

# Introduction

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## Supervised learning

We want a mapping  $f$  from inputs  $\mathbf{x} \in \mathcal{X}$  ( $\mathcal{X} = \mathbb{R}^D$ ) to outputs  $y \in \mathcal{Y}$ .

- ▶  $\mathbf{x}$  : features, predictors, covariates
- ▶  $y$  : label, target, response
- ▶  $\mathcal{D} = \{(\mathbf{x}_i, y_i)\}_{i=1}^N$  : training set ( $(\mathbf{x}_i, y_i)$  is the  $i$ th observation or example)
- ▶  $D$  : number of features
- ▶  $N$  : number of training examples
- ▶  $\mathbf{X} \in \mathcal{R}^{N \times D}$  : Design matrix

**Classification** :  $\mathcal{Y} = \{1, 2, \dots, C\}$   
where  $C$  is the number of classes

**Regression** :  $\mathcal{Y} = \mathbb{R}$

## Unsupervised learning

Only inputs  $\mathcal{D} = \{\mathbf{x}_i\}_{i=1}^N$  are observed without any corresponding outputs. We want to find structure, “meaning” in data.

## Reinforcement learning

An agent learns to interact with its environment. A policy defines which action to take in response to each possible environment state. A reward signal is received in response to the actions the agent takes.

## Other types of ML

- ▶ Semi-supervised learning
- ▶ Self-supervised learning

■ **"Pure" Reinforcement Learning (cherry)**

- ▶ The machine predicts a scalar reward given once in a while.
- ▶ **A few bits for some samples**

■ **Supervised Learning (icing)**

- ▶ The machine predicts a category or a few numbers for each input
- ▶ Predicting human-supplied data
- ▶ **10→10,000 bits per sample**

■ **Unsupervised/Predictive Learning (cake)**

- ▶ The machine predicts any part of its input for any observed part.
- ▶ Predicts future frames in videos
- ▶ **Millions of bits per sample**

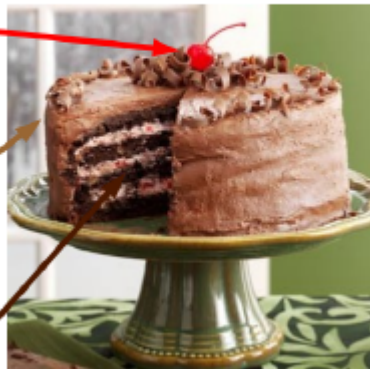


Figure: Machine learning types as layers of a cake by Yann Le Cun at NIPS'2016

## Models

- ▶ A (statistical) **model**  $\mathfrak{F}$  is a set of distributions or densities or functions.
- ▶ A **parametric model**  $\mathfrak{F}$  is a set of distributions or densities or functions that can be parametrized by a finite number of parameters.

$$\mathfrak{F} = \left\{ f(x; \theta) : \theta \in \Theta \right\}$$

sometimes just  $f(x; \theta)$  where  $\theta$  is the parameter

- ▶ A **non-parametric model** is a set  $\mathfrak{F}$  that cannot be parameterized by a finite number of parameters.

# Machine Learning Systems

A Machine learning system is generally composed of:

- ▶ Dataset
- ▶ Model (with parameters and hyperparameters)
- ▶ Loss function
- ▶ Optimization (learning) algorithm
- ▶ Regularization strategy

## Other related fields

- ▶ Predictive Analytics
- ▶ Statistics
- ▶ Data Mining
- ▶ Data Science
- ▶ Business Intelligence
- ▶ Artificial Intelligence

See ML glossary

<https://developers.google.com/machine-learning/glossary/>

# Prerequisites

## Must-have

- ▶ Programming
- ▶ Probability and statistics
- ▶ Linear algebra
- ▶ Convex optimization
- ▶ Multivariate calculus

## Recommended

- ▶ Analysis of algorithms
- ▶ Information Theory
- ▶ Numerical computation

## Practical skills

- ▶ Python
- ▶ Bash
- ▶ SQL
- ▶ Git
- ▶ Scrum
- ▶ DevOps
- ▶ Writing and presentation skills



