

# Introduction to supervised machine learning, k-fold cross-validation, nearest neighbors, and linear models

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# Supervised machine learning

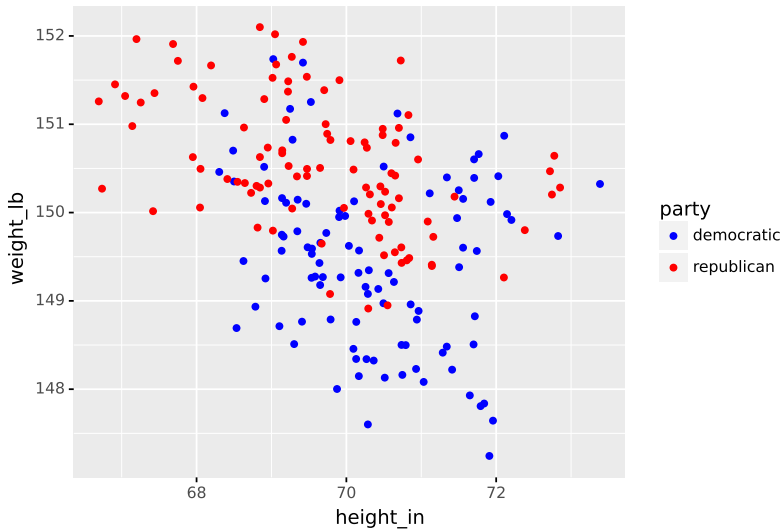
- ▶ Goal is to learn a function  $f(\mathbf{x}) = y$  where  $\mathbf{x}$  is an input/feature vector and  $y$  is an output/label.
- ▶  $x$  = image of digit/clothing,  $y \in \{0, \dots, 9\}$  (ten classes).
- ▶  $x$  = vector of word counts in email,  $y \in \{1, 0\}$  (spam or not).
- ▶  $x$  = image of retina,  $y$  = risk score for heart disease.
- ▶ This week we will focus on a specific kind of supervised learning problem called binary classification, which means  $y \in \{1, 0\}$ .

# Learning algorithm

- ▶ We want a learning algorithm `LEARN` which inputs a training data set and outputs a prediction function  $f$ .
- ▶ We typically represent a training data set with  $n$  observations and  $p$  features as a matrix  $\mathbf{X} \in \mathbb{R}^{n \times p}$  with a corresponding label vector  $\mathbf{y} \in \{0, 1\}^n$ .
- ▶ We will use three such data sets from Elements of Statistical Learning book by Hastie et al. (mixture slightly modified)

name	observations, $n$	inputs/features, $p$	outputs/labels
zip.test	images, 623	pixel intensities, 256	0/1 digits
spam	emails, 4601	word counts, 57	spam=1/not=0
mixture	people, 200	height/weight, 2	democratic/republican

## Visualize iris data with labels



## Visualize iris data without labels

- ▶ Let  $X \in \mathbb{R}^{150 \times 2}$  be the data matrix (input for clustering).