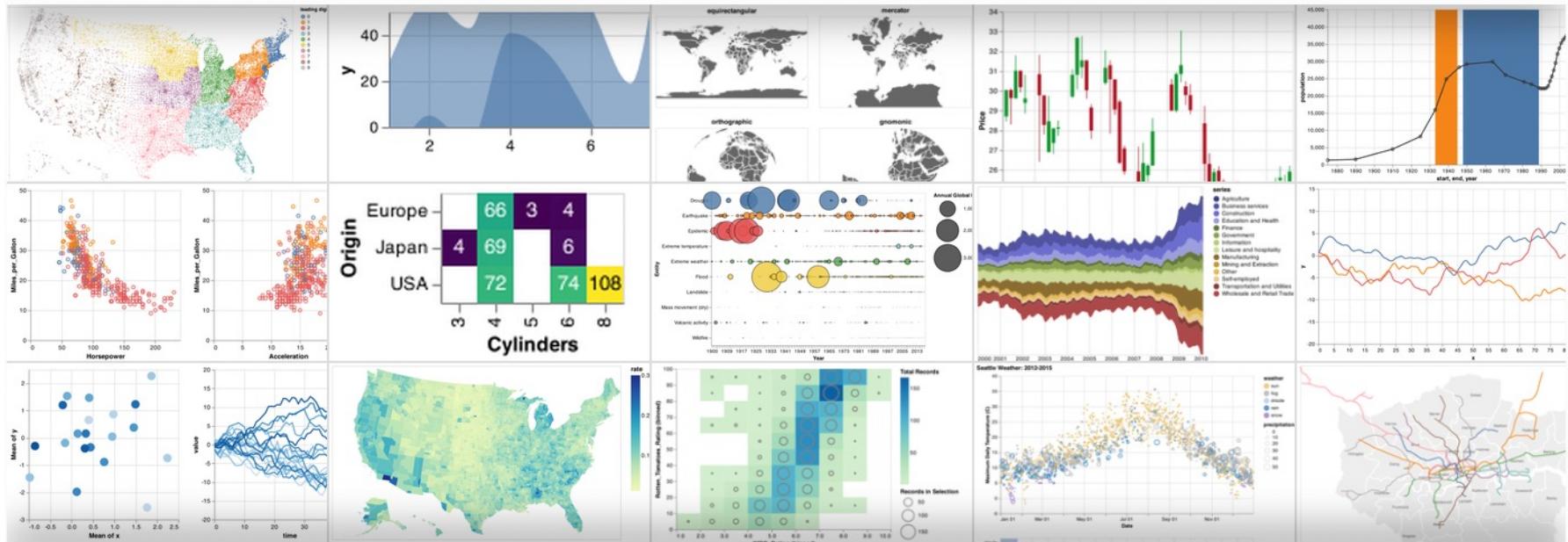


# Math 10, Fall 2022

## Introduction to Programming for Data Science

- First half of Math 10: Exploratory Data Analysis
- Second half: Introduction to Machine Learning



# What is Machine Learning?

Informal definition, from *Hands on Machine Learning* by Aurélien Géron:

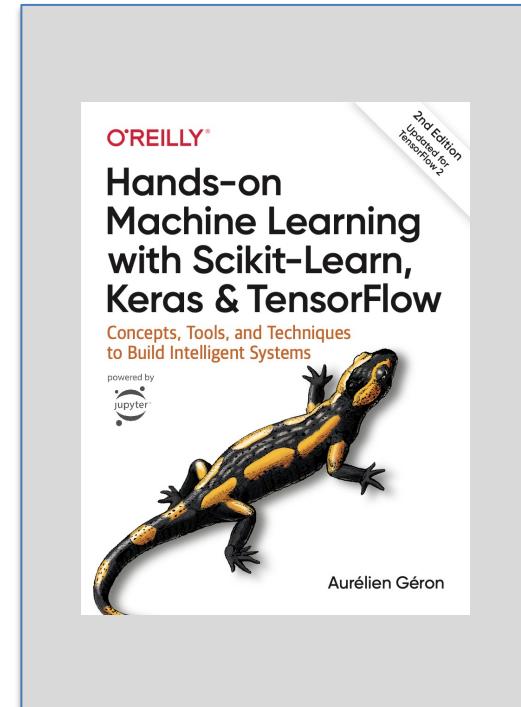
Machine Learning is the science (and art) of programming computers so they can learn from data.

(Pretty vague. Is computing a mean an example of Machine Learning?)

Two types of ML problems:

Supervised Learning: Have labeled data

Unsupervised Learning: No labels

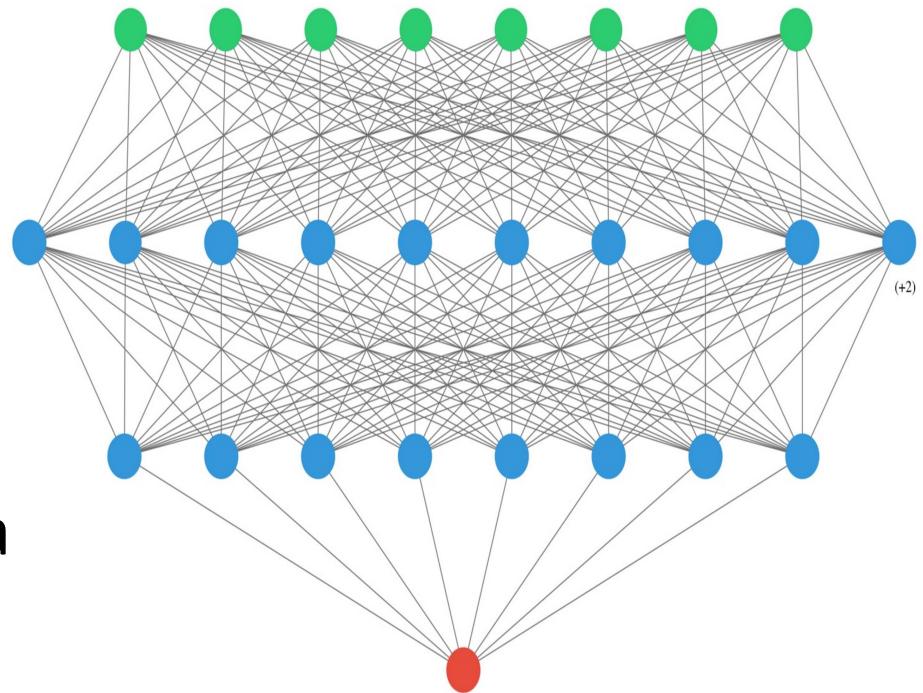


# Supervised Learning

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Two types of supervised learning problems:

- **Regression**: predicting a quantitative value.
- **Classification**: predicting a category.



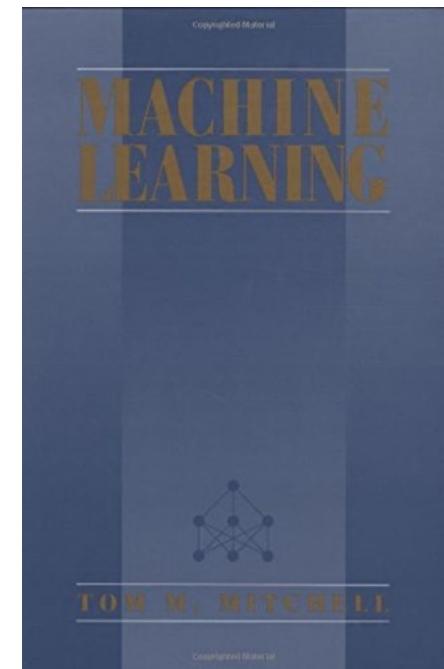
# A more formal Definition

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(Adapted from Mitchell, *Machine Learning*, 1997.)

## A definition of learning from data:

Consider a collection of tasks  $T$ , a performance measure  $P$ , a baseline strategy  $B$ , and an algorithm  $A$  which depends on a set of training data  $D$ . The algorithm  $A$  is said to learn from the data  $D$ , if its performance at tasks in  $T$ , as measured by  $P$ , is better than the baseline strategy  $B$ .



# Example: Classification

- **Task:** Determine if an email is spam.
- **Baseline strategy:** Predict spam if the email contains 3 or more exclamation points (!).
- **Training data:** Emails that have been identified as spam/not spam.
- **Performance measure:** Percentage correctly identified.

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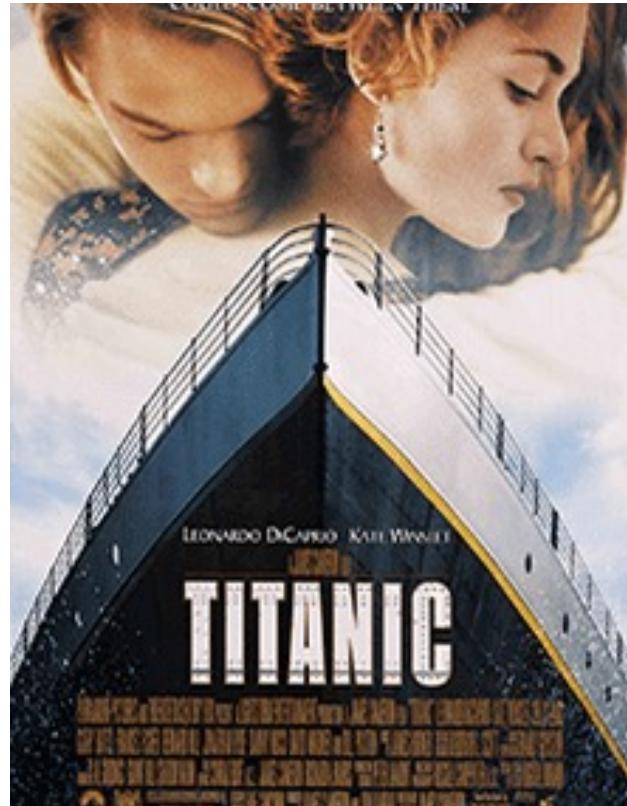
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# Example: Classification

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- **Task:** Estimate the probability that a Titanic passenger survived.
- **Baseline strategy:** Use the average survival rate as prediction.
- **Training data:** Survival outcomes and passenger characteristics.
- **Performance measure:** Log loss (severe penalty for being both confident and wrong)



# Example: Regression

- **Task:** Predict prices of houses in King County, Washington, based on characteristics.
- **Baseline strategy:** Always predict the median house value.
- **Training data:** Prices of certain houses in the county.
- **Performance measure:** Mean Absolute Error (less concern with outliers than Mean Squared Error)



Source: Wikimedia Commons, Hannah Lewis House, Jon Roanhaus

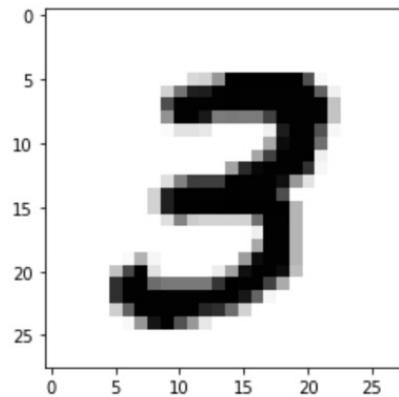
# Example: Classification

- **Task:** Identify the values of handwritten digits (given pixel values).
- **Baseline strategy:** Always predict 0.
- **Training data:** Handwritten digits together with their correct values.
- **Performance measure:** Percentage of digits correctly identified.

No. 4	No. 9	No. 0	No. 4	No. 6	No. 3	No. 8
4	9	0	4	6	3	8

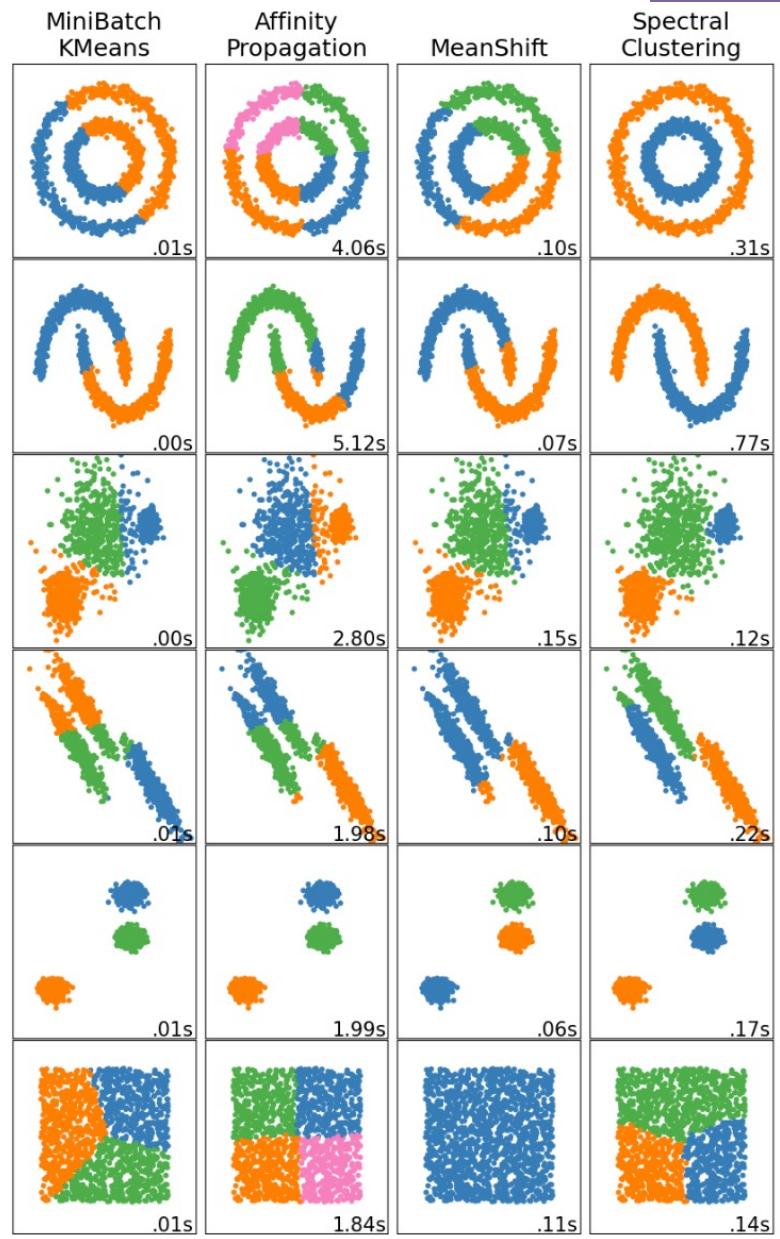
No. 5	No. 3	No. 6	No. 8	No. 7	No. 7	No. 9
5	3	6	8	7	7	9



# Example: Unsupervised Learning

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- **Task:** Divide data into K distinct clusters.
- **Baseline strategy:** Assign randomly.
- **Training data:** 100 sample points.
- **Performance measure:** Average distance of a sample point to the nearest centroid.



# Example: Unsupervised Learning

- **Task:** Generate art.
- **Baseline strategy:** Random pixel values.
- **Training data:** Collection of artworks together with evaluations by an expert.
- **Performance measure:** Was an expert tricked into thinking the artwork was made by a human?



# Example: Unsupervised Learning

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- **Task:** Reduce the dimensionality of NumPy arrays representing images of faces.
- **Baseline strategy:** Keep only the center-most 36 pixel values
- **Training data:** A collection of images of faces.
- **Performance measure:** Similarity of the reduced face to the original image.

