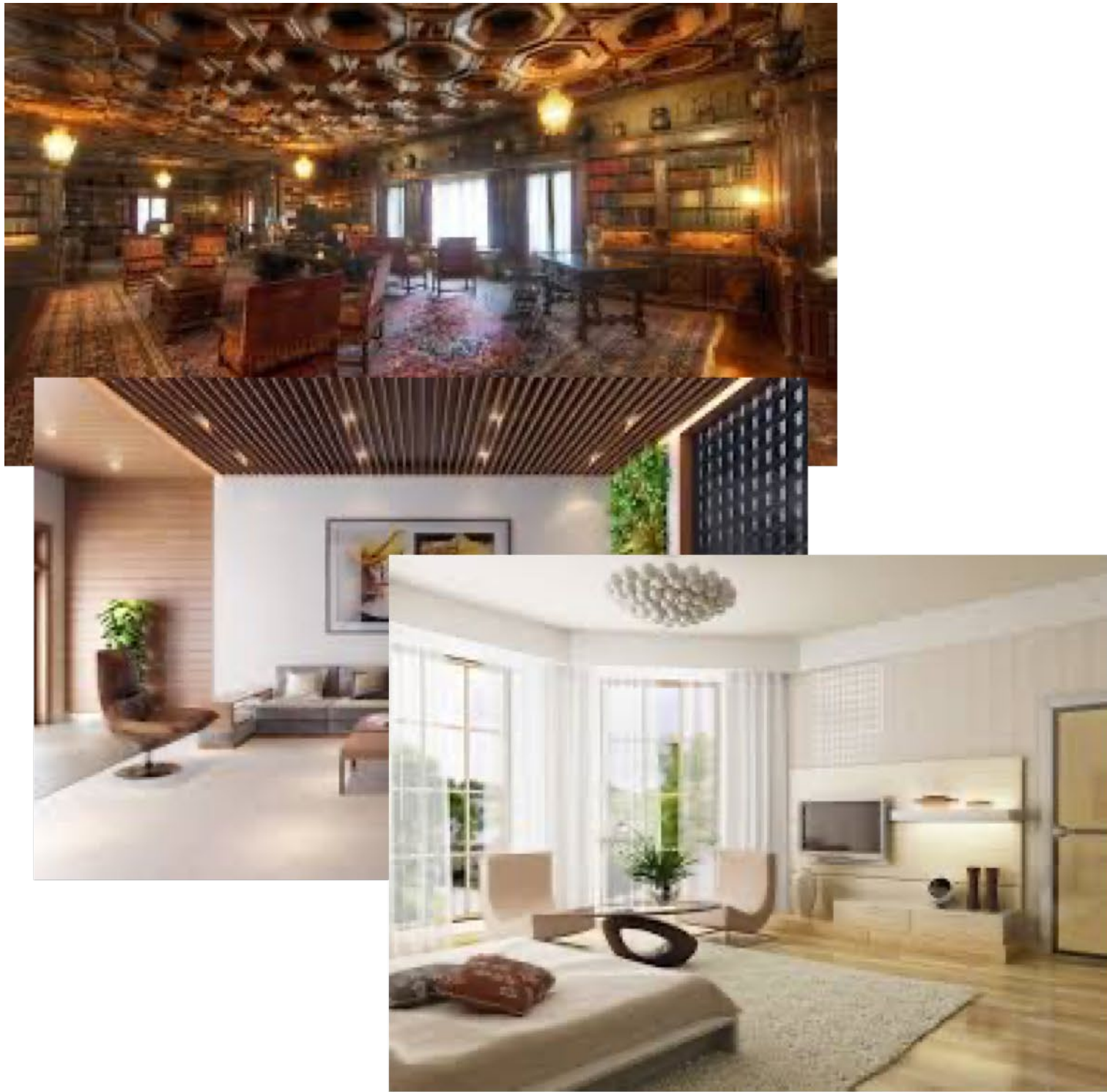
# Example 2: Classify Images

<http://www.image-net.org/>

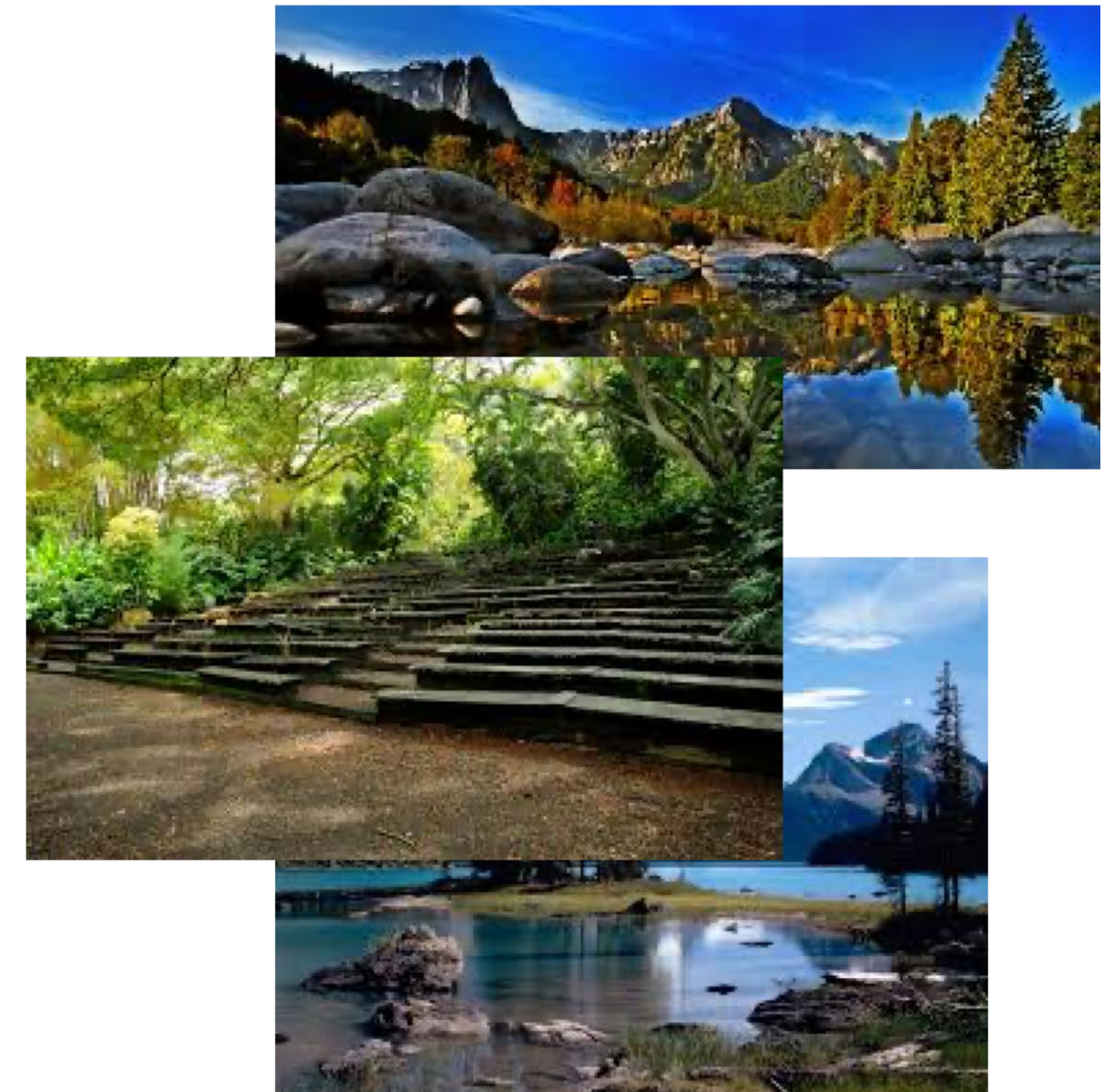




# Example 2: Classify Images



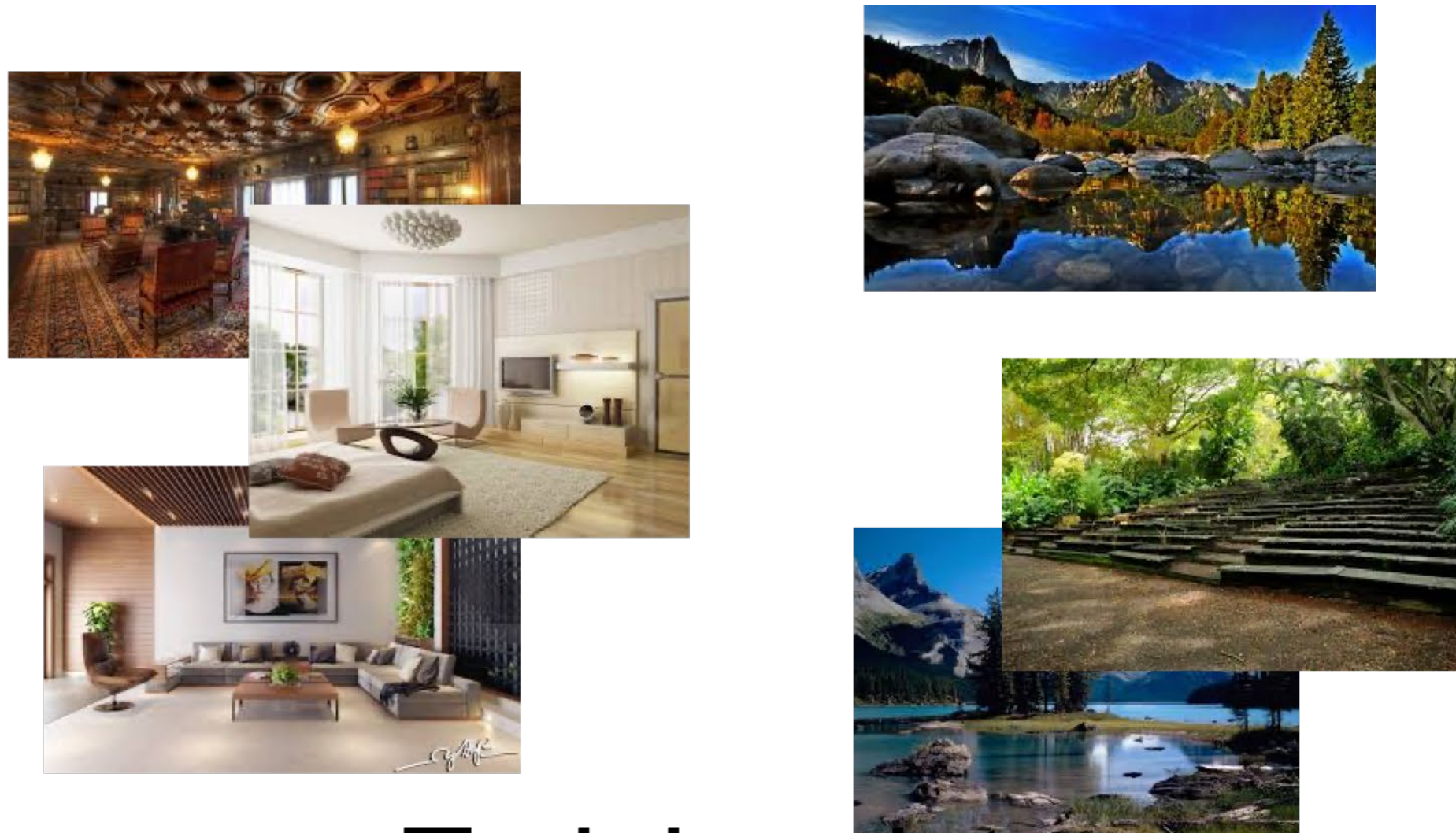
indoor



outdoor



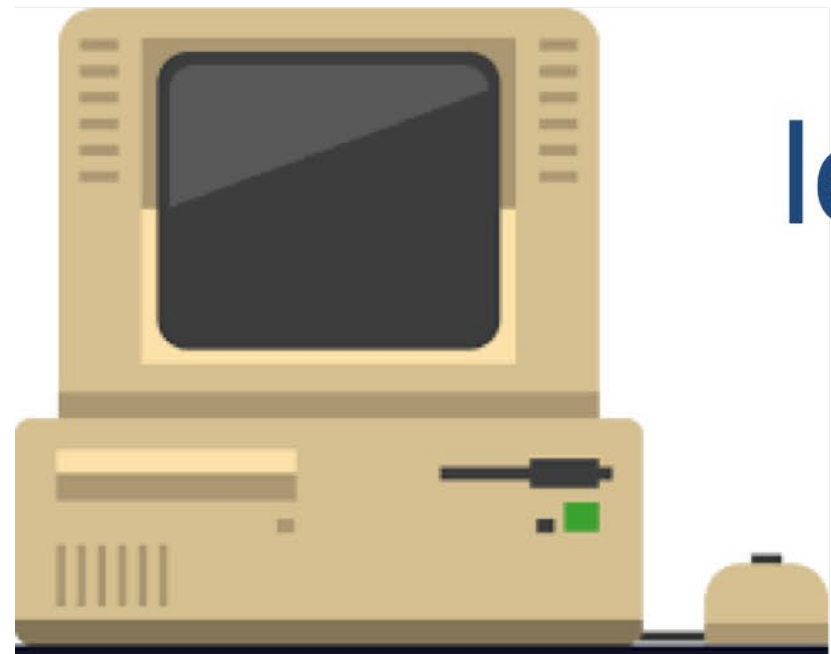
# Example 2: Classify Images



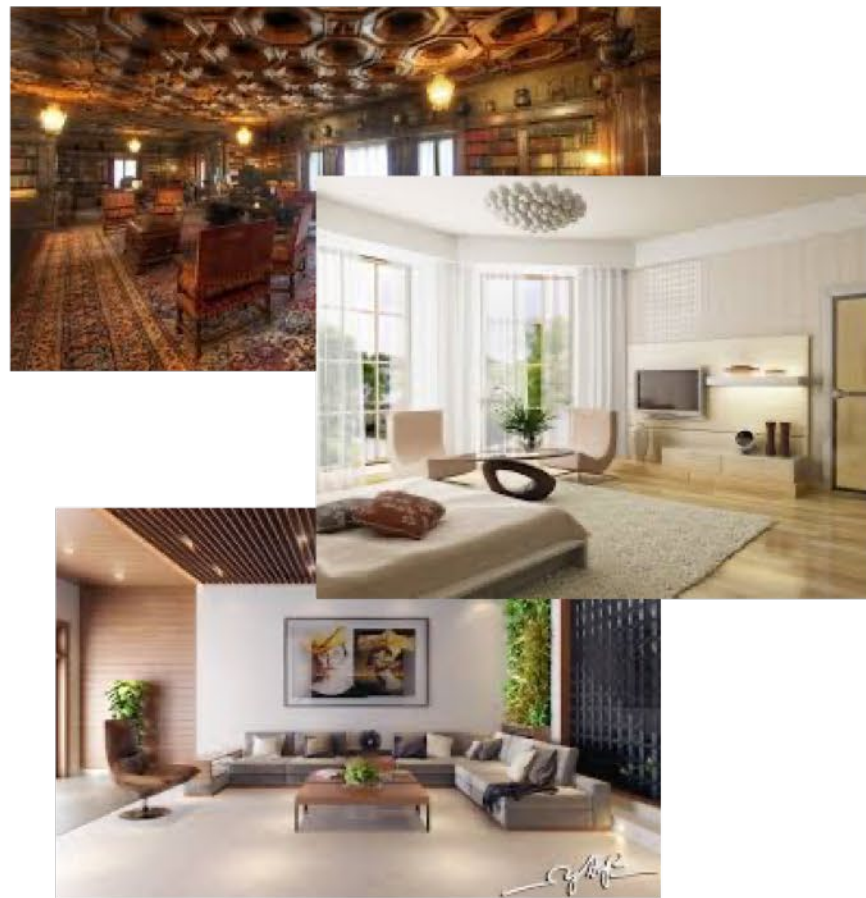
Training data



learning (i.e., training)







Training data

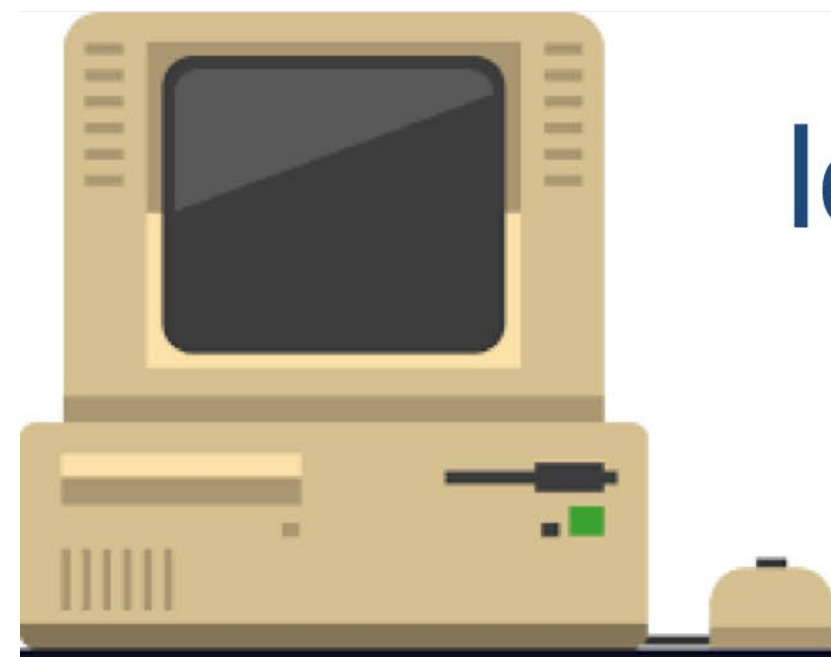


Label: outdoor



Label: indoor

Test data



learning (i.e., training)



testing



performance



# How to represent data?

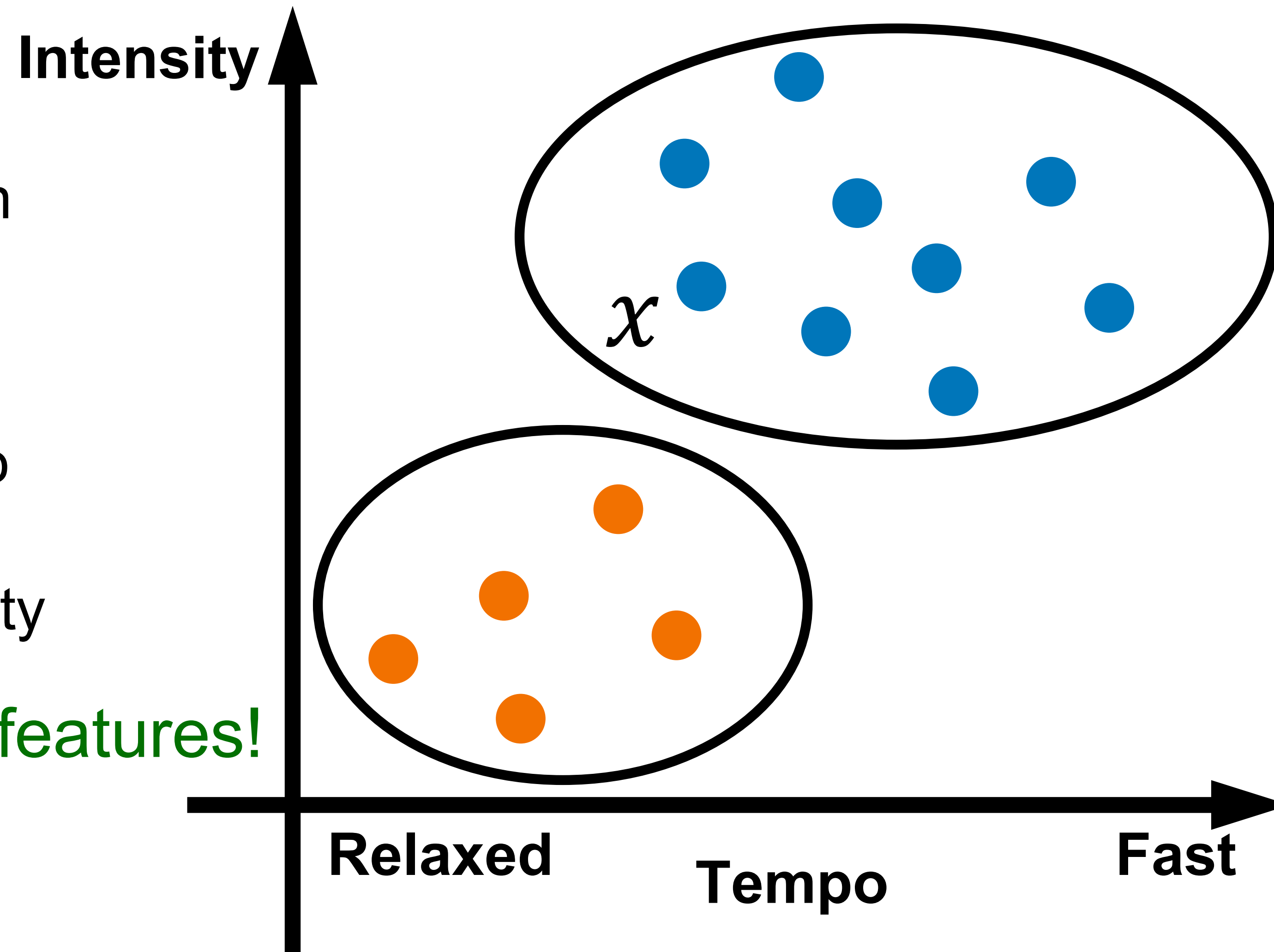
input data

$$x \in \mathbb{R}^d$$

$d$ : feature dimension

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \begin{matrix} \text{Tempo} \\ \text{Intensity} \end{matrix}$$

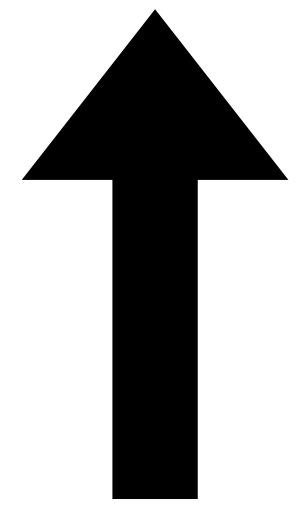
There can be many features!



# How to represent data?

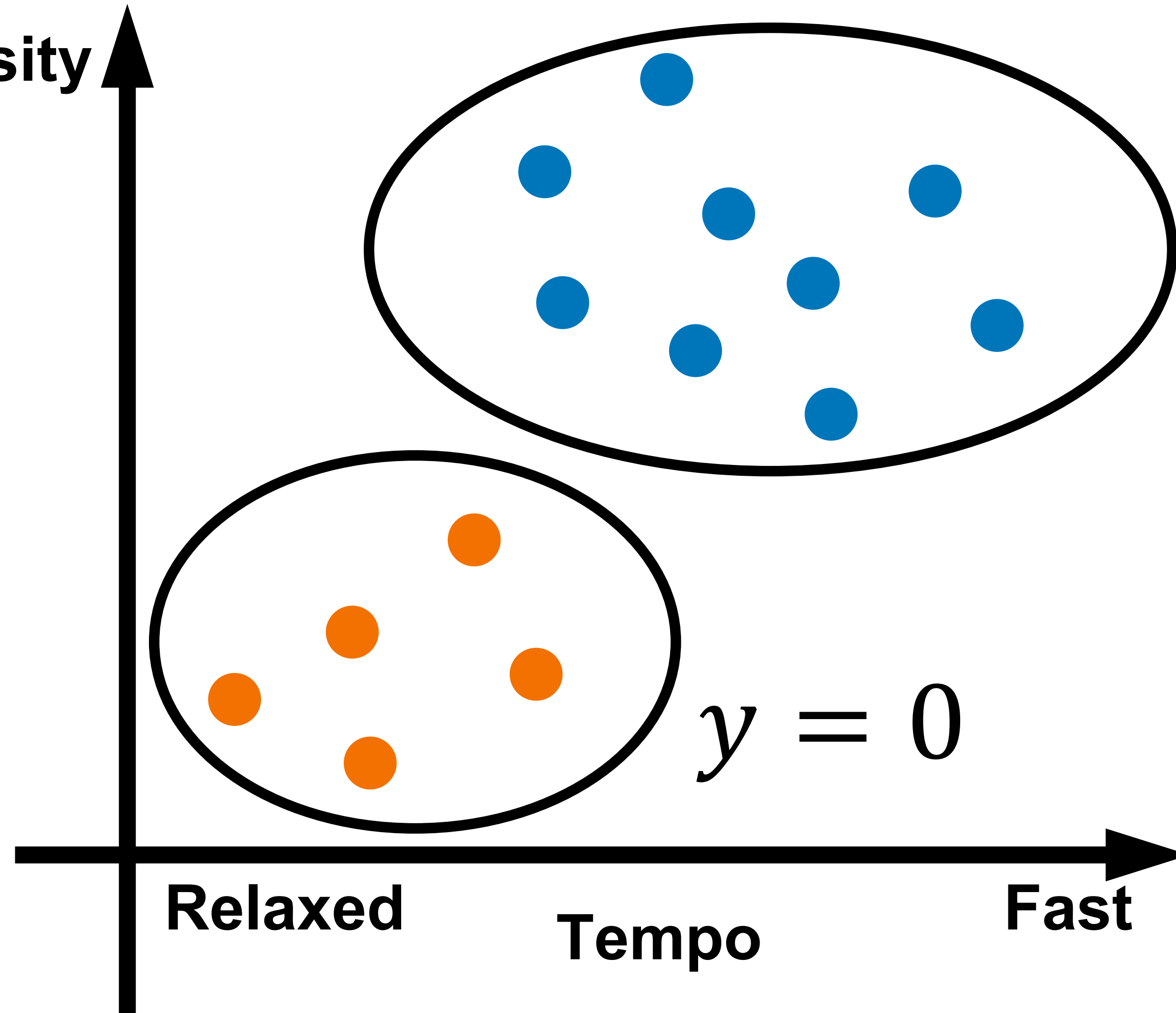
Label

$y \in \{0,1\}$



Where "supervision"  
comes from

Intensity



$y = 1$

$y = 0$

Relaxed

Tempo

Fast

# Represent various types of data

- Image
  - Pixel values
- Bank account
  - Credit rating, balance, # deposits in last day, week, month, year, #withdrawals



# Two Types of Supervised Learning Algorithms

**Classification**

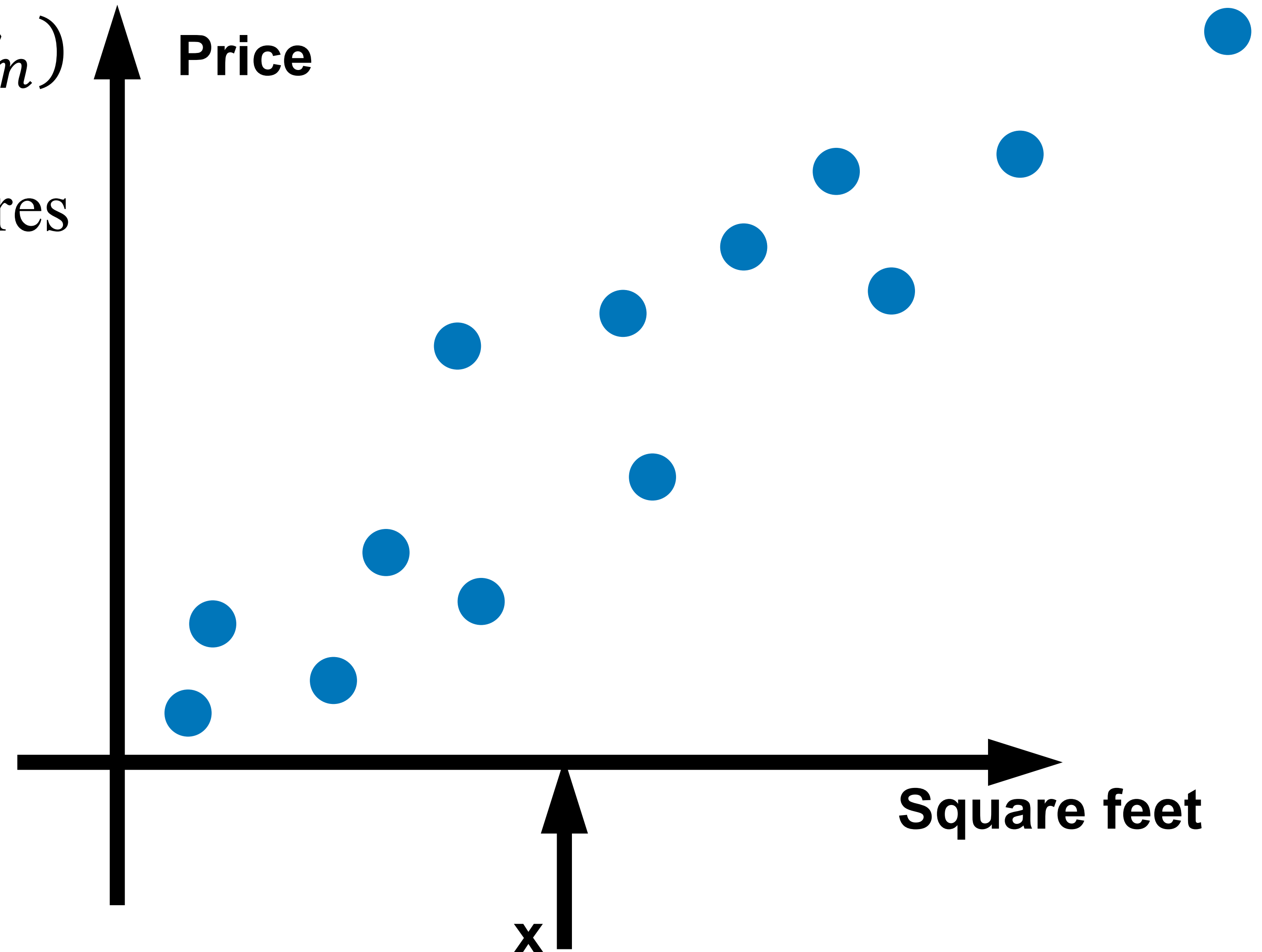
**Regression**

# Example of regression: housing price prediction

Given: a dataset that contains  $n$  samples

$(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

**Task:** if a residence has  $x$  squares feet, predict the price?





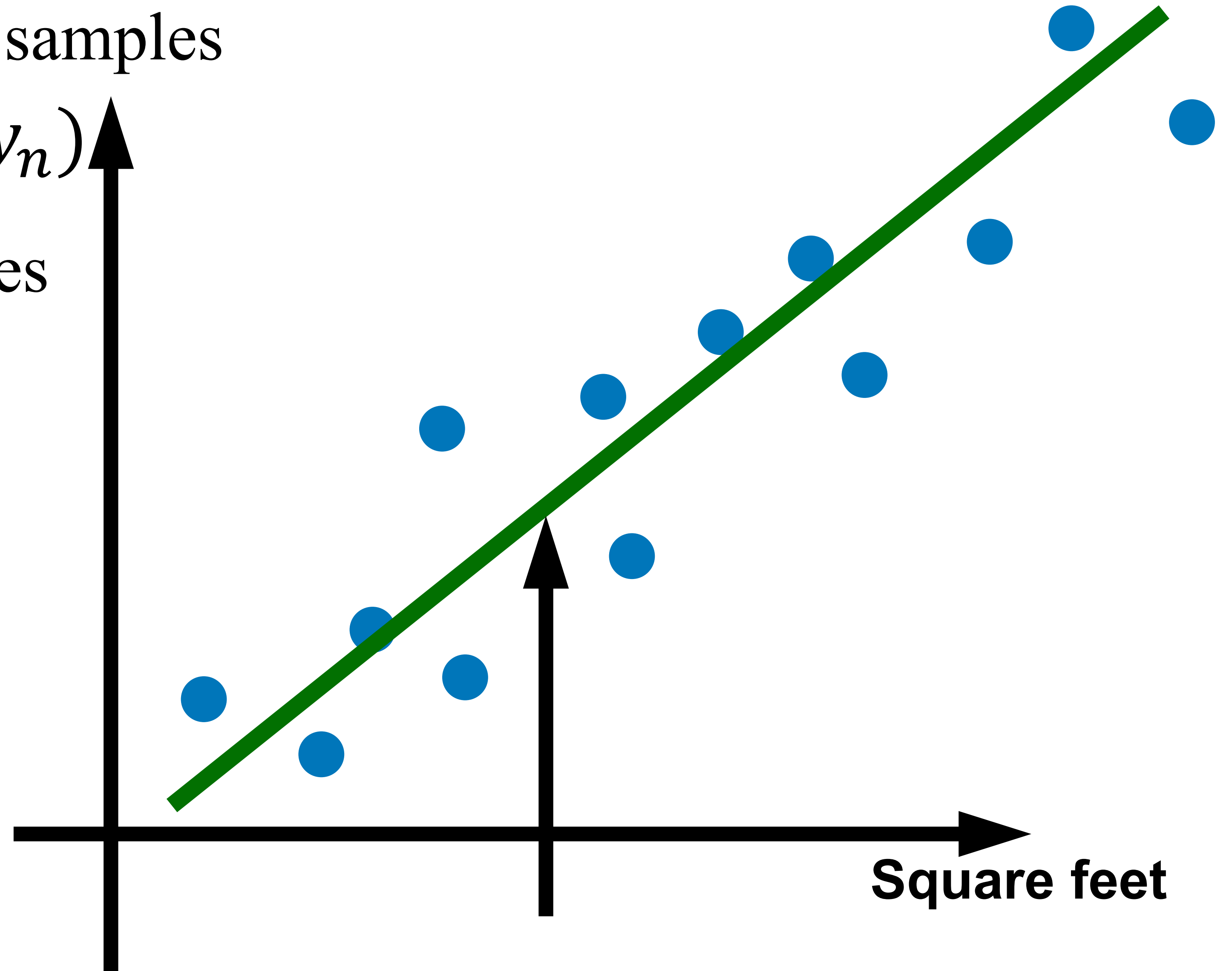
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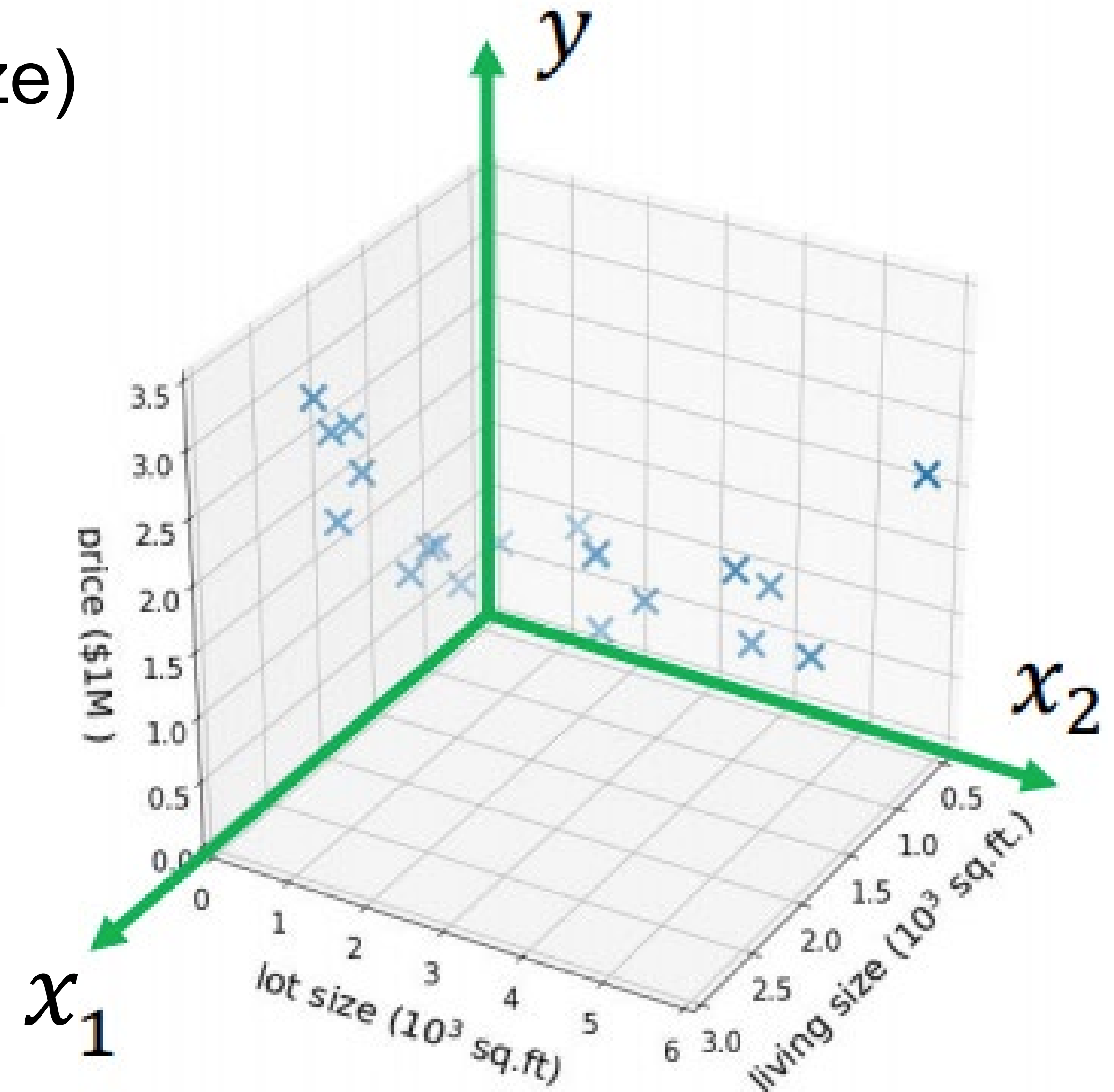
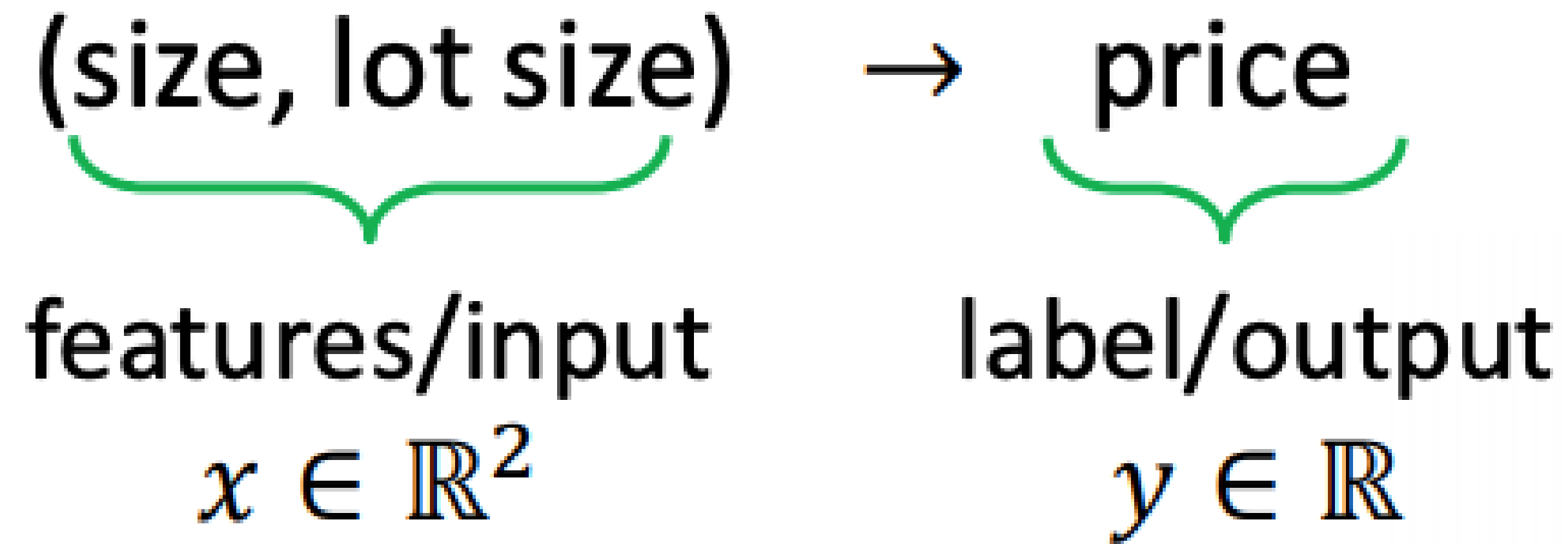
**Task:** if a residence has  $x$  squares feet, predict the price?

$$y \in \mathbb{R}$$



# Example of regression: housing price prediction

Input with more features (e.g., lot size)



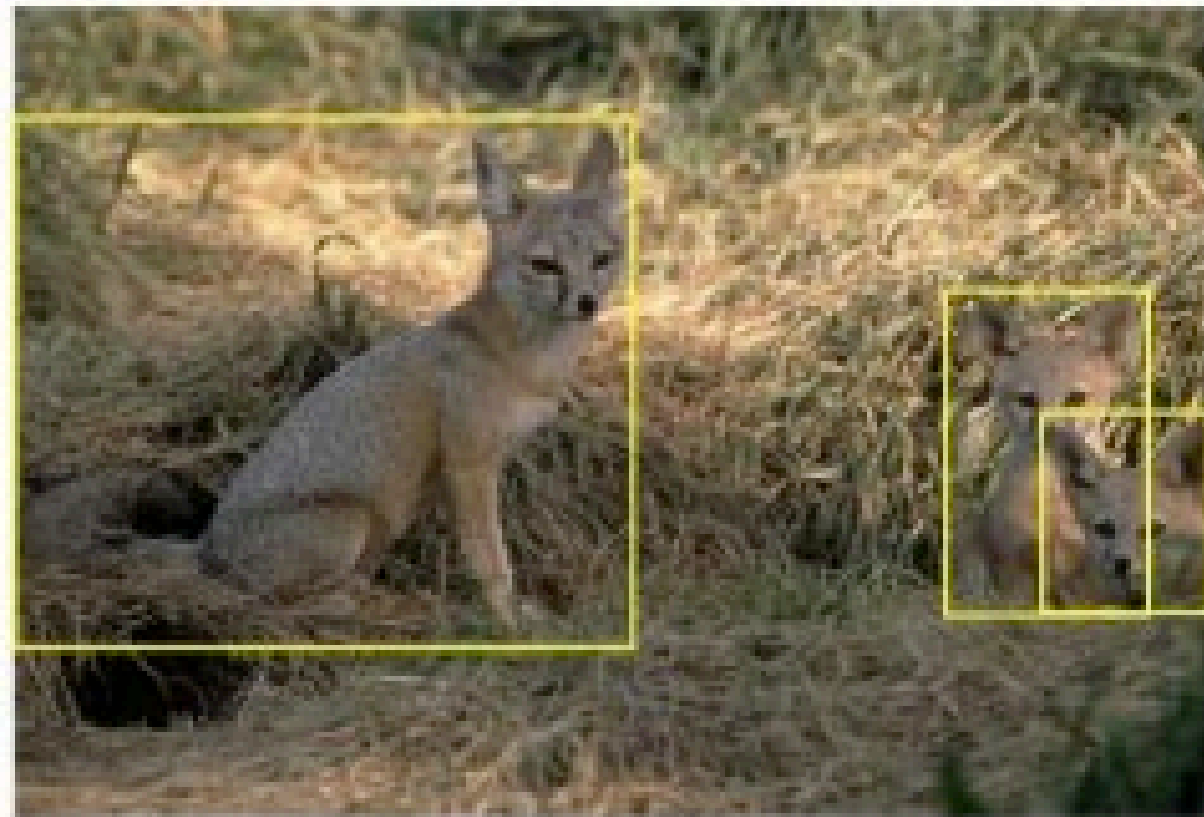
(credit: stanford CS229)



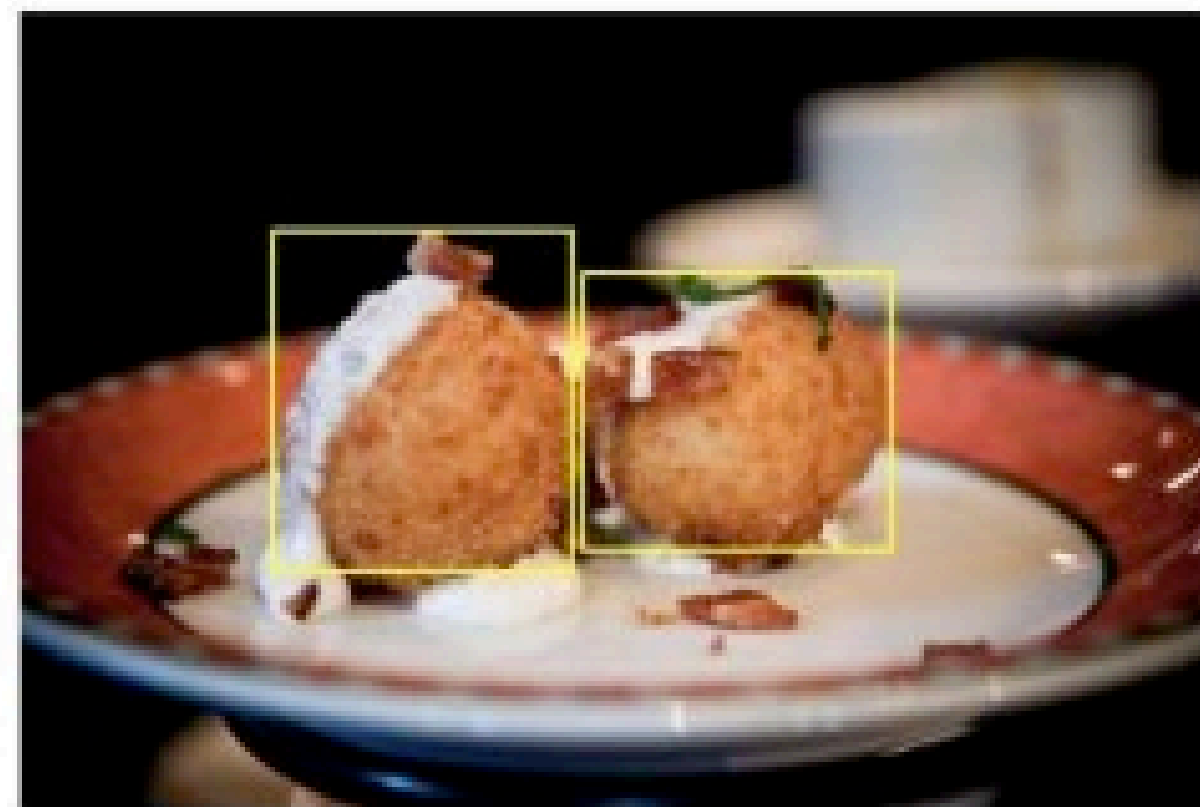
# Supervised Learning: More examples

$x$  = raw pixels of the image

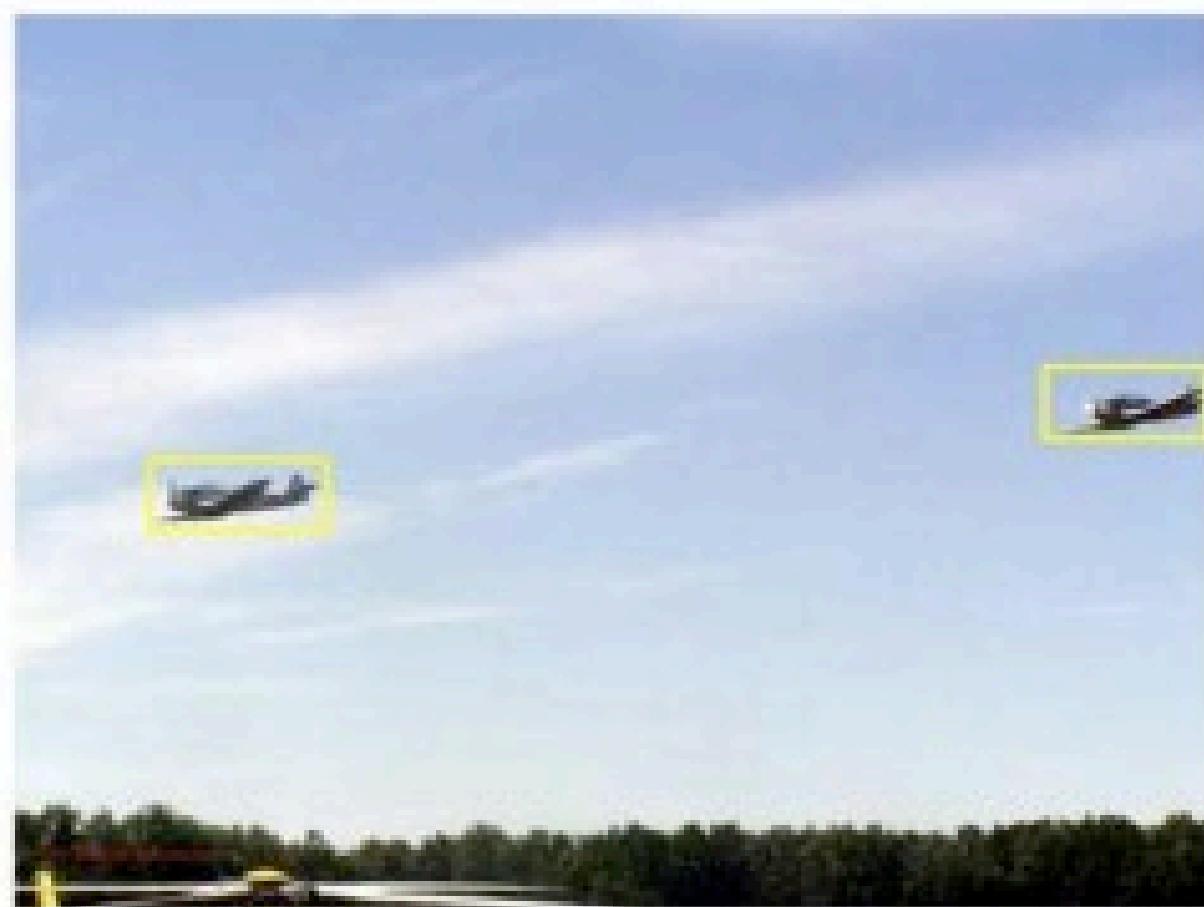
$y$  = bounding boxes



kit fox



croquette



airplane



frog

# Two Types of Supervised Learning Algorithms

## Classification

- the label is a **discrete** variable

$$y \in \{1, 2, 3, \dots, K\}$$

## Regression

- the label is a **continuous** variable

$$y \in \mathbb{R}$$

# Training Data for Supervised Learning

Training data is a collection of input instances to the learning algorithm:

$$(\text{input}, \text{label}), (x_2, y_2), (x_3, y_3), \dots, (x_n, y_n)$$

The training data is the “**experience**” given to a learning algorithm



# Goal of Supervised Learning

Given training data

$$(x_1, y_1), (x_2, y_2), (x_3, y_3), \dots, (x_n, y_n)$$

Learn a function mapping  $f: X \rightarrow Y$ , such that  $f(x)$  predicts the label  $y$  on **future** data  $x$  (not in training data)

# Goal of Supervised Learning

Training set error

- 0-1 loss for classification  $\ell = \frac{1}{n} \sum_{i=1}^n (f(\mathbf{x}_i) \neq y_i)$
- Squared loss for regression:  $\ell = \frac{1}{n} \sum_{i=1}^n (f(\mathbf{x}_i) - y_i)^2$

A learning algorithm optimizes the training objective

$$f^* = \operatorname{argmin} \mathbb{E}_{(x,y)} \ell(f(x), y)$$

Details in upcoming lectures :)

# Quiz Break

Q1-1: Which is true about feature vectors?

- A. Feature vectors can have at most 10 dimensions
- B. Feature vectors have only numeric values
- C. The raw image can also be used as the feature vector
- D. Text data don't have feature vectors



# Quiz Break

Q1-1: Which is true about feature vectors?

- A. Feature vectors can have at most 10 dimensions
- B. Feature vectors have only numeric values
- C. The raw image can also be used as the feature vector
- D. Text data don't have feature vectors

- A. Feature vectors can be high dimensional
- B. Some feature vectors can have other types of values like strings
- D. Bag-of-words is a type of feature vector for text

# Quiz Break

Q1-2: Which of the following is not a common task of supervised learning?

- A. Object detection (predicting bounding box from raw images)
- B. Classification
- C. Regression
- D. Dimensionality reduction



# Quiz Break

Q1-2: Which of the following is not a common task of supervised learning?

- A. Object detection (predicting bounding box from raw images)
- B. Classification
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- D. Dimensionality reduction**





## Part II: Unsupervised Learning (no teacher)

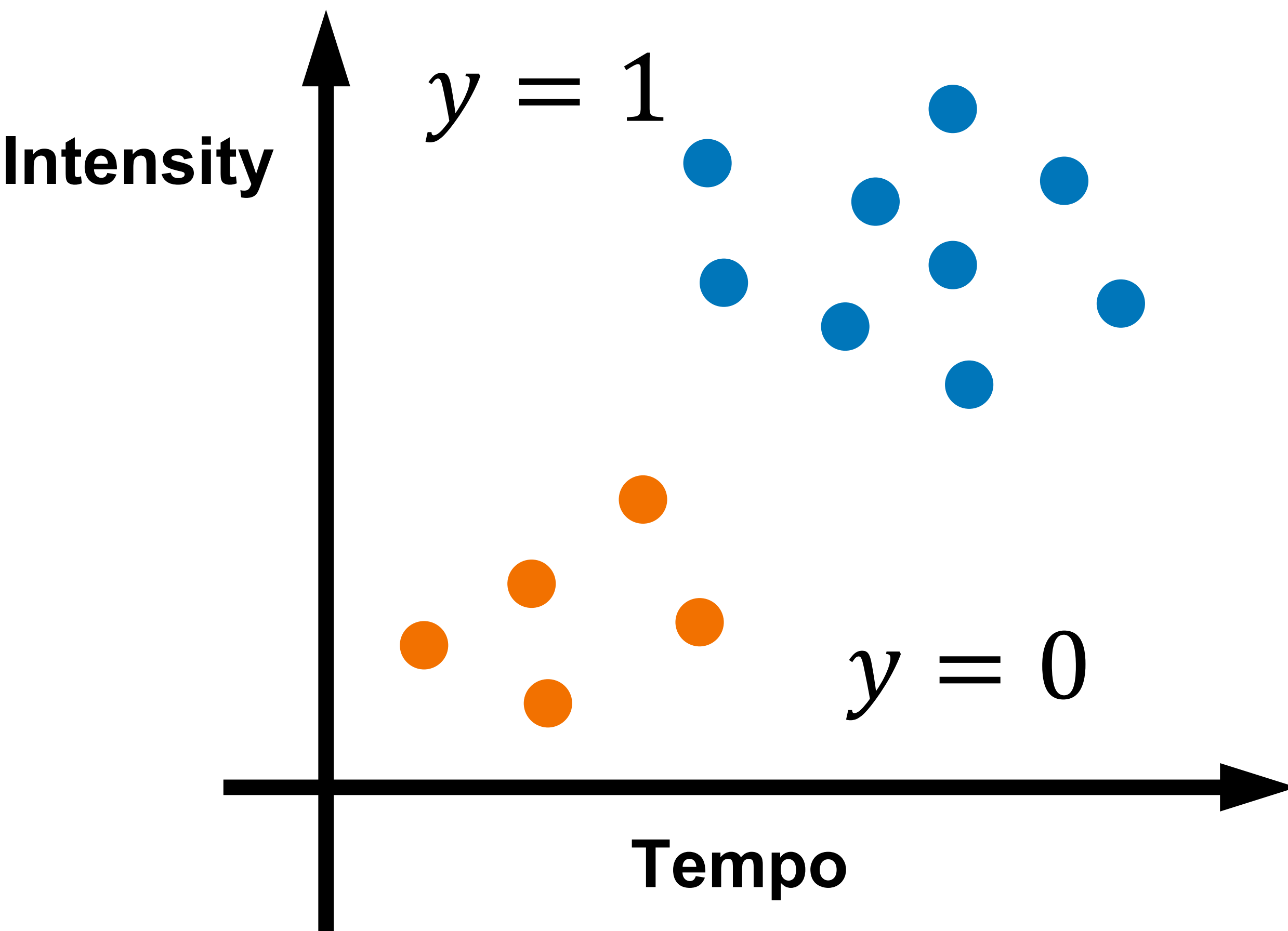


# Unsupervised Learning

- Given: dataset contains **no label**  $x_1, x_2, \dots, x_n$
- **Goal:** discover interesting patterns and structures in the data

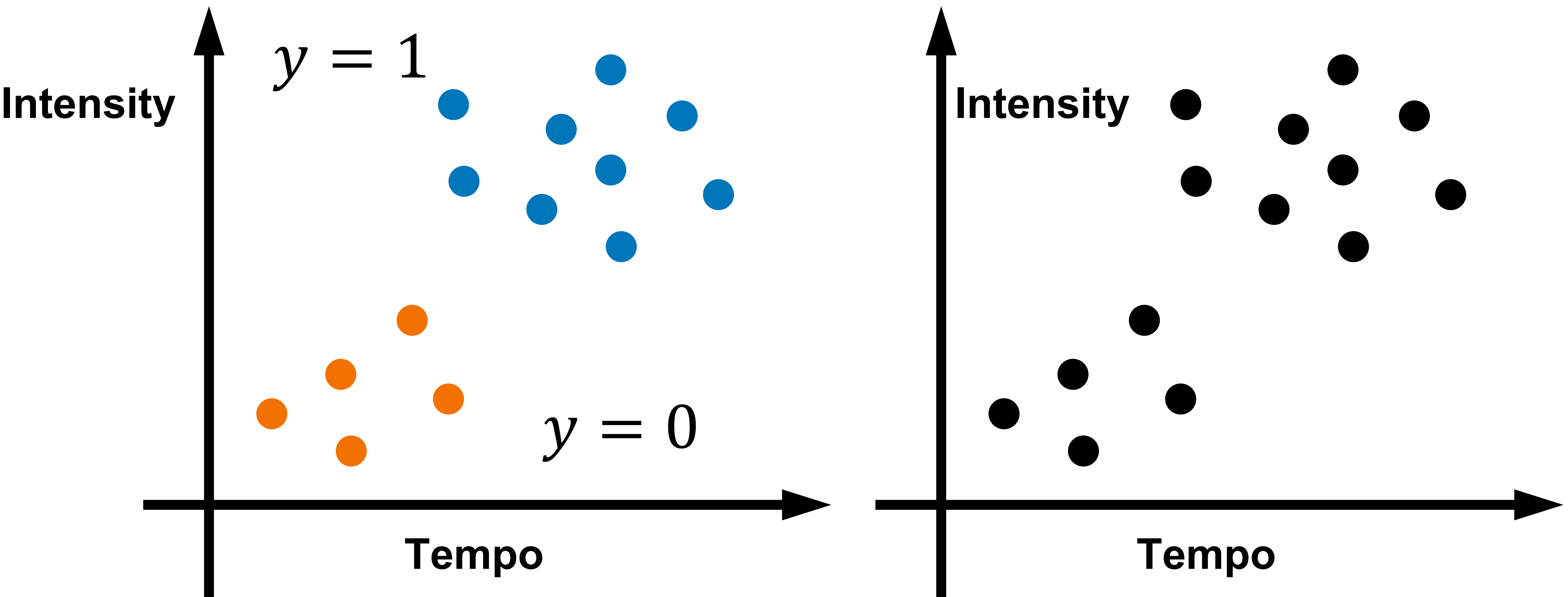
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# Unsupervised Learning

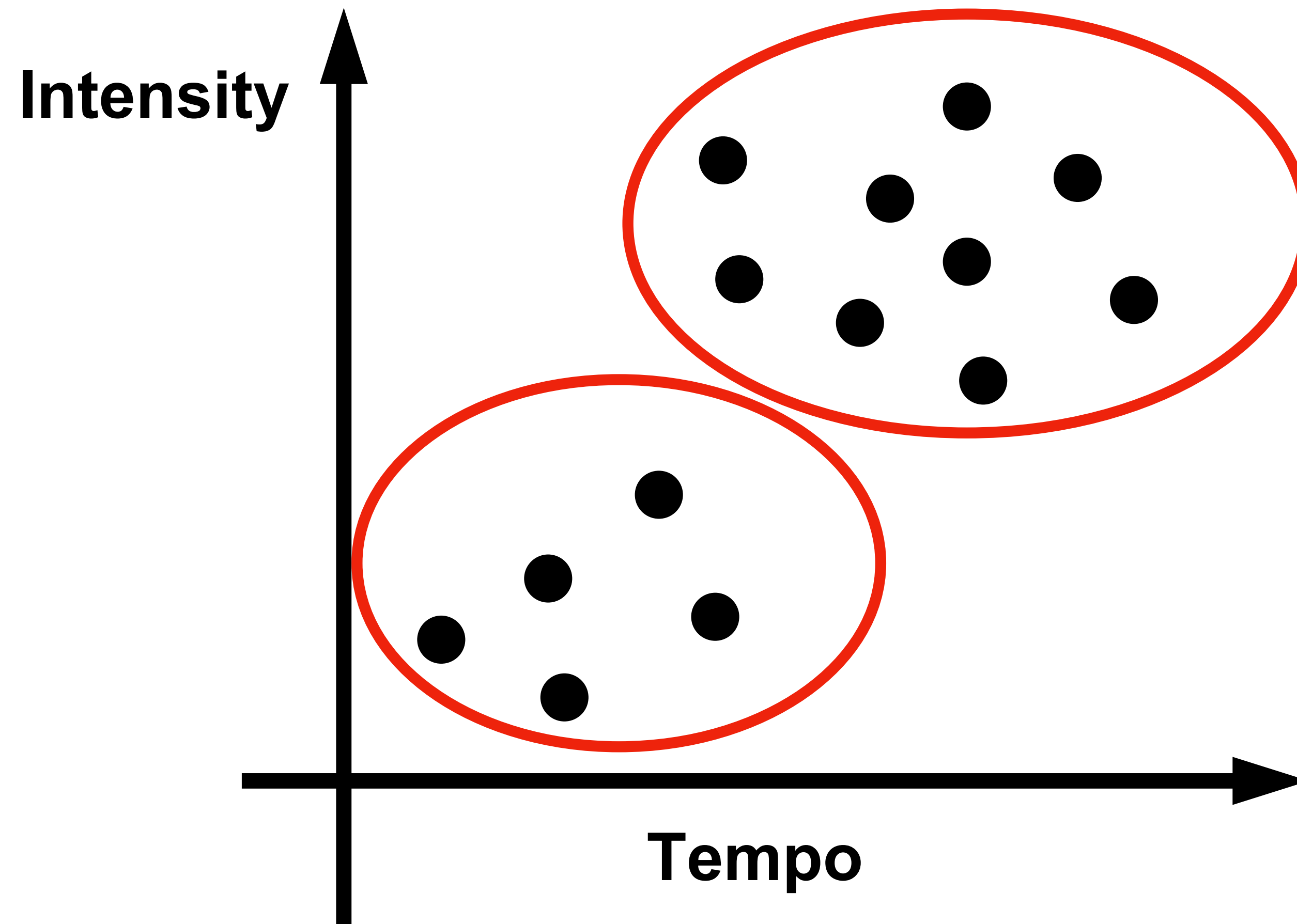
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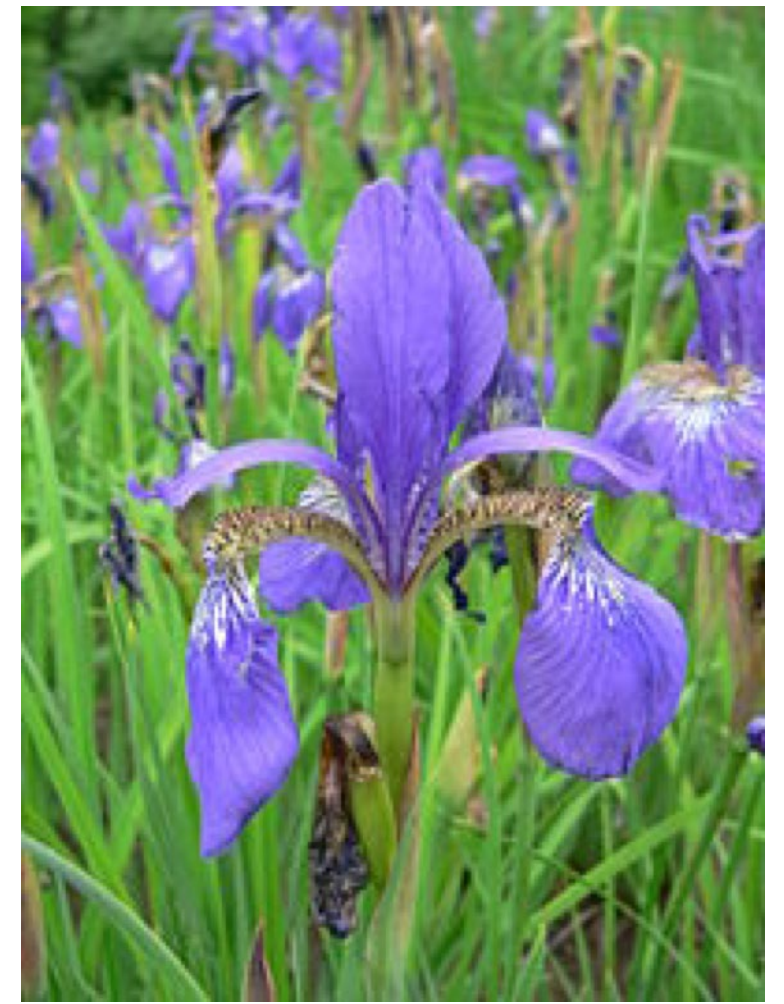
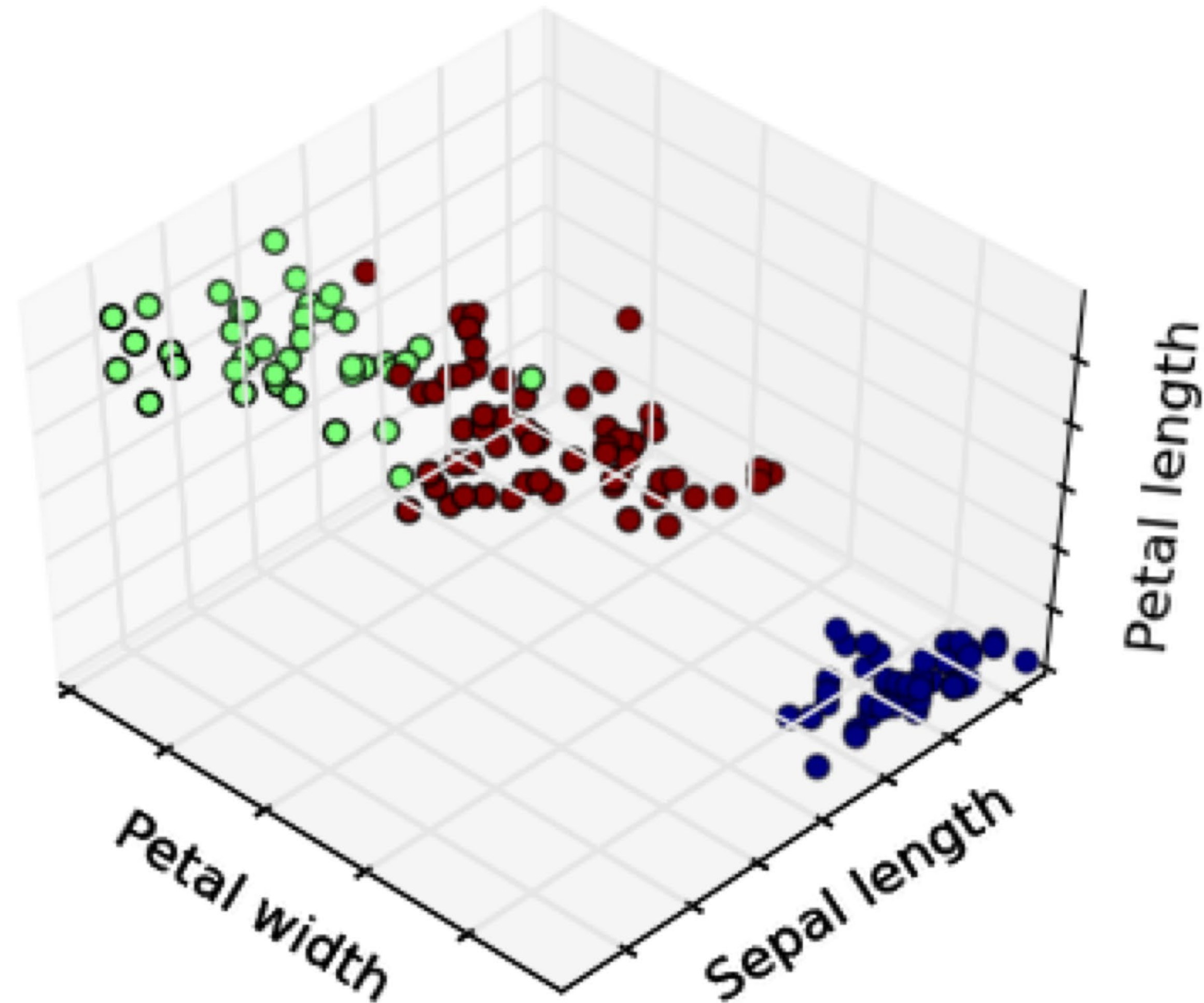


# Clustering

- Given: dataset contains **no label**  $x_1, x_2, \dots, x_n$
- **Output:** divides the data into clusters such that there are intra-cluster similarity and inter-cluster dissimilarity



# Clustering



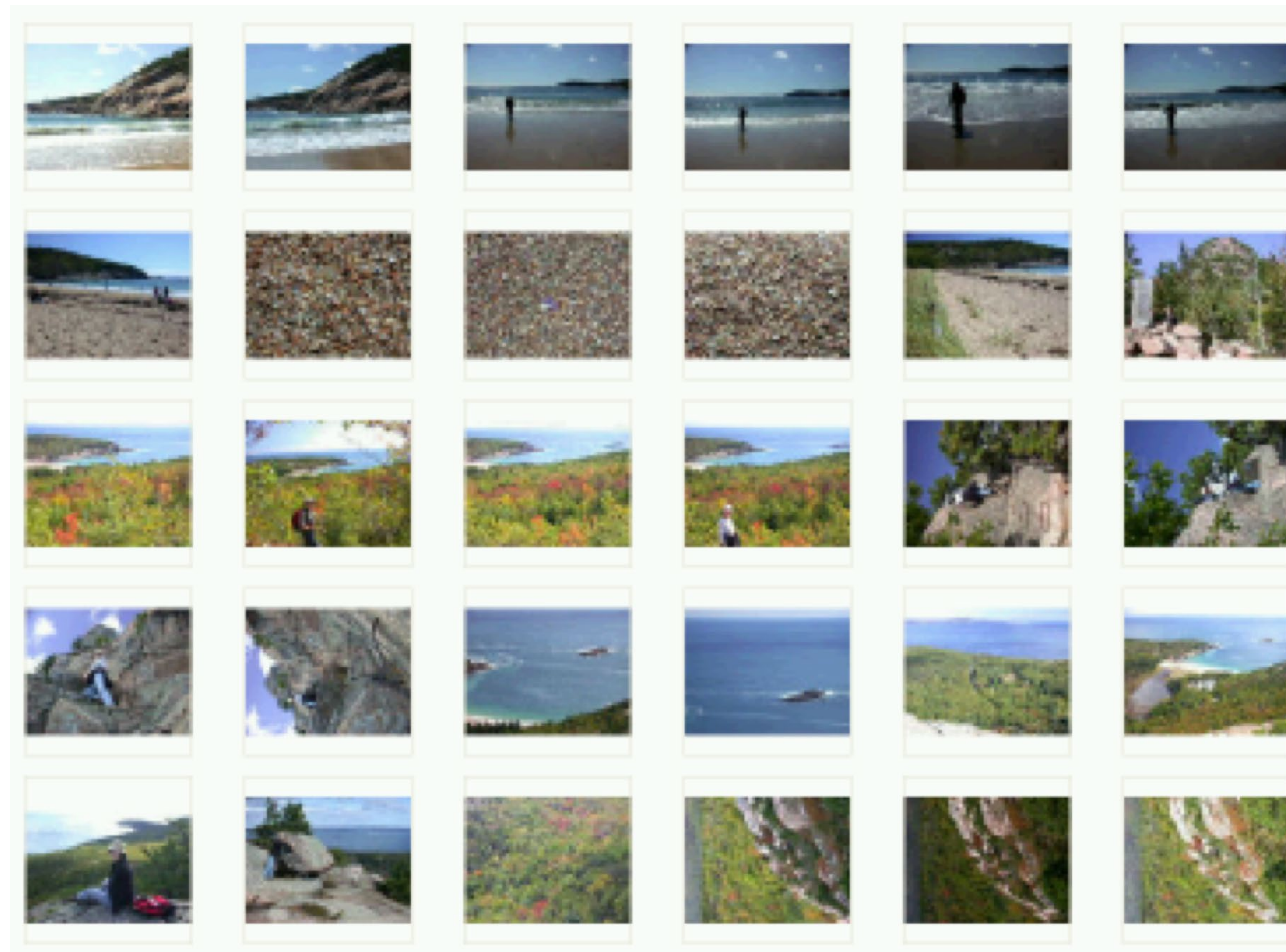
Clustering Irises using three different features

The colors represent clusters identified by the algorithm, **not** y's provided as input



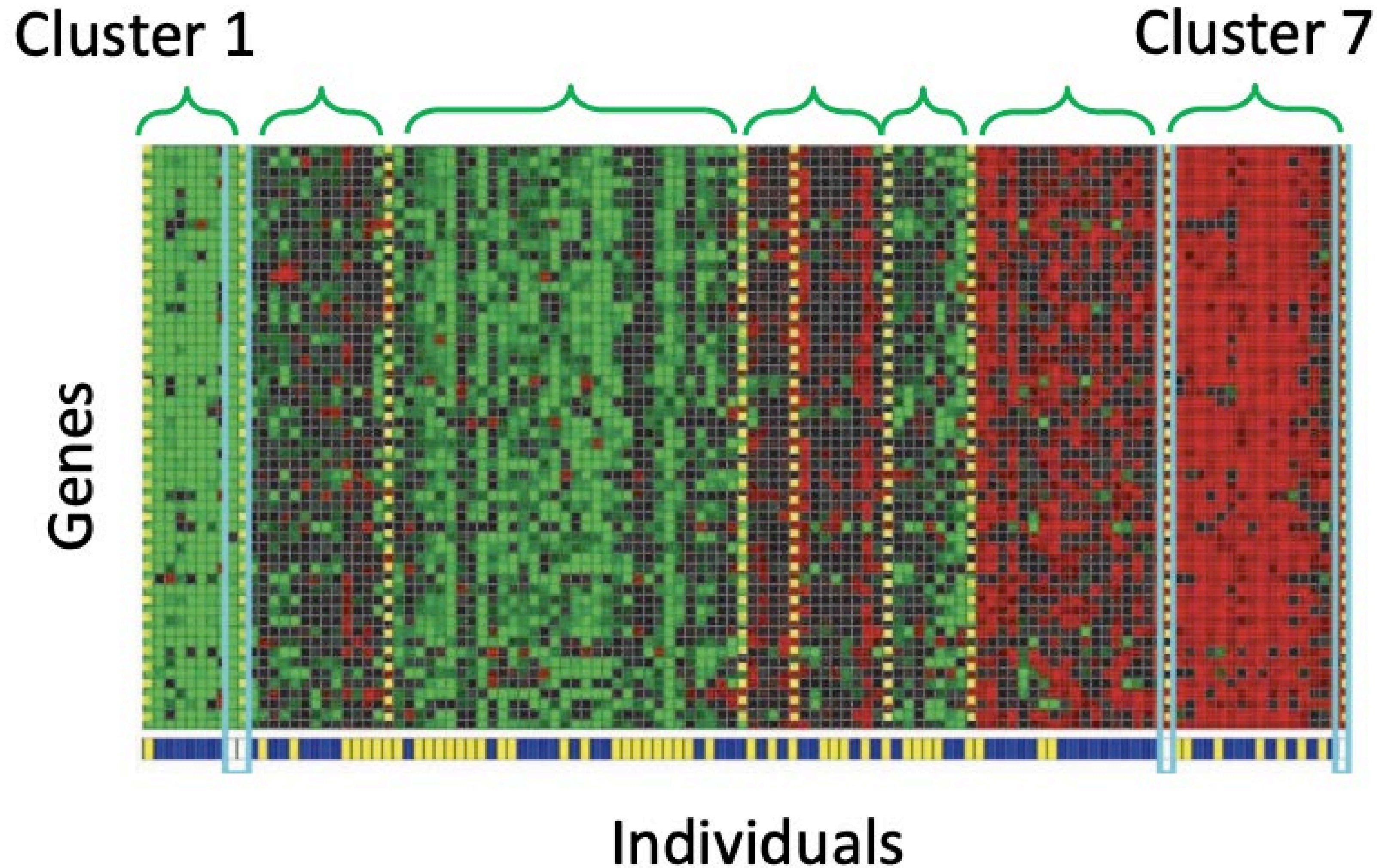
# Clustering

- You probably have  $>1000$  digital photos stored on your phone
- After this class you will be able to organize them better (based on visual similarity)



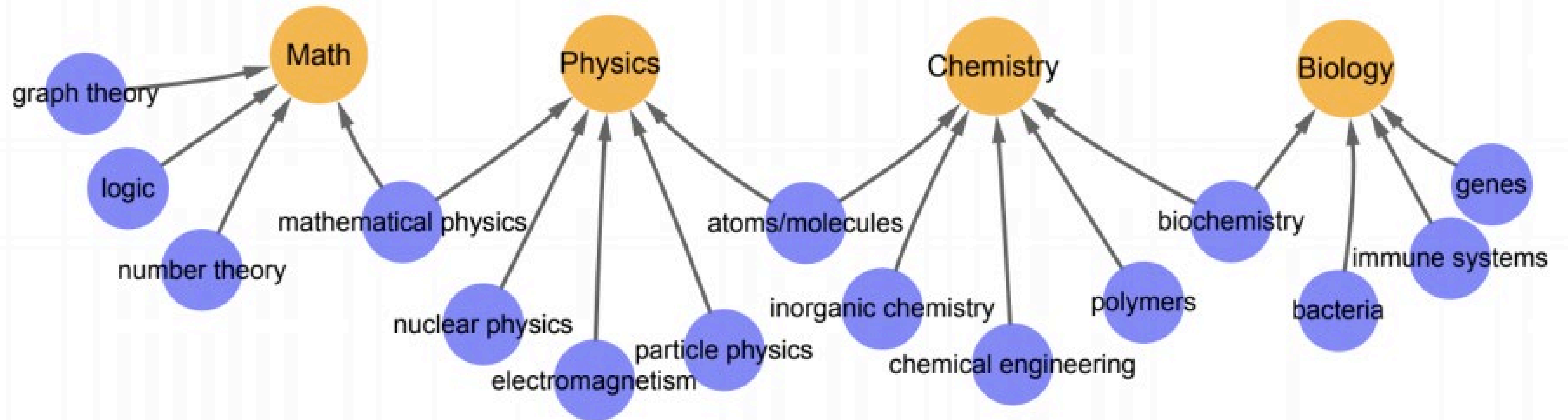


# Clustering Genes



Identifying Regulatory Mechanisms using Individual Variation Reveals Key Role for Chromatin Modification. [Su-In Lee, Dana Pe'er, Aimee M. Dudley, George M. Church and Daphne Koller. '06]

# Clustering Words with Similar Meanings



[Arora-Li-Liang-Ma-Risteski, TACL'17,18]

# How do we perform clustering?

Many clustering algorithms.

We will look at the two most frequently used ones:

- *K-means clustering*: we specify the desired number of clusters, and use an iterative algorithm to find them
- *Hierarchical clustering*: we build a binary tree over the dataset



# Quiz Break

Q2-1: Which is true about machine learning?

- A. The process doesn't involve human inputs
- B. The machine is given the training and test data for learning
- C. In clustering, the training data also have labels for learning
- D. Supervised learning involves labeled data

# Quiz Break

Q2-1: Which is true about machine learning?

- A. The process doesn't involve human inputs
- B. The machine is given the training and test data for learning
- C. In clustering, the training data also have labels for learning
- D. Supervised learning involves labeled data**

- A. The labels are human inputs
- B. The machine should not have test data for learning
- C. No labels available for clustering

# Quiz Break

Q2-2: Which is true about unsupervised learning?

- A. There are only 2 unsupervised learning algorithms
- B. Kmeans clustering is a type of hierarchical clustering
- C. Kmeans algorithm automatically determines the number of clusters  $k$
- D. Unsupervised learning is widely used in many applications



# Quiz Break

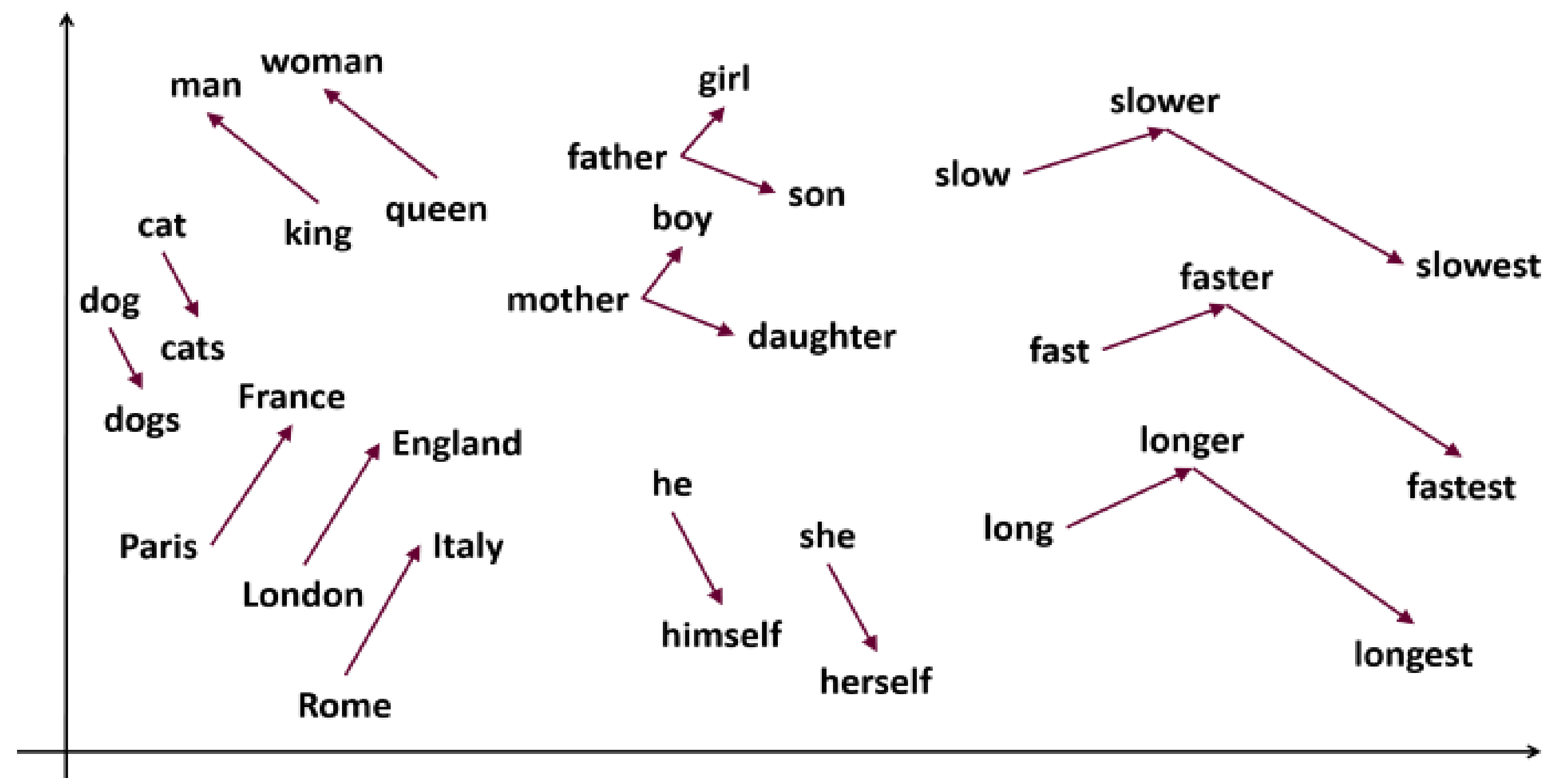
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# Self-Supervised Learning

- Given: dataset contains **no label**  $x_1, x_2, \dots, x_n$
- **Goal:** discover interesting patterns and structures in the data
- **Approach:** generate supervision signal from data.  
Solve a *pretext task*

Example: word embeddings



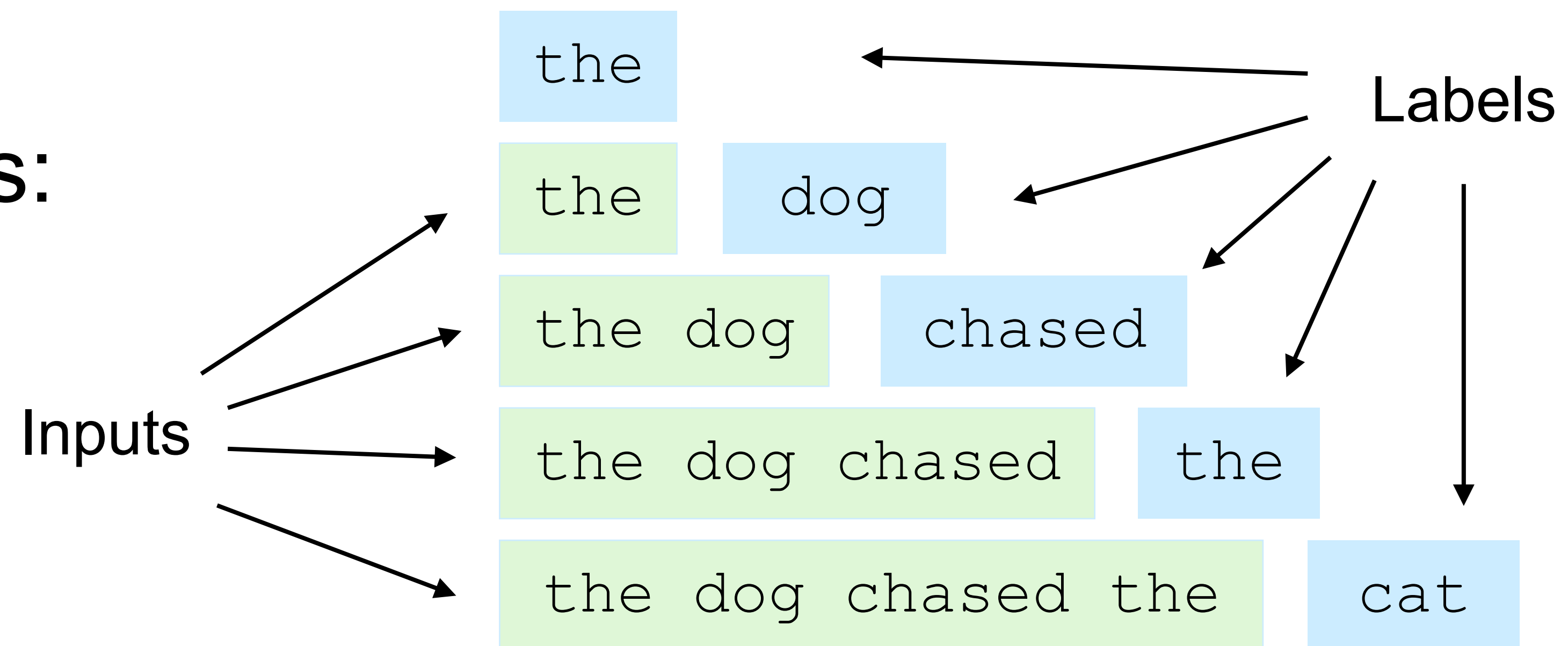
# Self-Supervised Learning for LLMs

- Pretext task for large language models:  
**next-word prediction**

- Original text:

the dog chased the cat

- Split into five labeled problems:



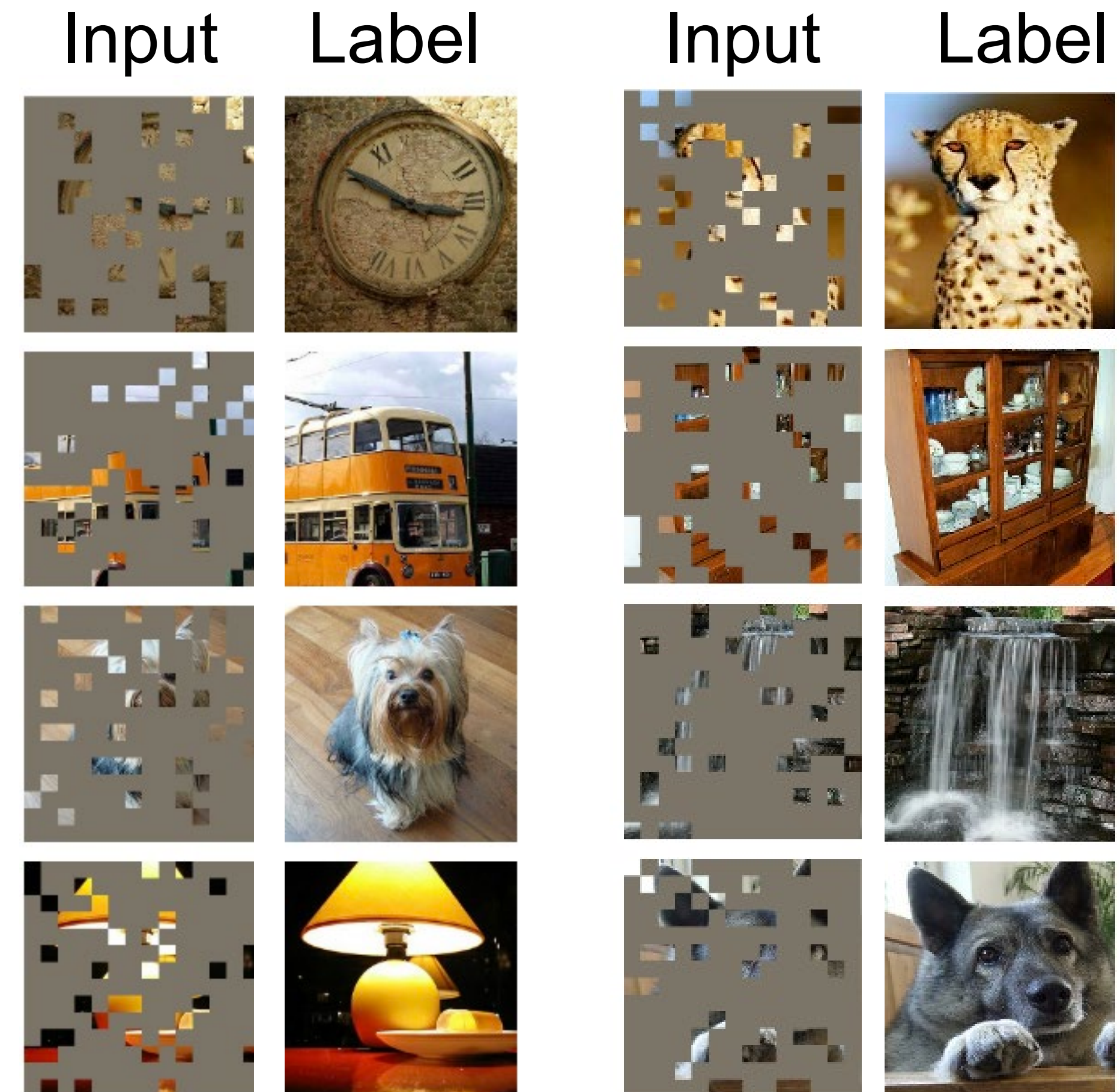


# Self-Supervised Learning in Vision

- Another common pretext task: **image inpainting**

- High-dimensional label!

- Type of **autoencoder**
  - “Auto-” = “self”







## Part III: Reinforcement Learning (Learning from rewards)







# Reinforcement Learning

- Given: an agent that can take actions and a reward function specifying how good an action is.
- **Goal:** learn to choose actions that maximize future reward total.



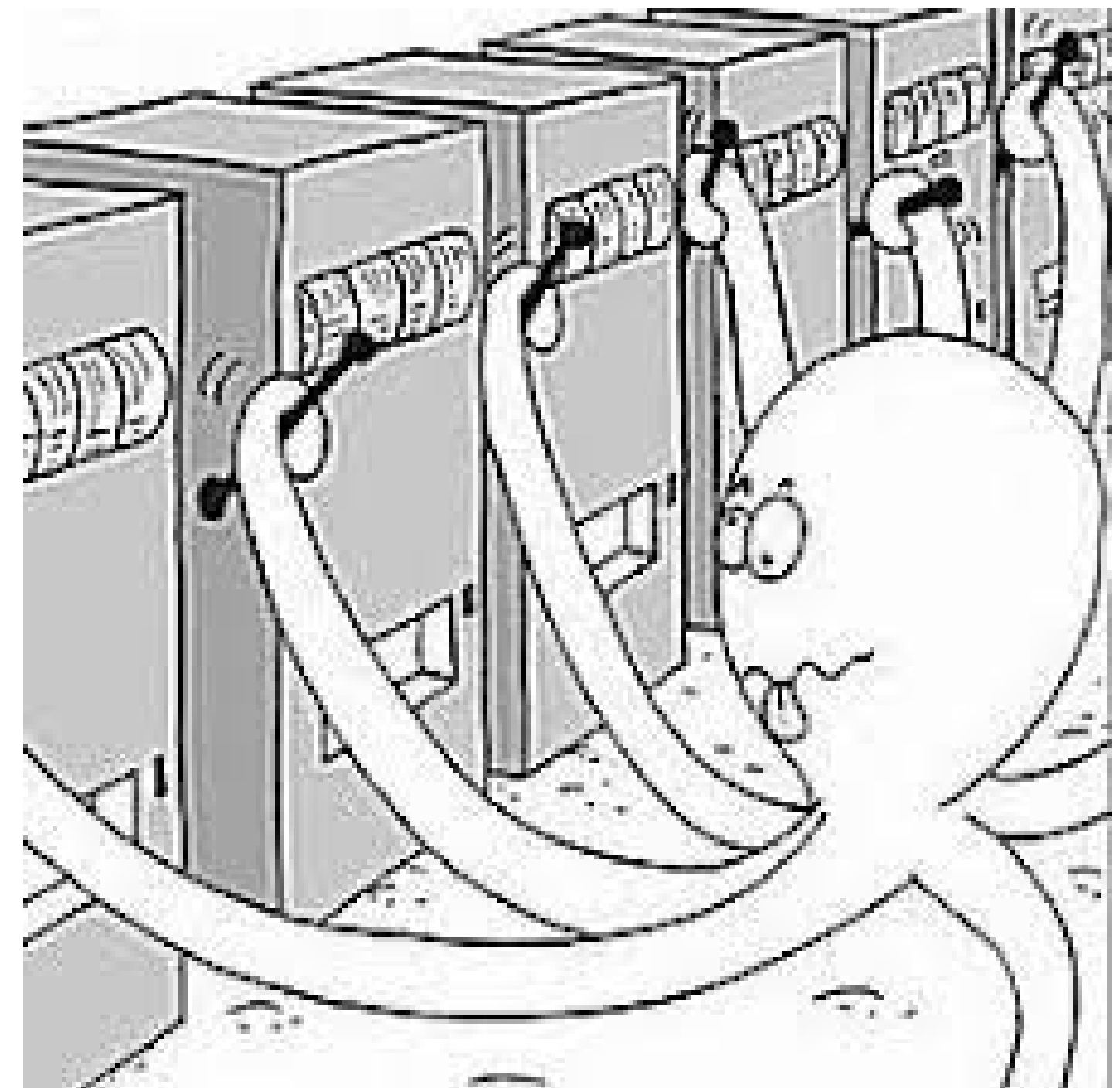
Google Deepmind

# Reinforcement Learning Key Problems

1. Problem: actions may have delayed effects.
  - Requires **credit-assignment**
2. Problem: maximal reward action is unknown
  - Exploration-exploitation trade-off

“..the problem [exploration-exploitation] was proposed [by British scientist] to be dropped over Germany so that German scientists could also waste their time on it.”

- Peter Whittle



Multi-armed Bandit

# Today's recap

- What is machine learning?
- Supervised Learning
  - Classification
  - Regression
- Unsupervised Learning
  - Clustering
  - Self-Supervised Learning
- Reinforcement Learning



# Suggested Readings

- Textbook: Artificial Intelligence: A Modern Approach (4th edition). Stuart Russell and Peter Norvig. Pearson, 2020. Sections 19.1





**Thanks!**