

## MACHINE LEARNING ALGORITHMS

- In 1959, Arthur Samuel defined **machine learning** as a "Field of study that gives computers the ability to learn without being explicitly programmed"
- **Machine learning** explores the study and construction of **algorithms** that can learn from and make predictions on data

# Hypothesis testing

A **hypothesis test** is a statistical **test** that is used to determine whether there is enough evidence in a sample of data to infer that a certain condition is true for the entire population

## Elements of a hypothesis test:

**Null hypothesis** - Statement regarding the value(s) of unknown parameter(s). Typically will imply **no association** between **explanatory** and **response variables** in our applications (will always contain an equality)

**Alternative hypothesis** - Statement contradictory to the null hypothesis (will always contain an inequality)

**Test statistic** - Quantity based on sample data and null hypothesis used to test between null and alternative hypotheses

**Rejection region** - Values of the test statistic for which we reject the null in favor of the alternative hypothesis

# Example - Efficacy Test for New drug

- Drug company has new drug, wishes to compare it with current standard treatment
- Federal regulators tell company that they must demonstrate that new drug is better than current treatment to receive approval
- Firm runs clinical trial where some patients receive new drug, and others receive standard treatment
- Numeric response of therapeutic effect is obtained (higher scores are better).
- Parameter of interest:  $m_{\text{New}} - m_{\text{Std}}$

# Example - Efficacy Test for New drug

**Null hypothesis** - New drug is no better than standard trt

$$H_0: \mu_{\text{new}} - \mu_{\text{std}} \leq 0 \quad (\mu_{\text{new}} - \mu_{\text{std}} = 0)$$

**Alternative hypothesis** - New drug is better than standard trt

$$H_A : \mu_{\text{new}} - \mu_{\text{std}} > 0$$

# Types of Machine Learning

Machine learning tasks are typically classified into **three** broad **categories**, depending on the **nature of the learning** "signal" or "feedback" available to a **learning system**. These are:

1. Supervised learning
2. Unsupervised learning
3. Reinforcement learning

- **Supervised learning:** The computer is presented with example inputs and their desired outputs, given by a "teacher", and the goal is to learn a general rule that maps inputs to outputs
- **Supervised learning** is the machine learning task of inferring a function from labeled training data.
- **Unsupervised learning: No labels** are given to the learning algorithm, leaving it on its own to find structure in its input. Unsupervised learning can be a goal in itself (discovering hidden patterns in data) or a means towards an end.
- **Unsupervised learning** is the machine learning task of inferring a function to describe hidden structure from unlabeled data

**Reinforcement learning:** A **computer program** interacts with a dynamic environment in which it must perform a certain goal (such as driving a vehicle), **without a teacher** explicitly telling it whether it has come close to its goal or not. Another example is **learning to play a game by playing against an opponent.**

Between supervised and unsupervised learning is semi-supervised learning, where the **teacher** gives an **incomplete training signal**: a training set with some (often many) of the target outputs missing.

**Transduction** is a special case of this principle where the entire set of problem instances is known at learning time, except that part of the targets is missing.



- A **support vector machine** is a **classifier** that divides its **input space** into **two regions**, separated by a **linear boundary**. Here, it has learned to distinguish black and white circles.
- Among other categories of machine learning problems, learning to learn learns its own inductive bias based on previous experience.