



## Introduction to Machine Learning and AI Boot Camp

Hands-on tutorials to help build skills required in Precision Child Health, data science and bioinformatics for SickKids staff and trainees.

Topics include:

- Introduction to machine learning
- Coding practice: regression and classification models
- Performance metrics and validation
- Neural network architectures
- Model training and optimization
- Open-source repositories

Prerequisite: Prior knowledge of R is required to follow along with the coding practice section.

To register please RSVP at:

<https://ccm20251111.eventbrite.com/>

Organized by the Centre for Computational Medicine ([ccm.sickkids.ca](http://ccm.sickkids.ca)) at the SickKids Research Institute with support from other groups:



[www.sickkids.ca/research](http://www.sickkids.ca/research)



Digital Research  
Alliance of Canada

[alliancecan.ca](http://alliancecan.ca)



[computeontario.ca](http://computeontario.ca)

## Hands-on Bioinformatics Tutorials for Biologists

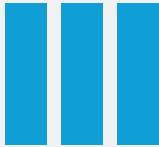
Tuesday Nov 11, 2025  
10 AM – 1 PM

Multimedia Room,  
PGCRL 3<sup>rd</sup> floor,  
or on Zoom



November 11, 2025

(A Whirlwind)  
Introduction to  
Machine Learning  
in R



# Agenda

Introduction to machine learning (ML)

ML approaches

- Supervised learning
- Unsupervised learning

Dataset overview

Coding practice

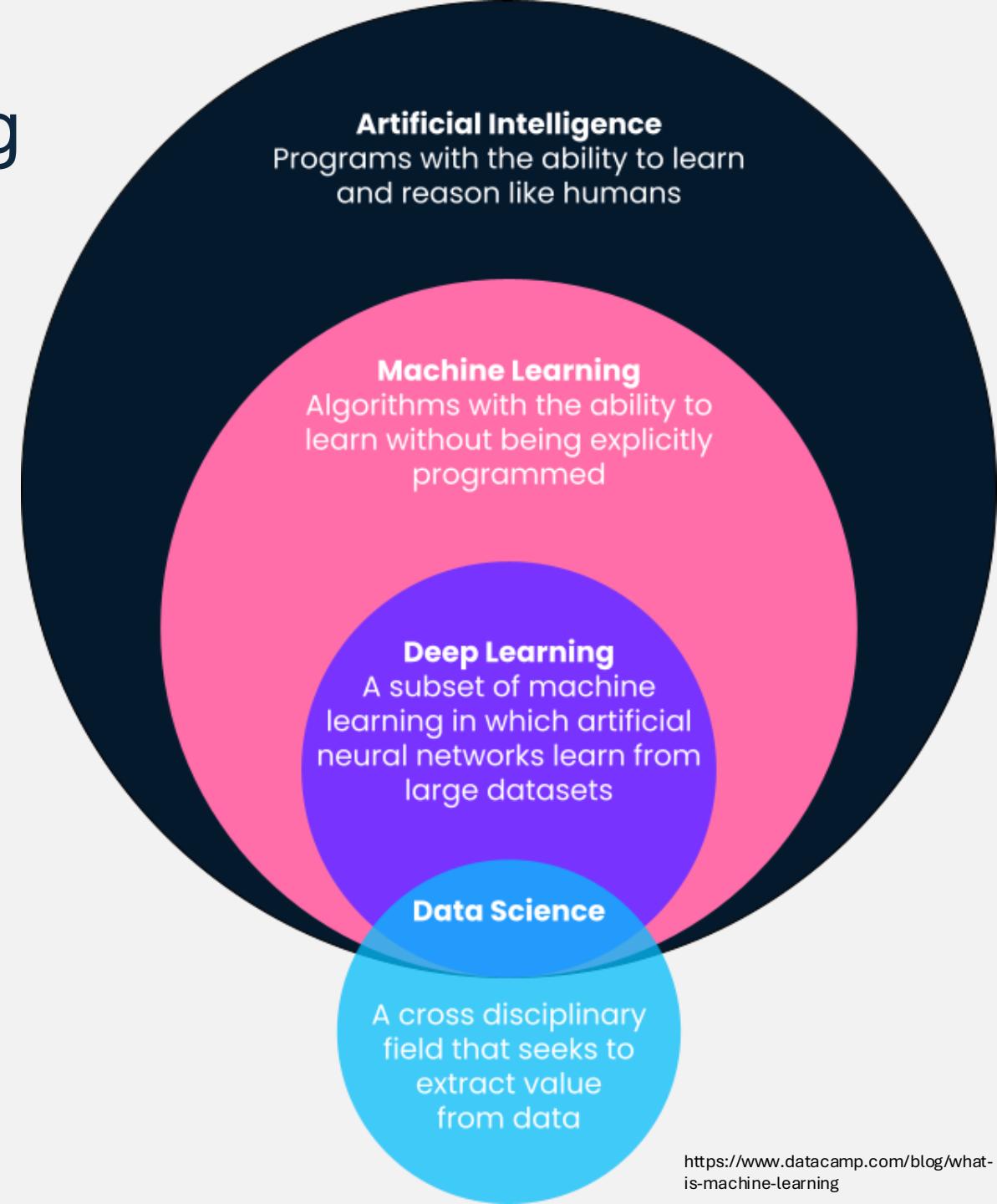
# Introduction to Machine Learning

What is machine learning (ML)?

A **branch of artificial intelligence** that involves the development and evaluation of statistical models and algorithms that **learn from data without following explicit instructions**.

ML has applications across industries:

- Disease prediction
  - Drug discovery
  - Financial risk assessment
  - Computer vision for self-driving cars
  - Supply chain optimization
  - Social media personalization
- ...and many others



# Machine Learning Approaches

## Supervised Learning

- Model is trained on **labeled data**.
- Classification: predict categorical outputs. Can be **binary** or **multi-class**.
- Regression: predict numerical values.
- Algorithms include:
  - Regression
  - Support vector machines
  - Decision trees



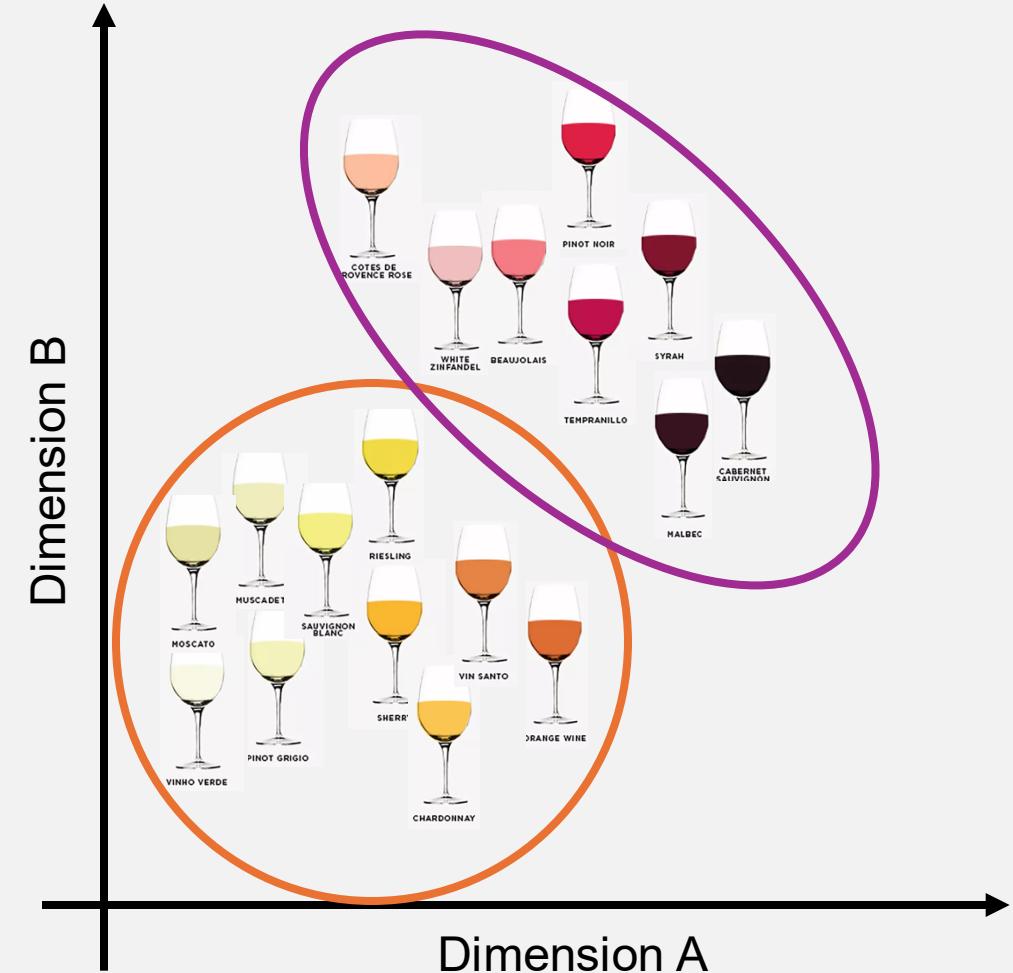
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## Unsupervised Learning

- Model is trained on **unlabeled data**.
- Clustering: find sub-groups in the data.
- Dimensionality reduction: reduce the number of features in the dataset without losing meaningful information.
- Algorithms include:
  - K-means clustering
  - Hierarchical clustering
  - Principal components analysis



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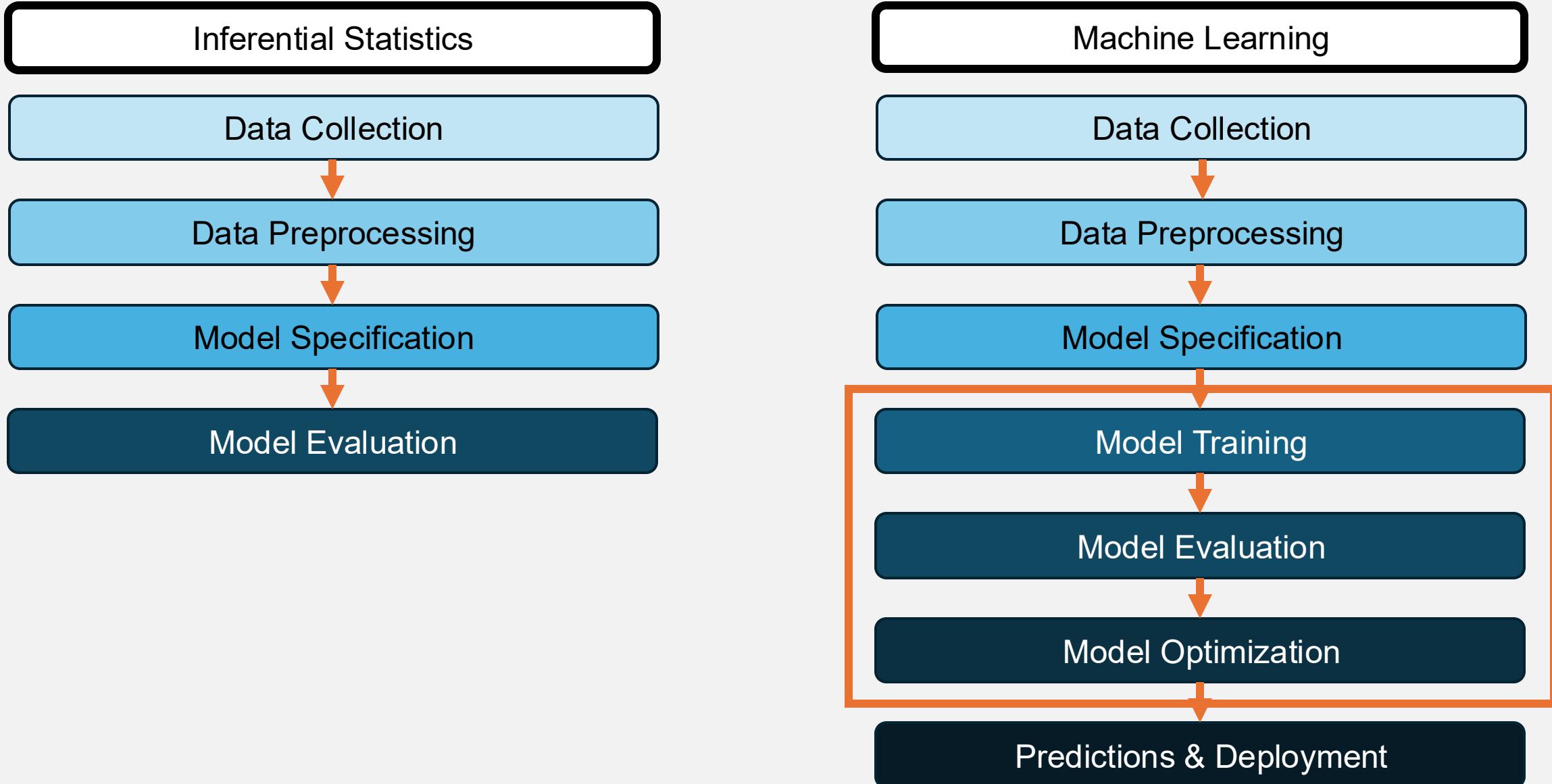
## Semi-supervised Learning

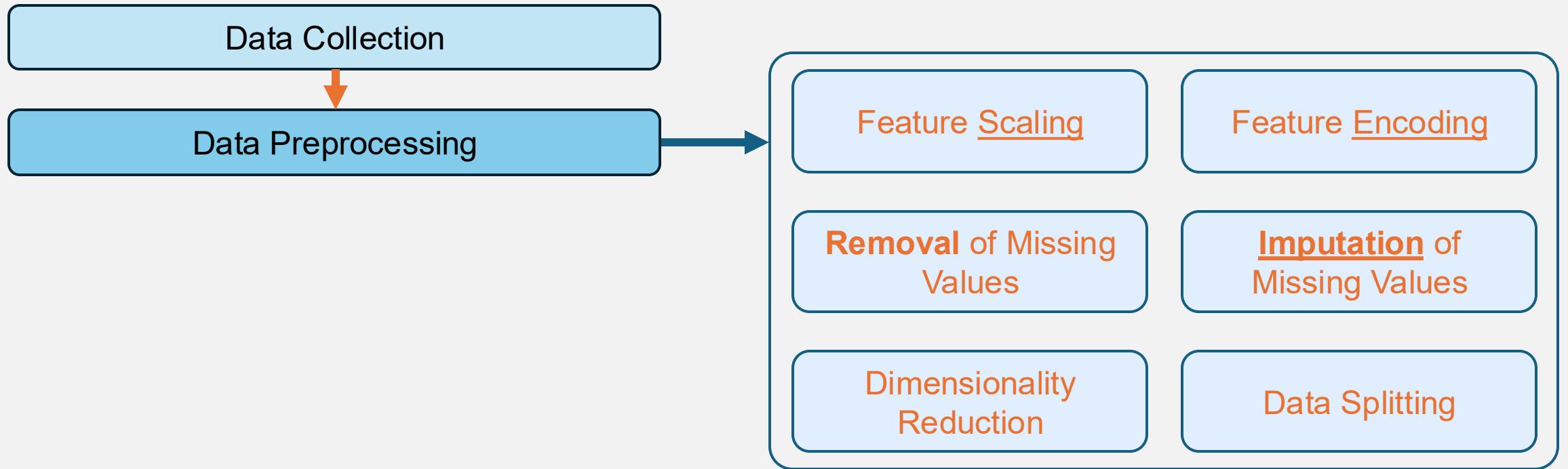
- Model is trained on both **labeled and unlabeled data**.
- Unlabeled data help identify patterns in the dataset while labeled data establish structure and guide learning (how many classes in the dataset?).
- Useful when working with limited or incompletely labeled data.

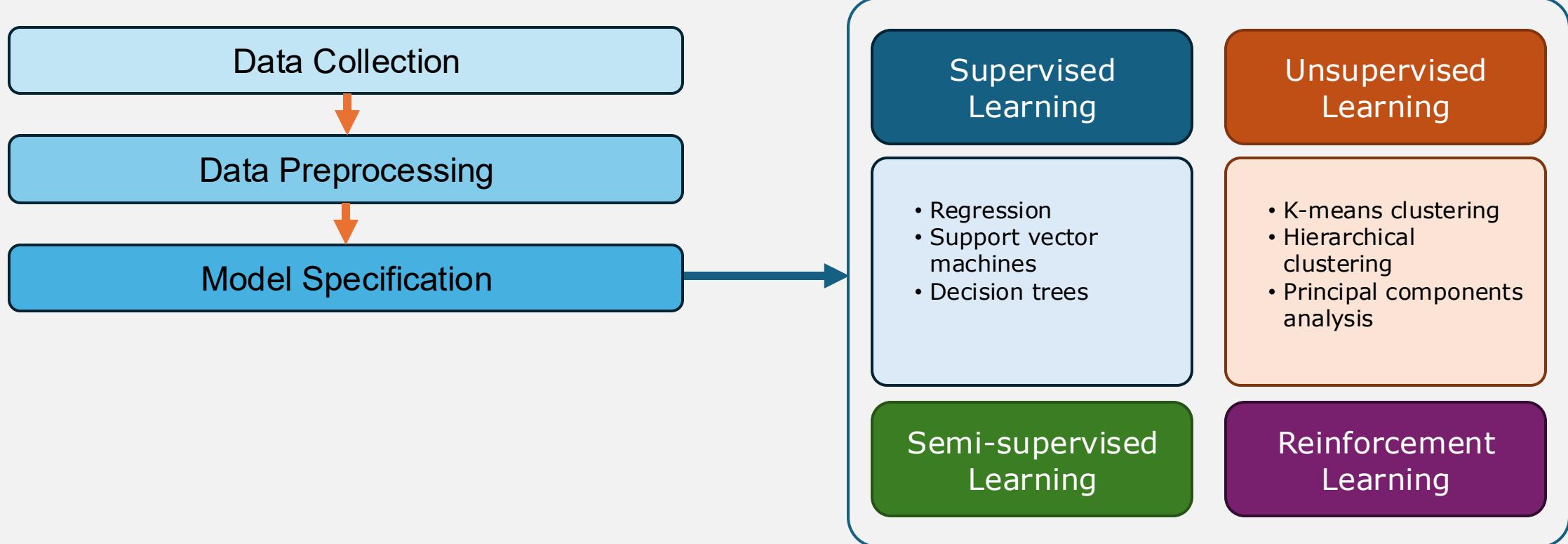
## Reinforcement Learning

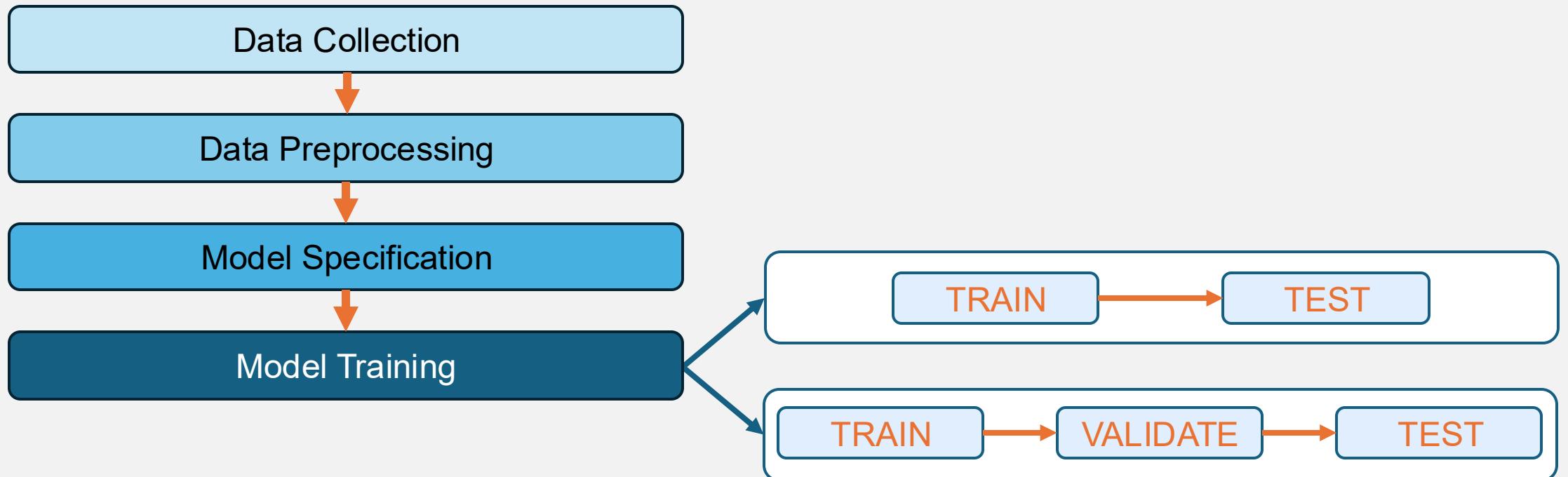
- Model is trained to learn from **positive (reward)** and **negative (punishment) feedback**.
- Algorithms include:
  - Value iteration
  - Markov decision process
  - Q learning
- Useful when learning can be sequential.

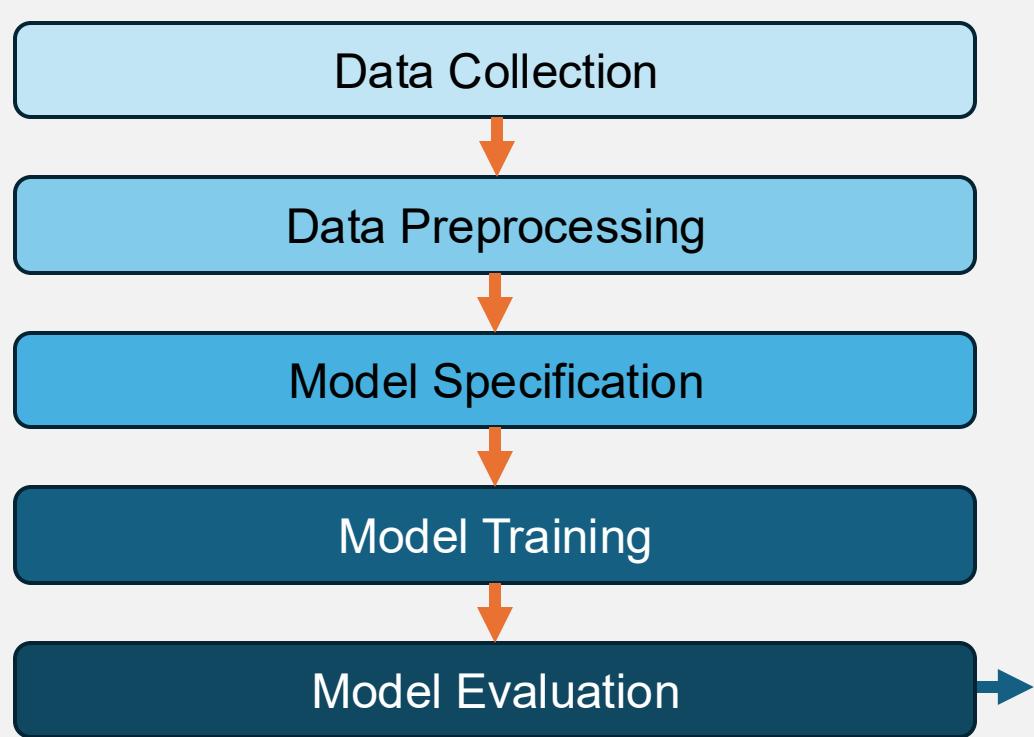
# Introduction to Machine Learning



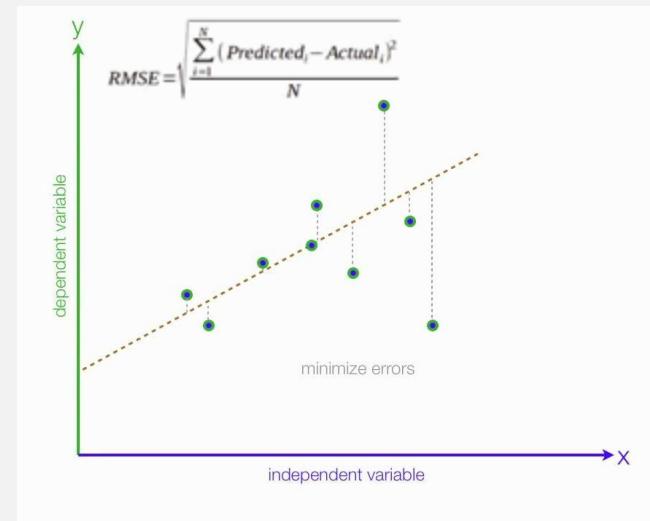
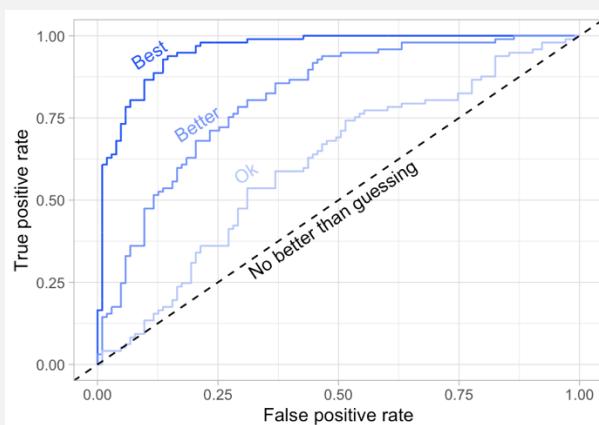
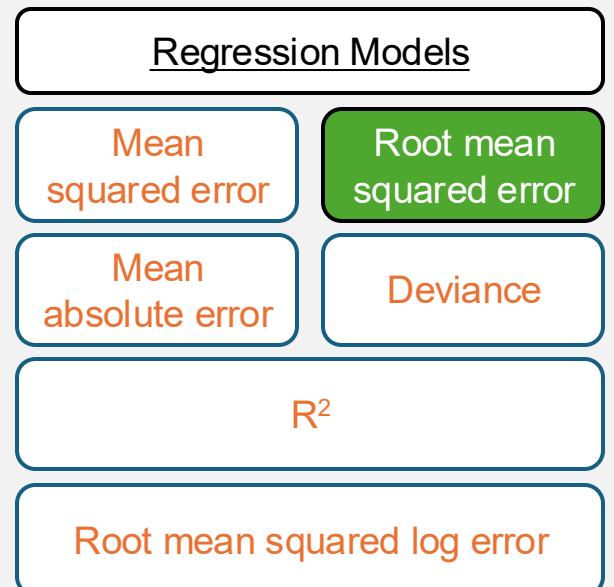
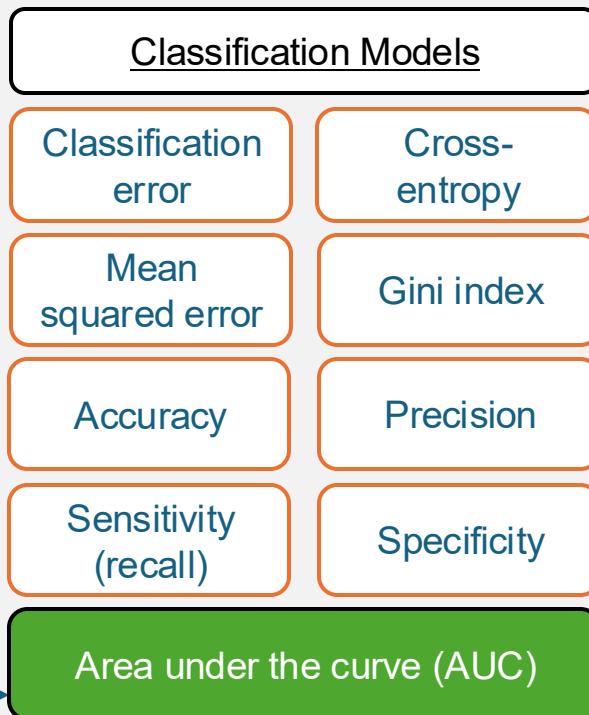
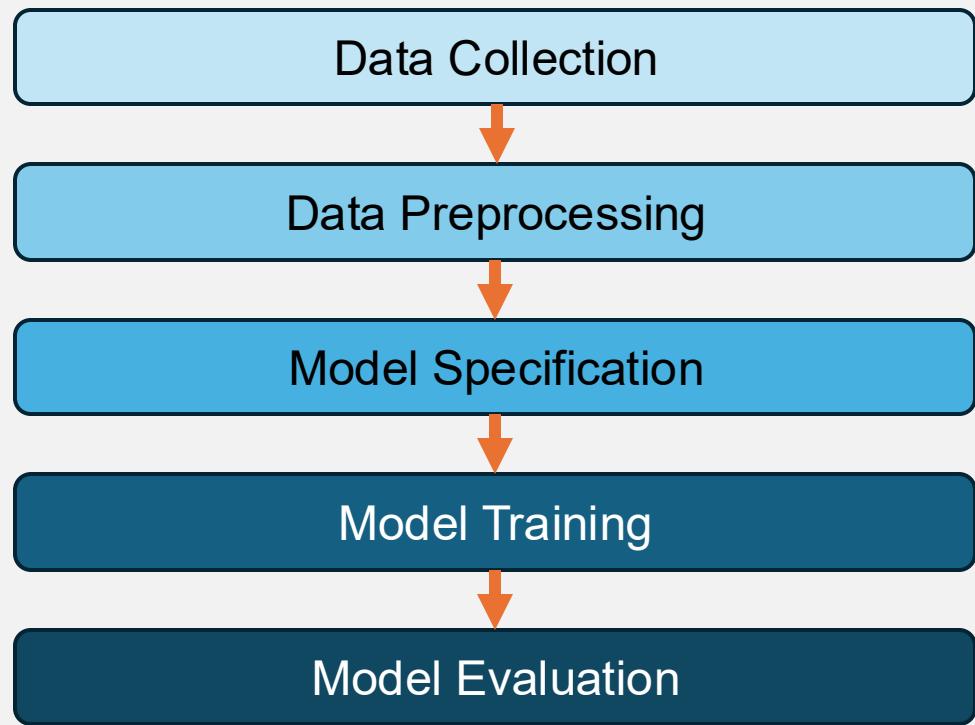


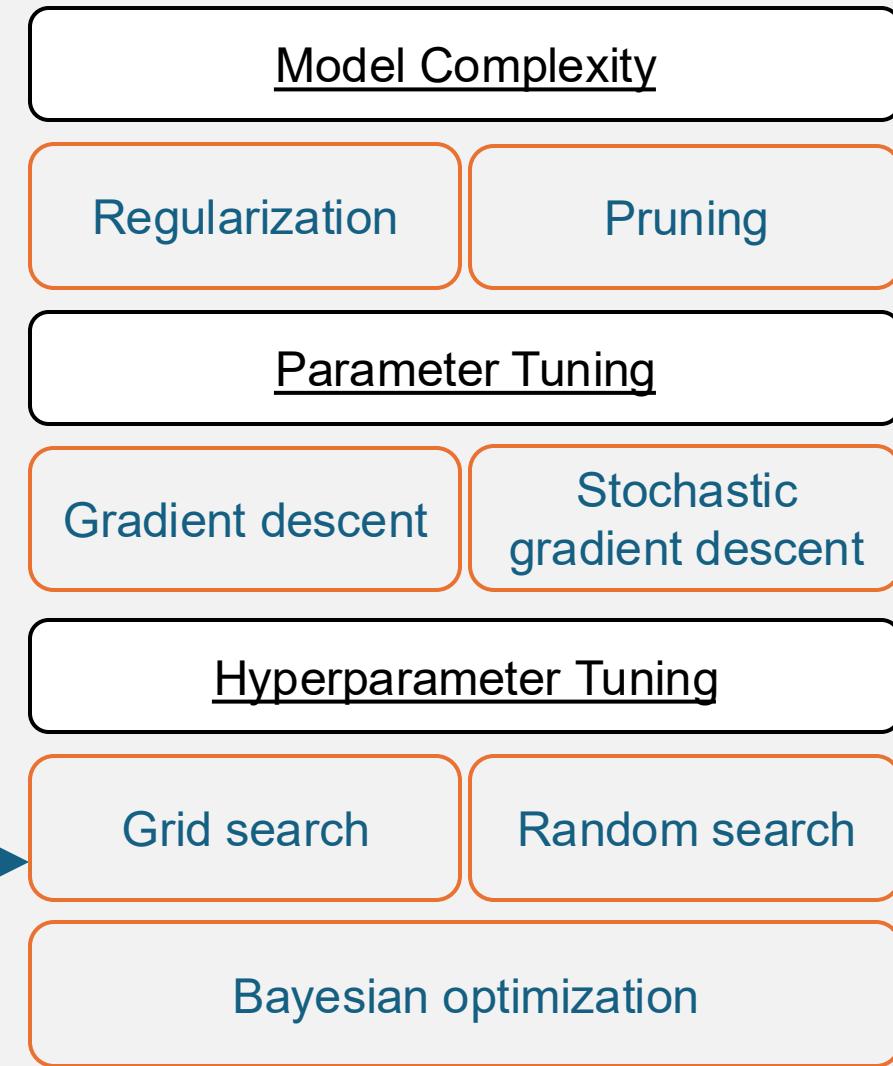
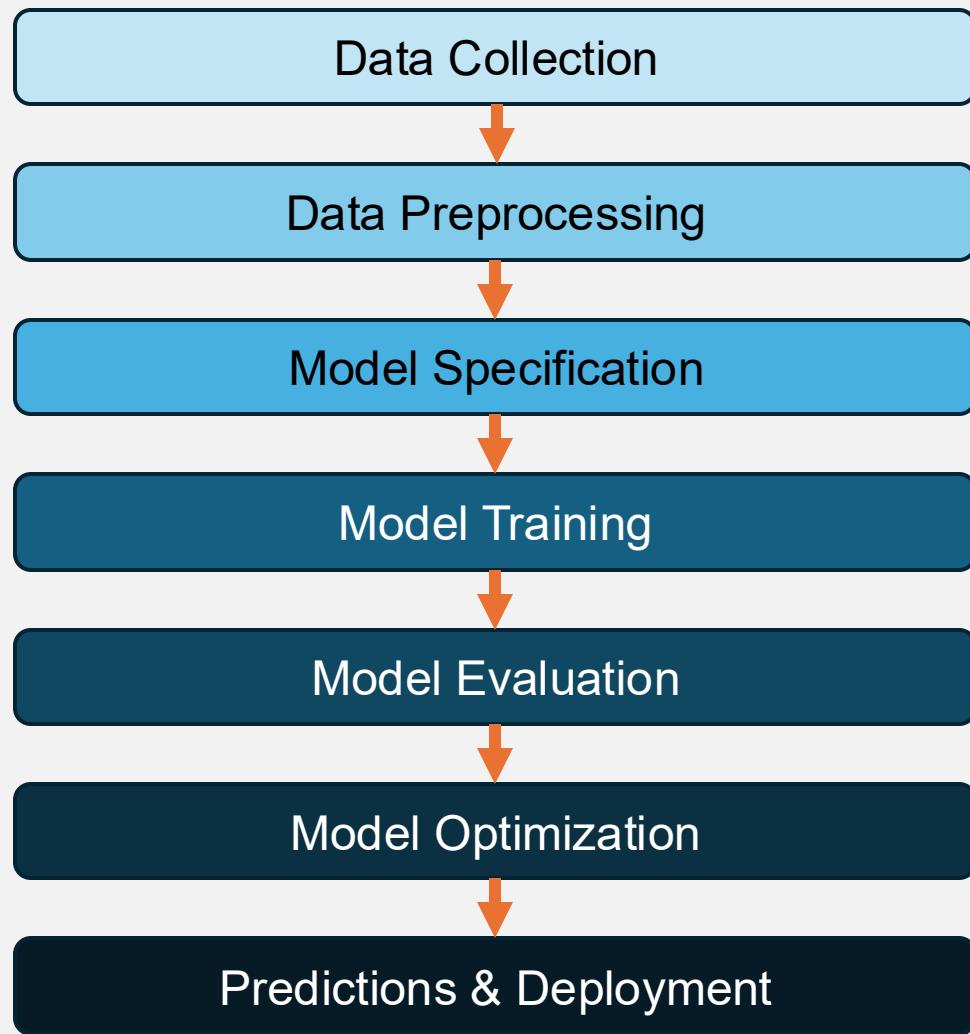






Classification Models		Regression Models	
Classification error	Cross-entropy	Mean squared error	Root mean squared error
Mean squared error	Gini index	Mean absolute error	Deviance
Accuracy	Precision	$R^2$	
Sensitivity (recall)	Specificity	Root mean squared log error	
Area under the curve (AUC)		True Outcome	
Predicted Outcome	Yes	True positive	False positive
	No	False negative	True negative





# CODING PRACTICE

# Framingham Heart Study

One of the longest prospective epidemiological studies of cardiovascular disease and its risk factors.

Began in 1948 in Framingham, MA

- Initially enrolled 5209 men and women aged 29-62 years old.
- Followed them over time, with assessments every 2 years.

Examinations included:

- Detailed medical history
- Physical exams
- Lab tests
- Lifestyle and habits
- Noninvasive imaging

# Framingham Heart Study Dataset

Subset of the data with 4000+ records and 16 variables.

(<https://www.kaggle.com/datasets/captainozlem/framingham-chd-preprocessed-data>)

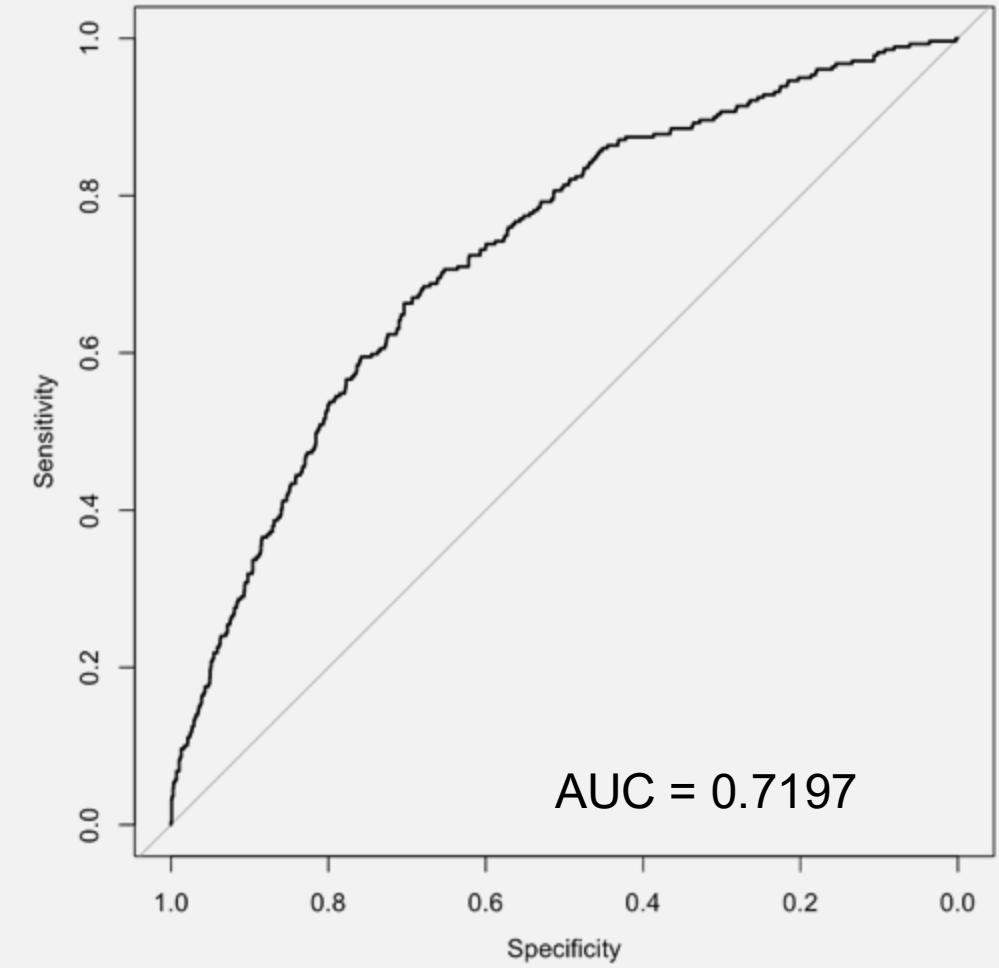
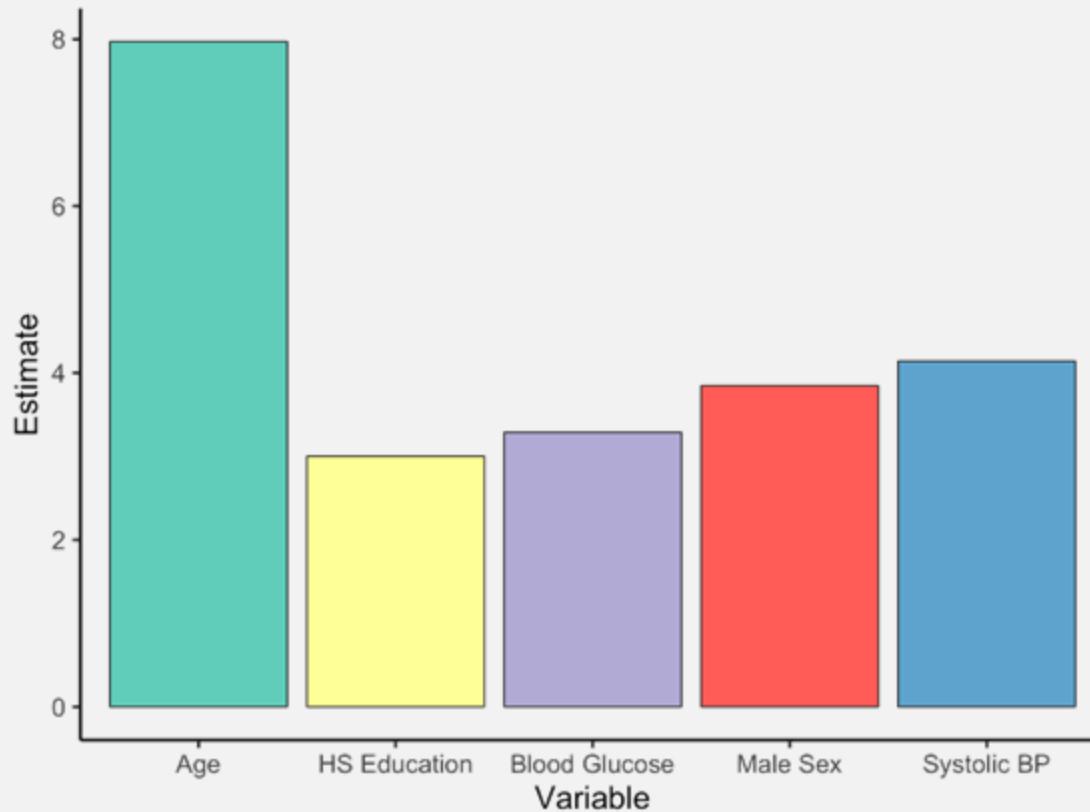
Variable	Description	Class/type
<b>male</b>	Sex (male or female)	Binary
<b>age</b>	Age	Continuous
<b>education</b>	Education (0-11 years, HS or GED, some uni, uni grad+)	Categorical
<b>currentSmoker</b>	Whether or not the patient is currently a smoker	Binary
<b>cigsPerDay</b>	Number of cigarettes smoked per day, on average	Continuous
<b>BPMeds</b>	Whether or not the patient is on blood pressure medication	Binary
<b>prevalentStroke</b>	Whether or not the patient had previously had a stroke	Binary
<b>prevalentHyp</b>	Whether or not the patient is hypertensive	Binary
<b>diabetes</b>	Whether or not the patient has diabetes	Binary
<b>totChol</b>	Total cholesterol level	Continuous
<b>sysBP</b>	Systolic blood pressure	Continuous
<b>diaBP</b>	Diastolic blood pressure	Continuous
<b>BMI</b>	Body mass index	Continuous
<b>heartRate</b>	Heart rate	Continuous
<b>glucose</b>	Glucose level	Continuous
<b>TenYearCHD</b>	10-year risk of coronary heart disease	Binary

# Binary Classification

## Logistic Regression

Predictors: all variables

Outcome variable: heart disease



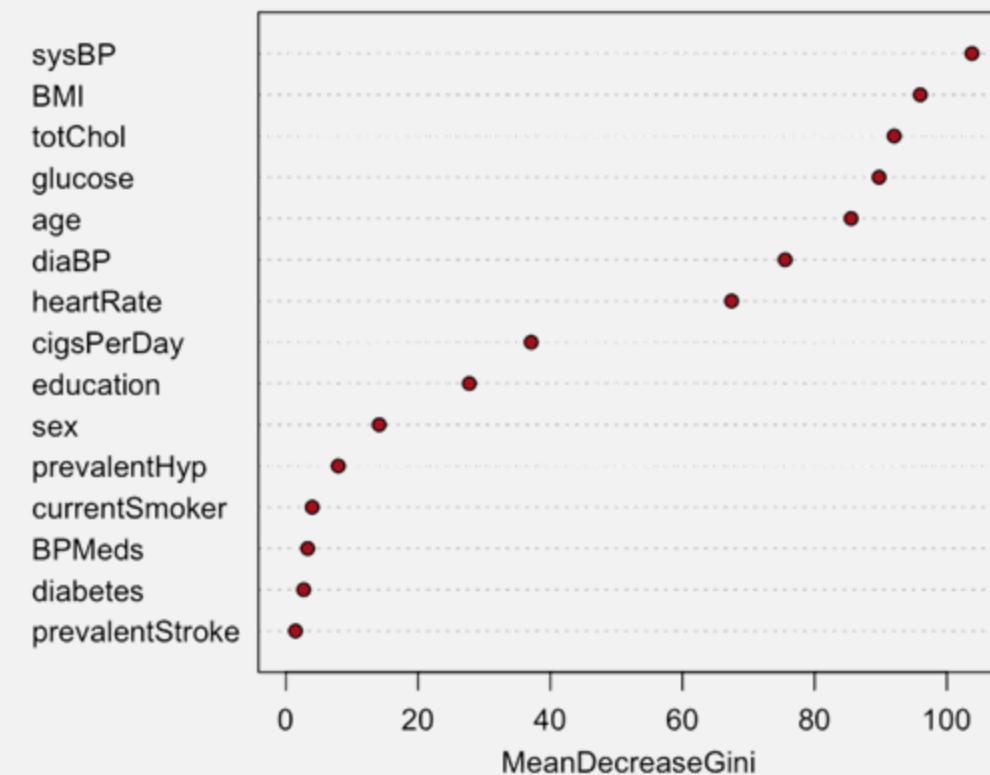
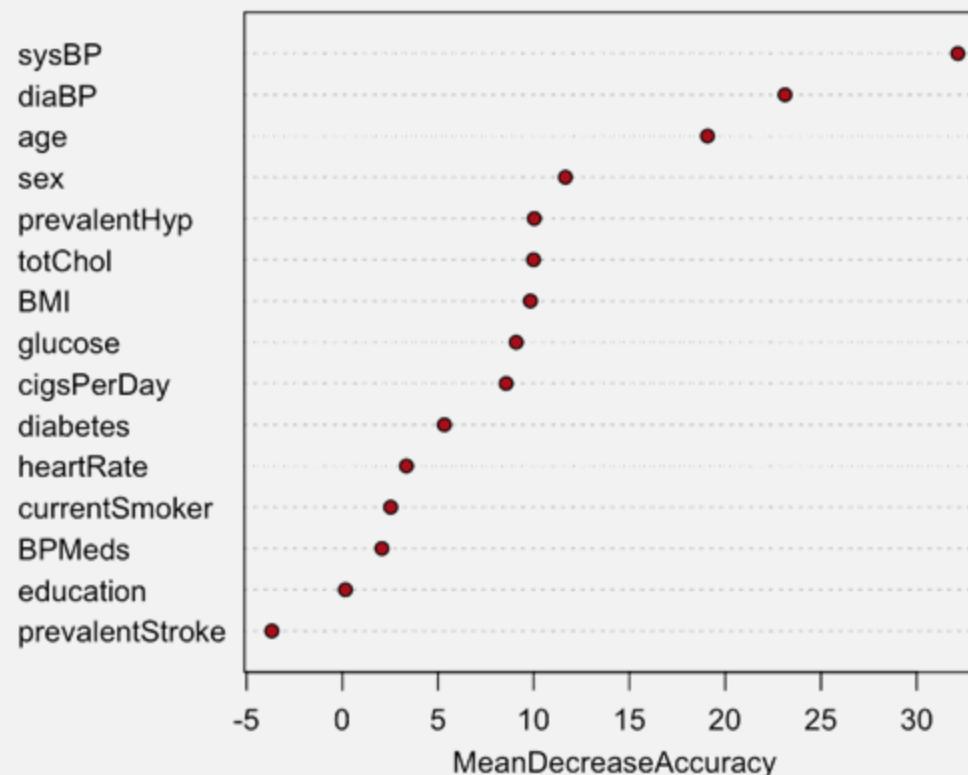
# Binary Classification

## Random Forest

Predictors: all variables

Outcome variable: heart disease

Variable Importance for Predicting Cholesterol Levels



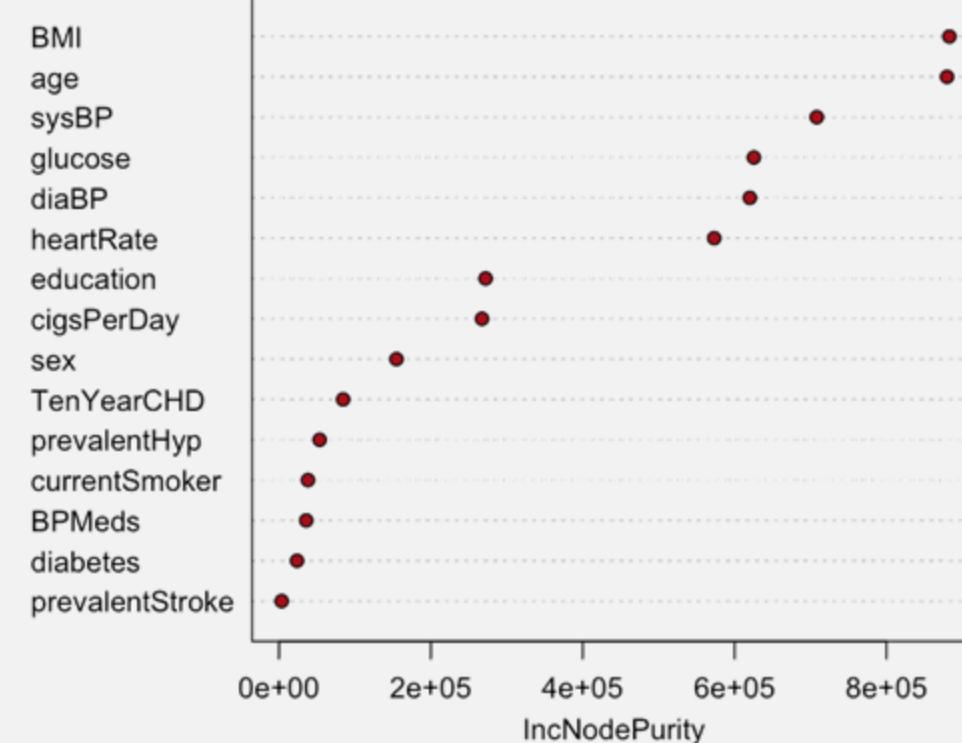
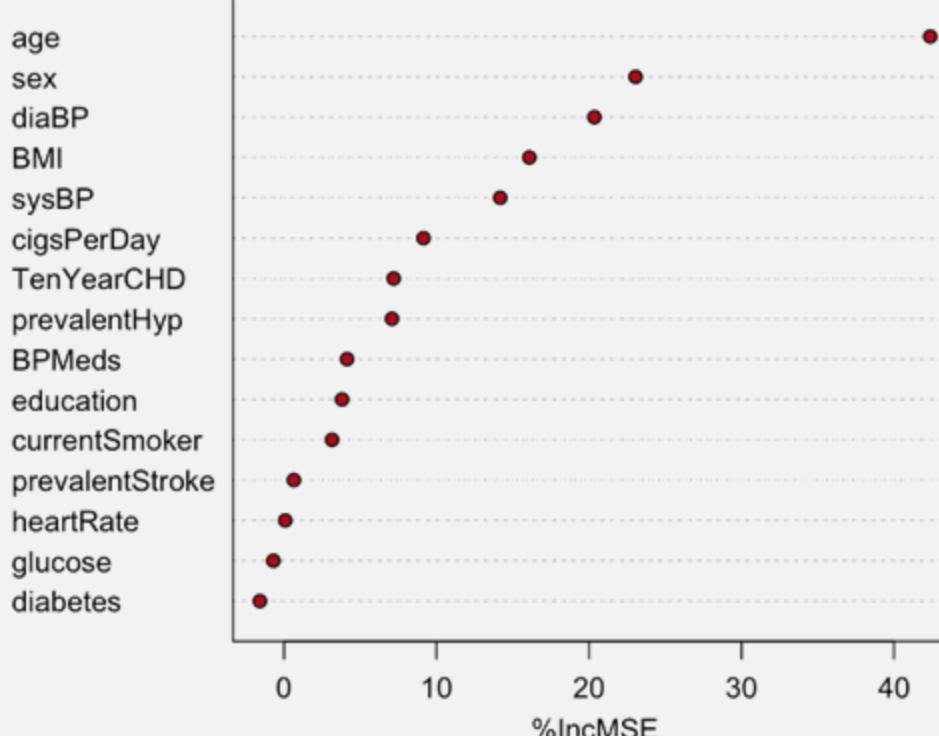
# Regression

## Random Forest

Predictors: all variables

Outcome variable: cholesterol levels

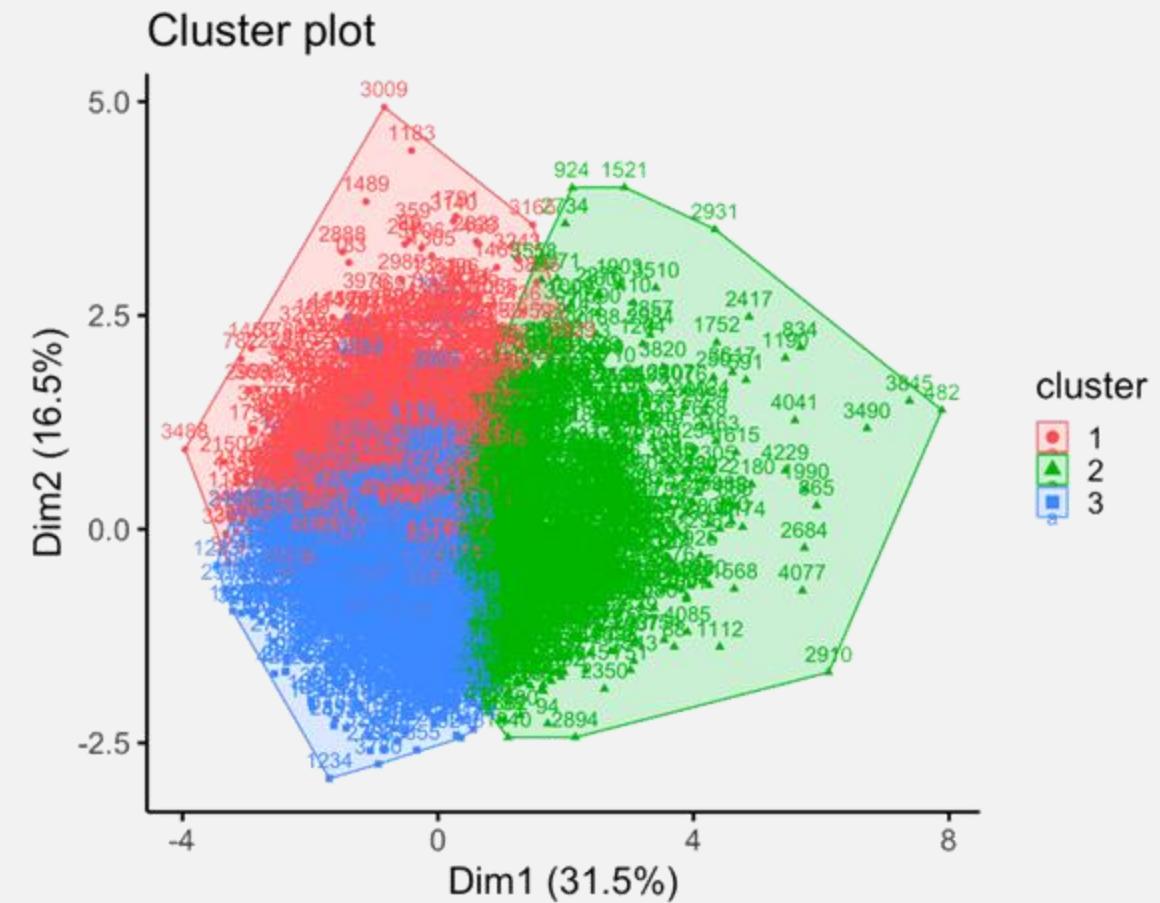
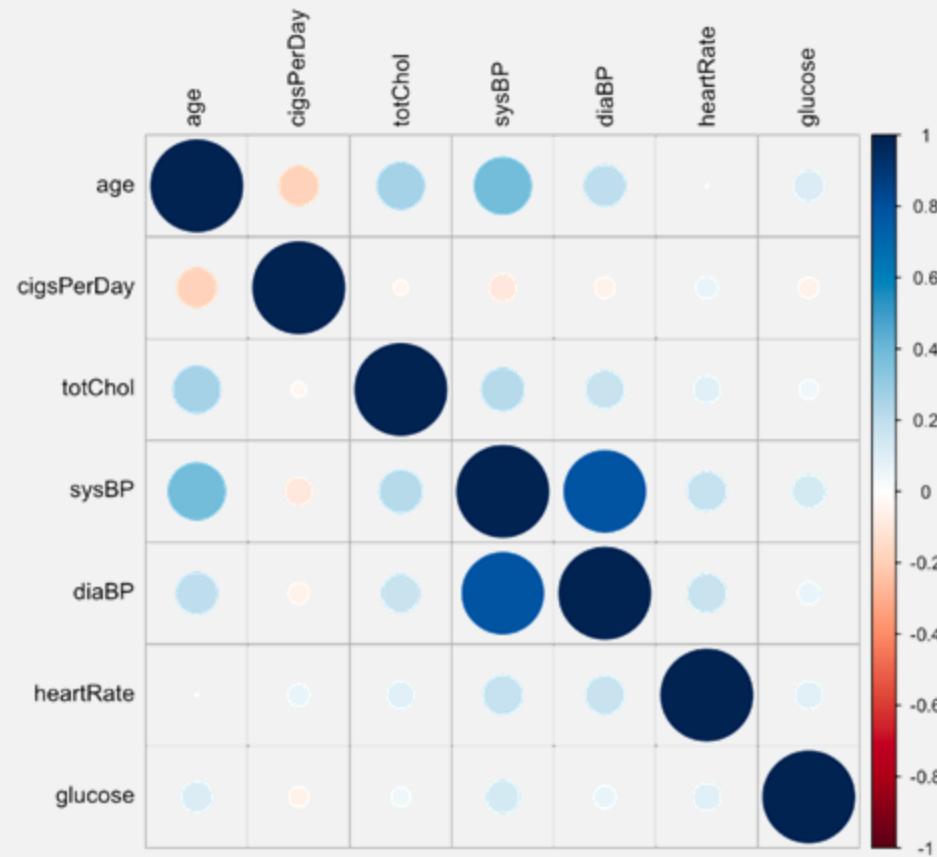
Variable Importance for Predicting Cholesterol Levels



# Clustering

## K-means

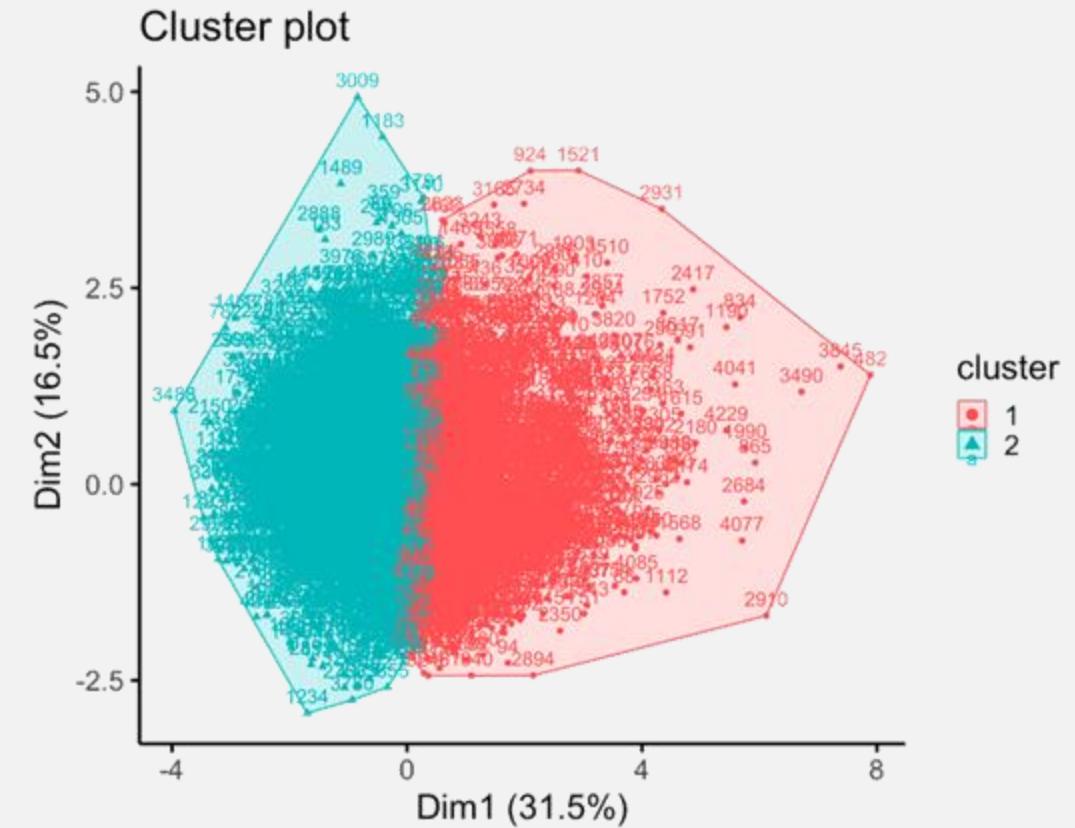
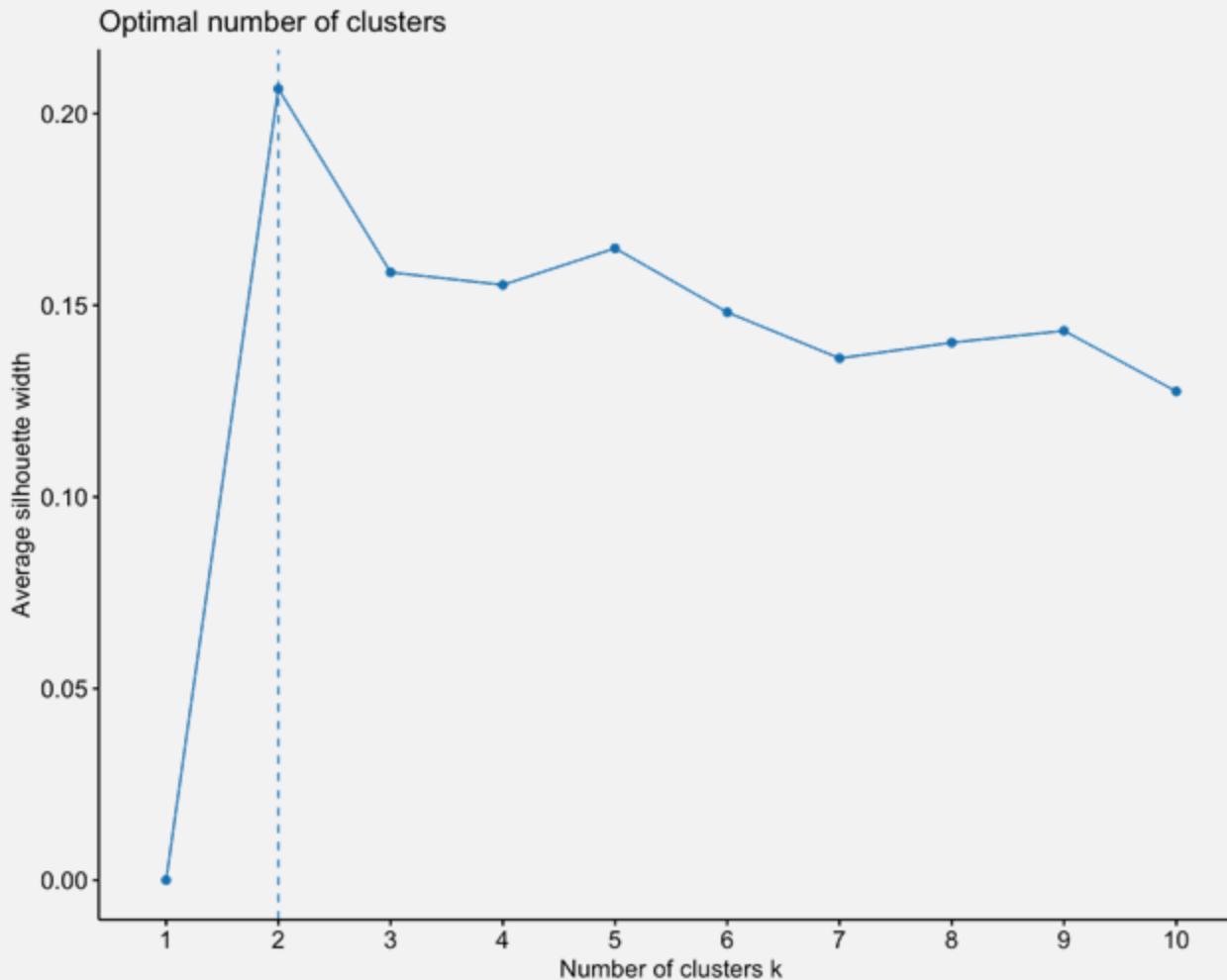
Predictors: all numeric/continuous variables



# Clustering

## K-means

Predictors: all numeric/continuous variables



Cluster	age	cigsPerDay	totChol	sysBP	diaBP	heartRate	glucose
1	1	55.2	6.04	258.	151.	91.8	78.7
2	2	45.7	11.0	223.	120.	76.9	73.7