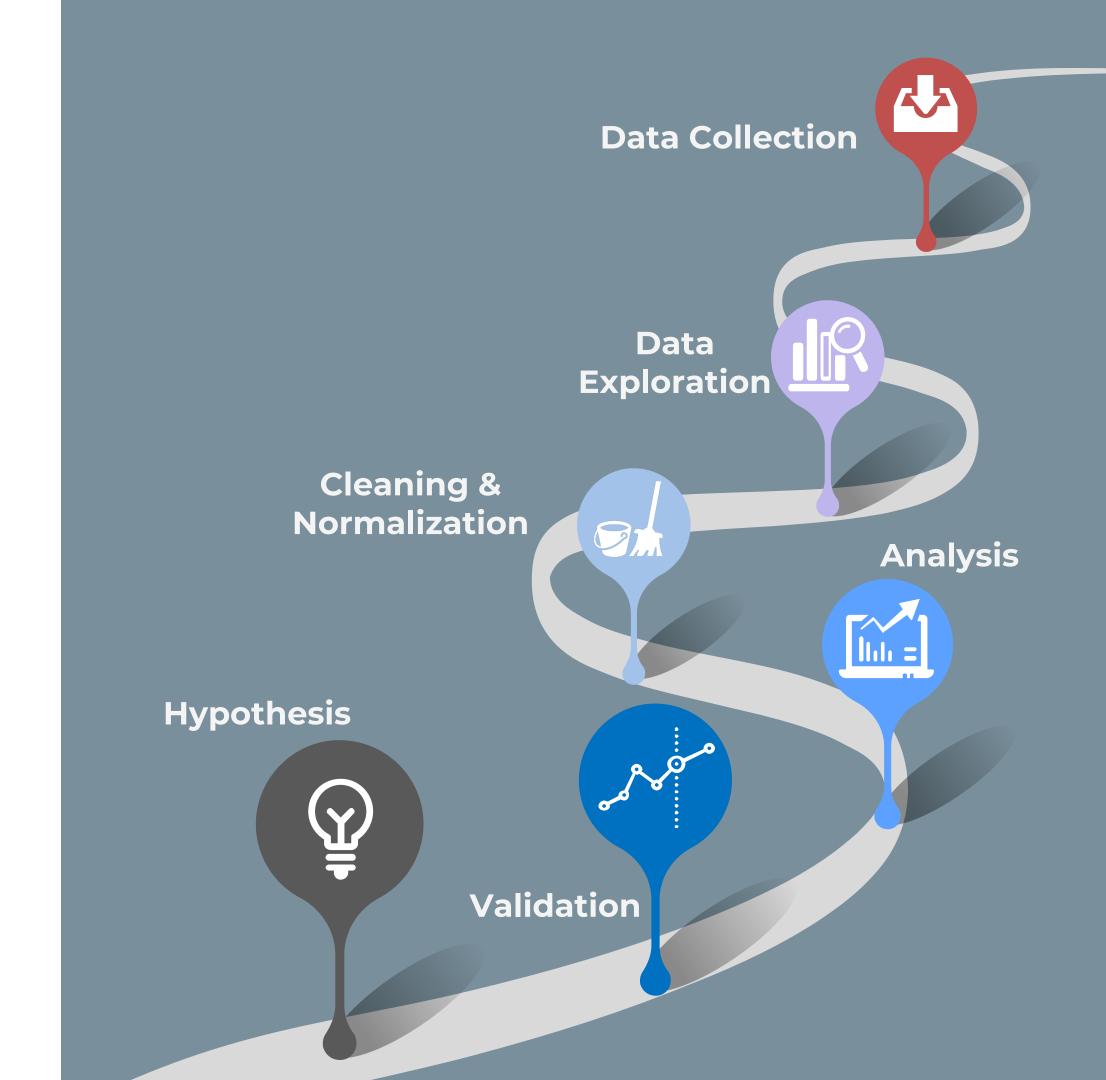
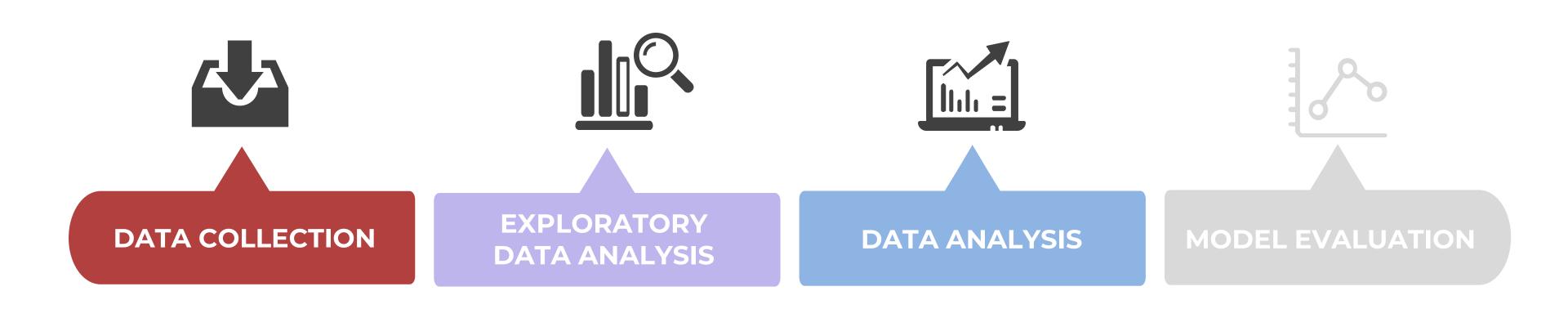
THE DATA'S JOURNEY



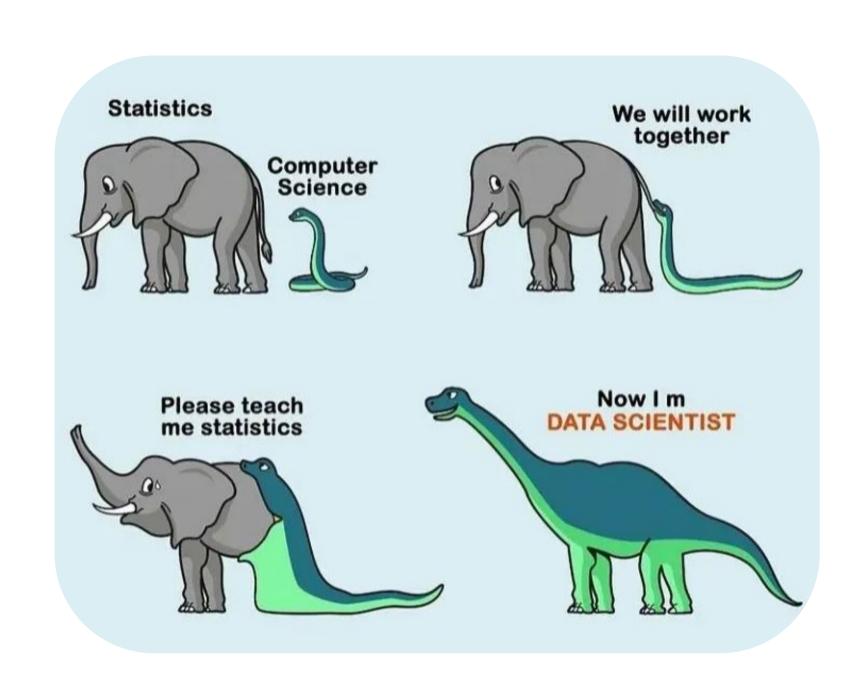


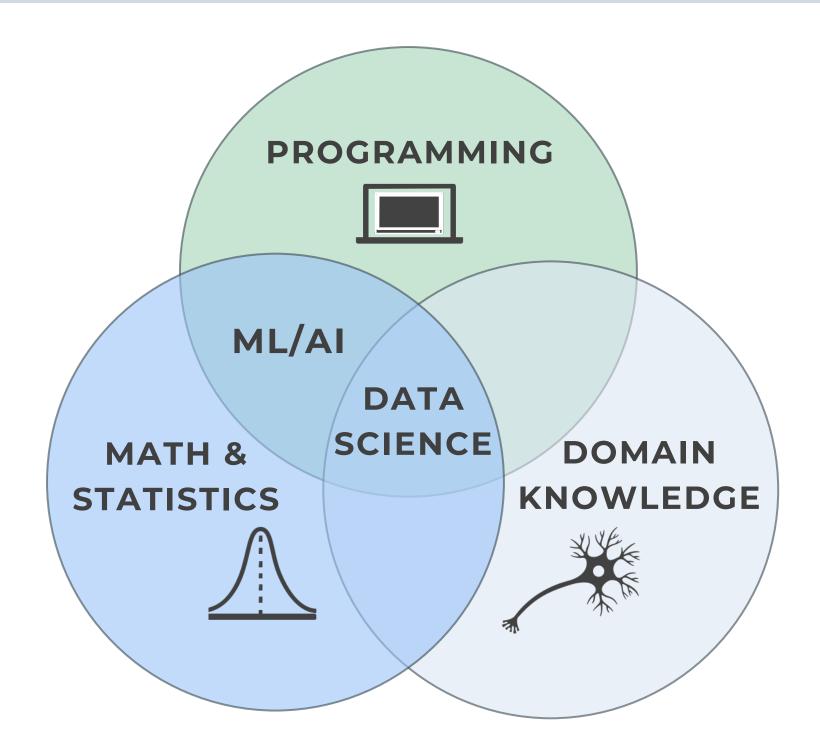
CONTINUING OUR JOURNEY





WHAT IS DATA SCIENCE?







DATA SCIENCE ANALYSIS

HYPOTHESIS TESTING

RQ: Is there a significant difference in the variable of interest between two or more groups?

PREDICTION/ CLASSIFICATION



RQ: Does this patient have a cataract?

RQ: Can we estimate cancer risk based on genetic risk, smoking and age?

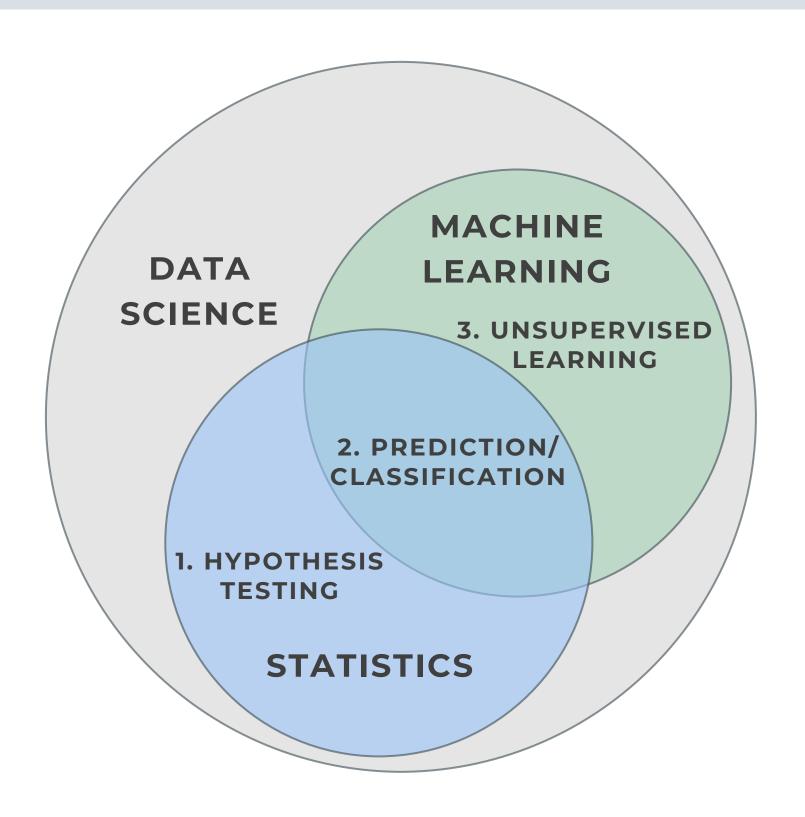
UNSUPERVISED LEARNING



RQ: What sets the cancer cells apart from the healthy cells? Are there subtypes within one cell type?



DATA SCIENCE ANALYSIS

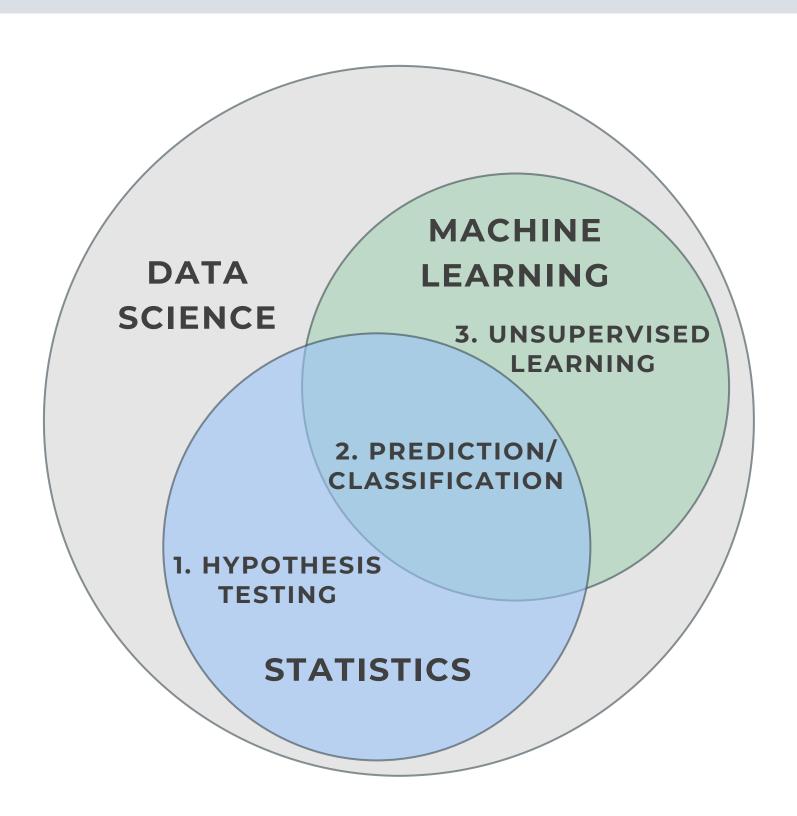


STATISTICS:

- Focus is on inference*, achieved through fitting of a probability model
- Quantitative measure of confidence of a 'true' effect (confidence interval)
- We have data/model assumptions which we must verify
- Statistical models can do predictions, but predictive accuracy is not their strength[1]. BUT interpretable.



DATA SCIENCE ANALYSIS



MACHINE LEARNING:

- Purpose is prediction/classification, finding patterns in large data
- Makes minimal assumptions about the data
- When the number of input variables exceed observation
- Appropriate for complicated nonlinear interactions
- Provide a degree of interpretability, but tends to sacrifice interpretability for predictive power[1].

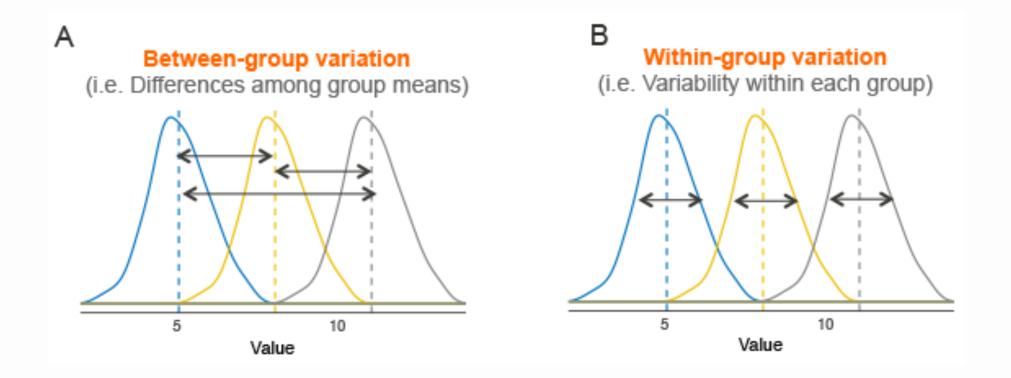


HYPOTHESIS TESTING

Questions that boil down to:

Is there a difference in feature A between these two or more groups?

- Is the bacterial load higher in colon swaps of cancer patients?
- Is the expression of gene A higher in tumor samples?
- Is there a difference in median height between men and women?



A difference test:

- t-test
- ANOVA
- Fisher's exact test



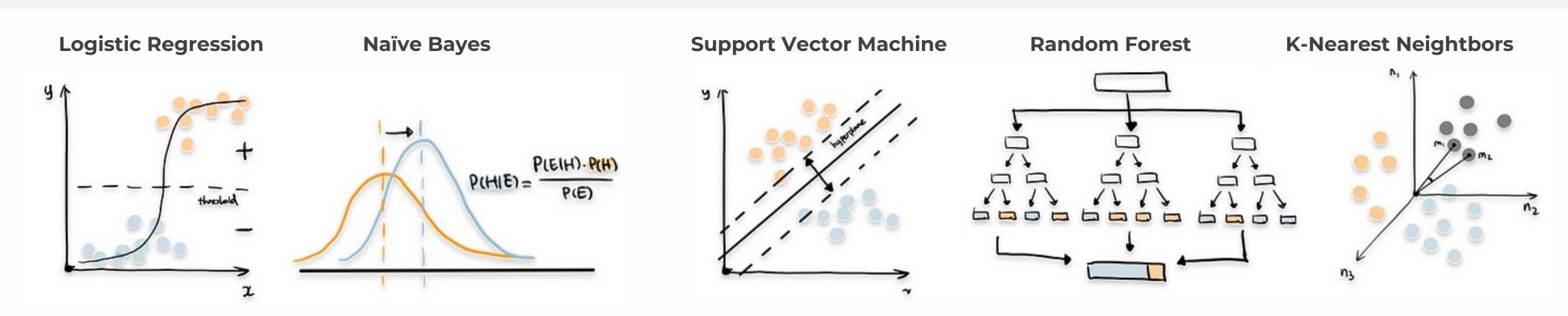
PREDICTIVE MODEL / CLASSIFICATION

Using statistical/ML models to predict outcomes / classify new data.

Discovering and quantifying the relationships between predictor variables and outcome.

Prediction (Regression) == outcome is continuous (weight of newborn)

Classification == outcome is a class/group (cystic fibrosis or healthy)



CLASSIFICATION OF FRUIT - EXAMPLE

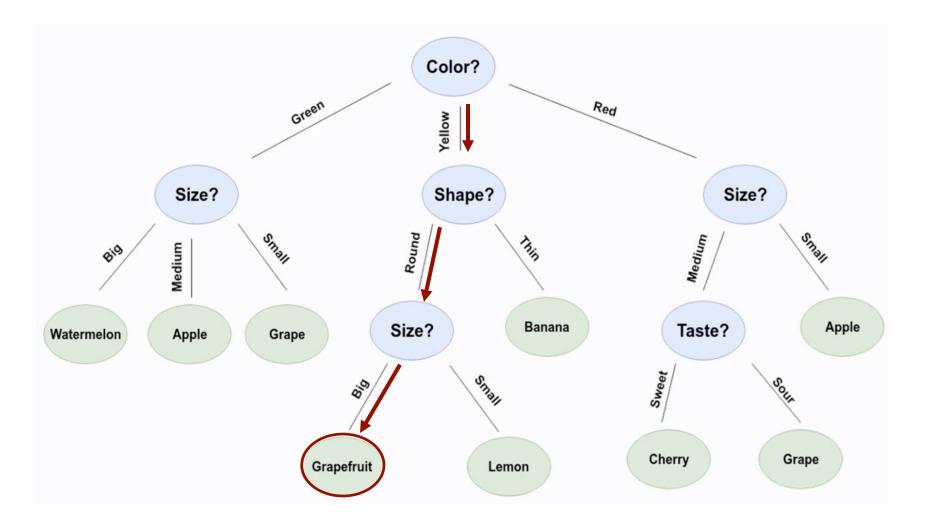
A model trained and evaluated using class labels is used to **classify** a new data point (label unknown).

- Probability (0.68 Grapefruit & 0.32 Lemon)
- Response (Grapefruit)

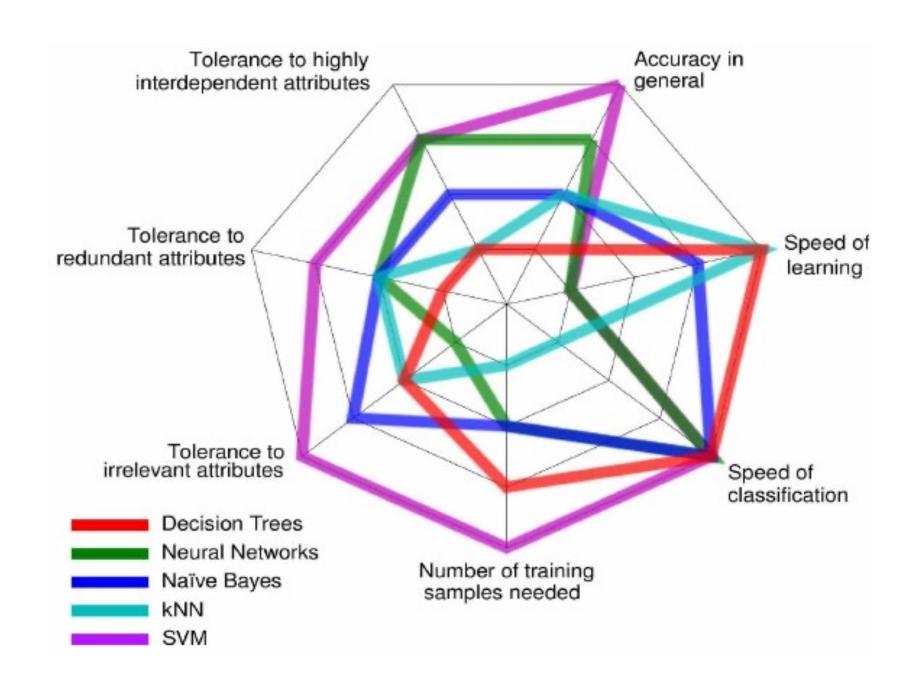
Some models have **feature importance** ranking.

What feature/variable is most important for classification, i.e. size, shape or color?

Decision Tree - Fruit Classification



CLASSIFICATION ALGORITHMS – PROS & CONS



- Pros and cons of different classification methods
- Depending on the scientific question and the data type/size a specific method may be favorable
- Here is a plot comparing algorithms, it is called a spider / radar plot

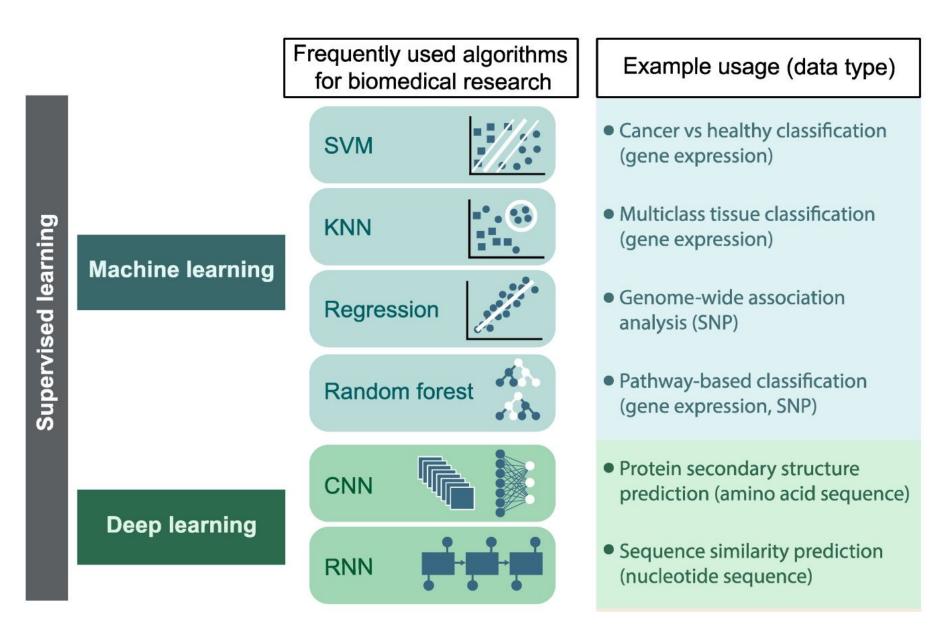


SUPERVISED LEARNING

Classification: Fit/train a model which can assign a new observation to a given class (i.e. cancer vs healthy).

Classification problems are solved using supervised or unsupervised learning

Supervised Classification: Our model is trained using labeled observations - we know the 'ground truth'.





In your groups discuss:

- What is a research question within your field(s) where you would use predictive modelling for classification?
- What model do/could you use? (see previous slides)
- If you had no 'ground truth' (no labels) to use for model training, do you think you could solve a classification problem regardless? If so, what would the scientific question be?



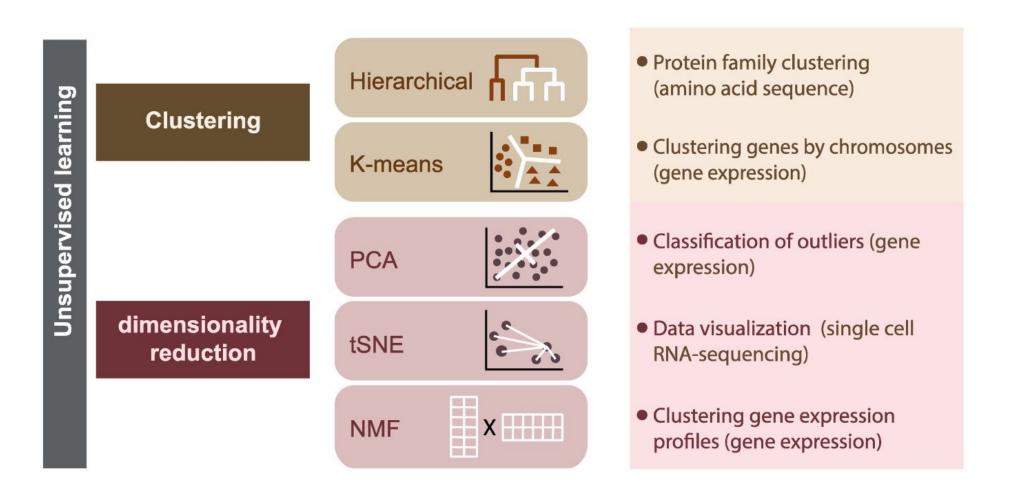
UNSUPERVISED LEARNING

What if we do not know the groups our data partition into, no labels...

A scientific question could be:

Do our observations stratify into groups and what data characteristics drive this partitioning?

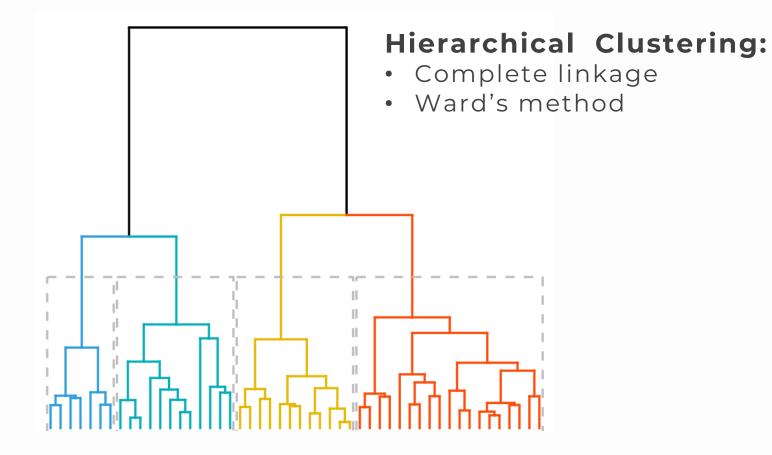
For this we use **unsupervised learning** methods (PCA is one example).





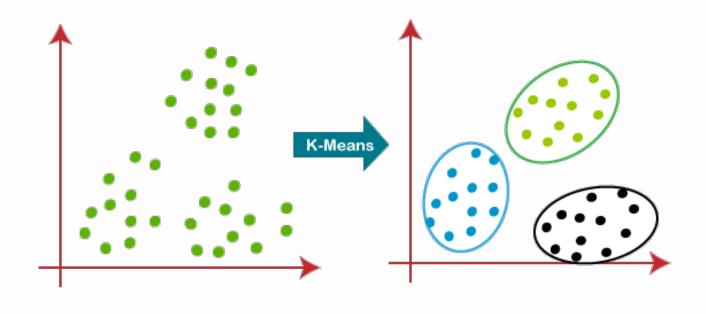
CLUSTERING

- Clustering == type of unsupervised learning.
- Metric to define similarity of observations.
- Items of a group share features that we care about.
- · RQ:
 - Optimal number of clusters for this dataset?
 What characterizes a certain cluster?
 - Which cluster does a new data point belong to?



K-Means Clustering:

- Euclidean distance
- Manhatten distance





DIMENSIONALITY REDUCTION

Dimensionality reduction (DR) methods project data from a high dimensional space into a low dimensional space.

DR methods are unsupervised.

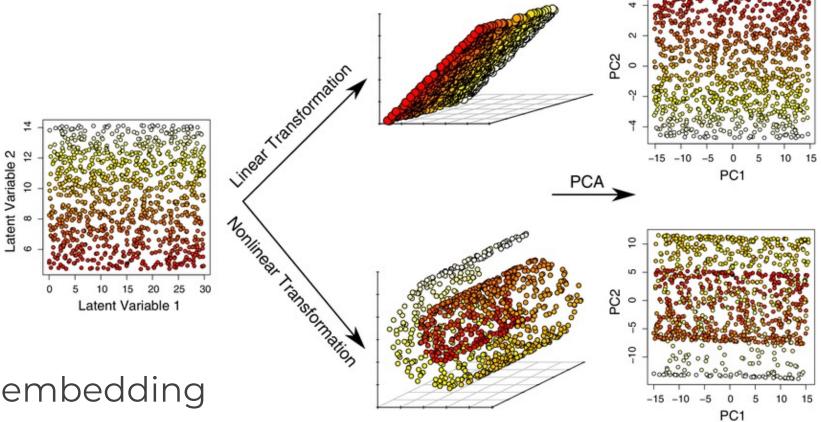
Linear projection:

PCA = principal component analysis

Non-linear projection:

t-sne = t-distributed stochastic neighbor embedding

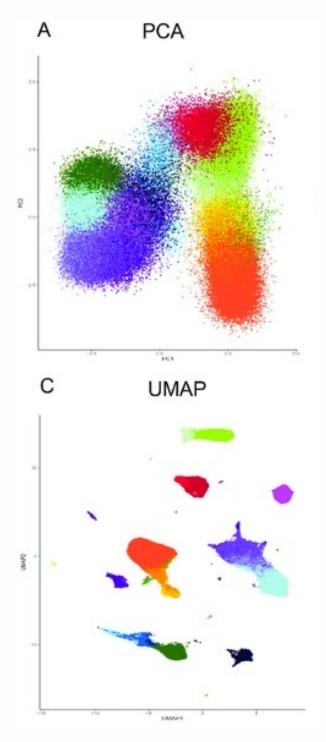
UMAP = Uniform Manifold Approximation and Projection





VISUAL CLUSTERING BASED ON DR

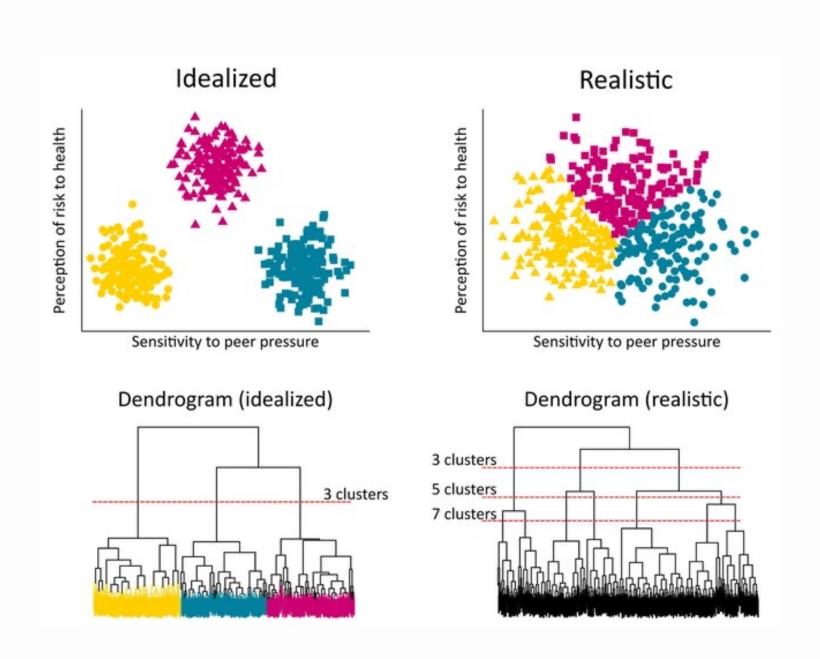
- DR is not a method for clustering, **BUT** it is used in this way.
- Clustering based on DR == visual inspection of plot to identify clusters.
- Coloring observations by meta-information (i.e. cancer grade, drug treatment or cell type,...) *may* support ground truth.
- DR methods can have parameters which should be specified correctly as they will affect results.





CHALLENGES

- How to define similarity?
- Human tendency to see clusters in randomness (clustering illusion)
- Algorithm may require the user to specify the number of clusters (chicken-and-egg problem)
- Non-linear clustering algorithms are powerful but can have unexpected behaviors (parameters)

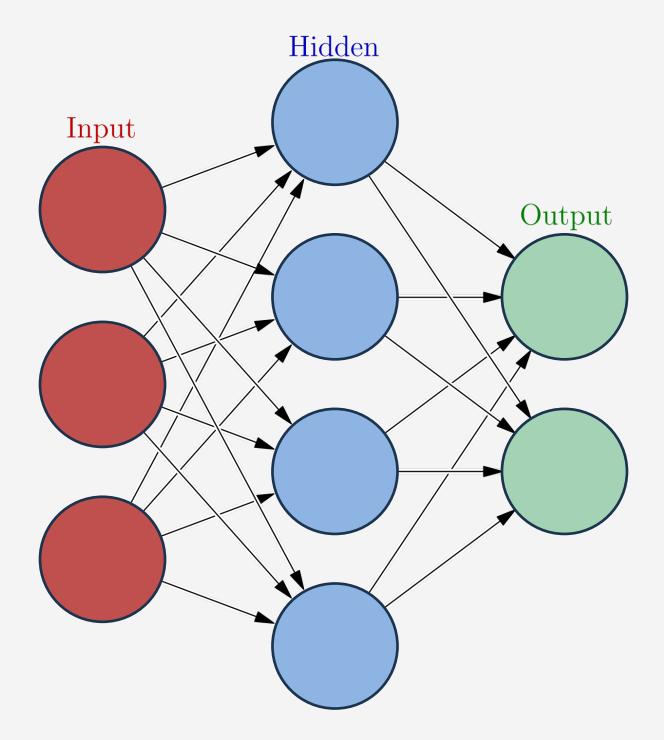


Engl, Elisabeth, Peter Smittenaar, and Sema K. Sgaier. *Gates Open Research* 3 (2019).



NEURAL NETWORKS

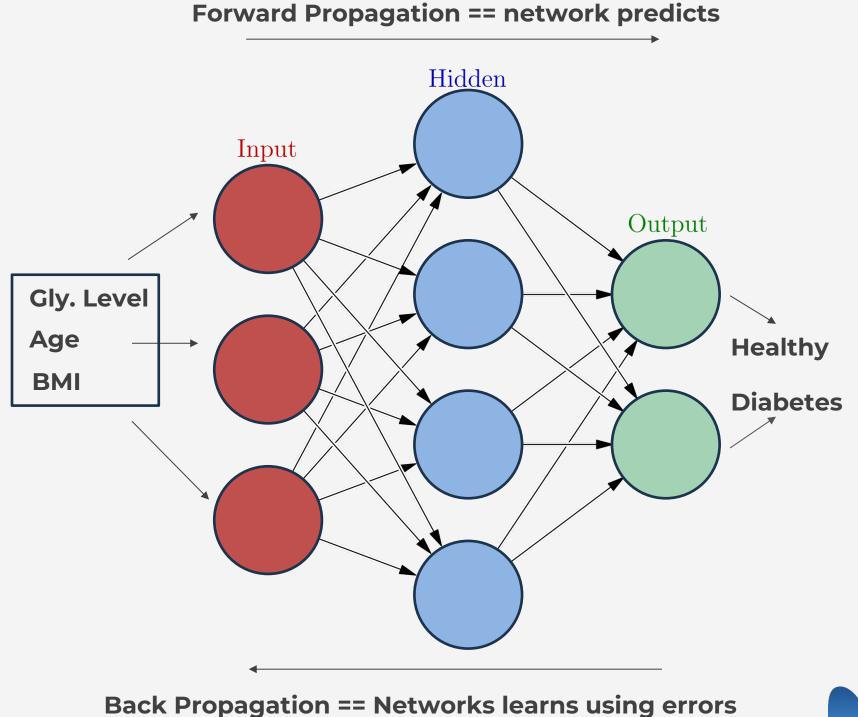
- Neural Networks (NN) mimic the function and structure of the human brain
- NN can be either supervised or unsupervised
- We use NN for medical image analysis, genomics analysis, protein structure prediction, etc.
- NN can be difficult to understand in detail but not in broad strokes, let's try...





NEURAL NETWORK FOR CLASSIFICATION

- Input layer: Data goes in (i.e. patient biometrics, expression data, medical images)
- Hidden layer: The model learns data patterns.
 - This is where the computation is performed.
 - Many hidden layers == deep neural network
- Output layer: Samples are classified as either cancer or healthy.





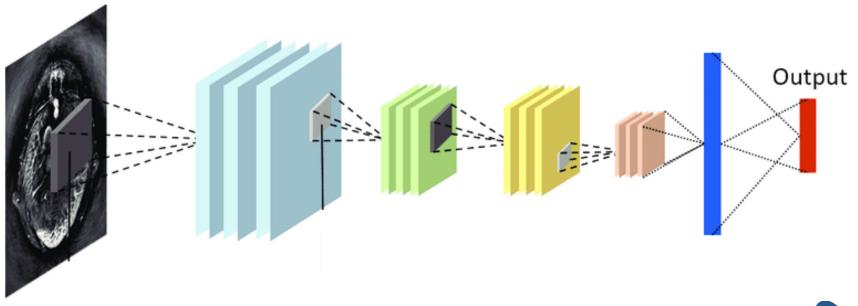
MEDICAL IMAGE ANALYSIS



- There are different types of neural networks (NNs)
- Different neural networks are good for different data and tasks
- NNs for **medical image** analysis are often convolutional neural networks (CNNs).

- The many Data Scientists do NOT make their own NNs, but they use them...
- Many implementations of NNs in different programming languages.

CONVOLUTIONAL NEURAL NETWORK

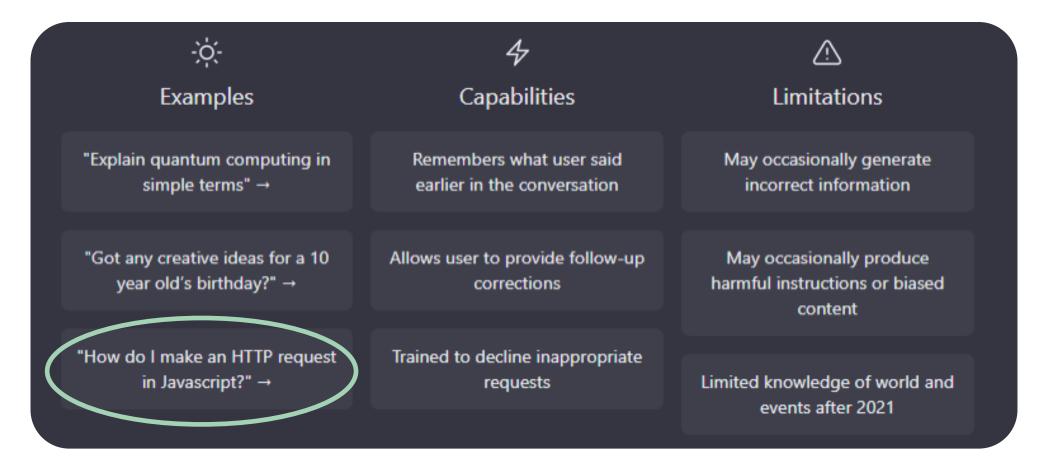


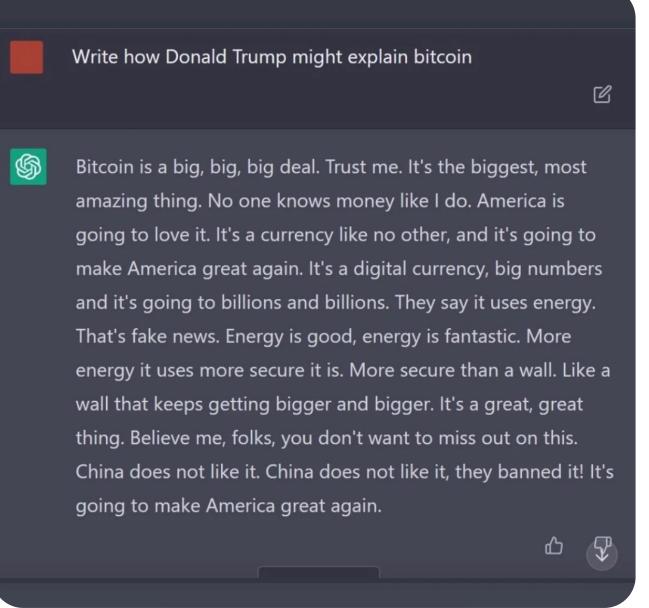
Yang, Changchun, et al. "Review of deep learning for photoacoustic imaging." *Photoacoustics* 21 (2021): 100215.





- You may already have used or heard about ChatGPT
- The algorithm behind is a *Transformer Neural Network*
- This AI bot can do many things, also programming!









Have you used any of the models we've talked about in this section?





Usage of ML Algorithms across Different Data Professional Roles

Job Title ■ Data Scientist ■ Research Scientist DBA/Database Engineer ■ Machine Learning Engineer Linear or Logistic Regression ■ Product/Project Manager Business Analyst **Evolutionary Approaches** Decision Trees or Random Forests Generative Adversarial Networks Convolutional Neural Networks Transformer Networks (BERT, gpt-3, etc) Gradient Boosting Machines (xgboost, lightgbm, etc) Dense Neural Networks (MLPs, etc) Bayesian Approaches Recurrent Neural Networks



WHICH APROACH SHOULD WE USE?

STATISTICAL ANALYSIS

- 'Simple' question
- Linear relationships
- Small dataset (few obs.)

The scientific questions asked are central to the methods chosen.

The **dataset size** guides choice of algorithm.

MACHINE LEARNING

- High dimensional data
- Non-linear relationships

SUPERVISED

Outcome is known

UNSUPERVISED

Outcome is unknown Fully data driven



In your groups discuss which of the three areas of DS analysis we talked about best applies to this question.

Is there a difference in weight between mice of strain A and strain B?

Further discuss:

What types of variables (data types) do we have? Do you know of any test you could use to answer this question?



In your groups discuss which of the three areas of DS analysis we talked about best applies to this question.

You have protein abundance data from skin samples (~ 10.000 different protein species). These samples were collected from patients with psoriasis (normal adjacent -and affected skin) and from healthy controls.

Is protein abundance predictive of the skin phenotype? And if so, are the levels of all proteins equally predictive/informative?



In your groups discuss which of the three areas of DS analysis we talked about best applies to this question.

Is the amount of bacterial load in a swap of the oral epithelium (gums) based on skin type, diet and whether the person recently had antibiotic treatment?

Further discuss:

What are the outcome variable(s) and the explanatory variable(s) in this scenario?





In your groups discuss which of the three areas of DS analysis we talked about best applies to this question.

Gene expression data and biometrics (height, weight, age, etc.) from patients with colorectal cancer. You are interested in exploring if there are any potential subgroups of cancer patients within your dataset, in order to pair each subgroup with the appropriate healthy controls.

What type of analysis could you use for this? N.B while avoiding a fishing-expedition





A BREAK IN OUR JOURNEY



20:00



Breaktime for PowerPoint by Flow Simulation Ltd.

Pin controls when stopped <a>

