Supervised vs. unsupervised learning

Protein sequence space



Supervised classification problem

- 1. Training data (we need examples to learn from)
- 2. Learning (how to learn from examples)
- 3. Validation (how to find trade off between accuracy and generalization)

Training data:

• A collection of records (objects) **x**. Each record contains a set of features and the class *C* that it belongs to.

Age	tumor-size	inv-nodes	irradiat	Class
21	23	12	yes	recurrence-events
32	12	3	yes	no-recurrence-events
10	3	2	no	no-recurrence-events
45	3	6	yes	recurrence-events
				1 77
		$\{\mathbf{X}_i, \mathbf{C}_i\}$	} l	$= 1 \dots N$

How to choose feature space?

adults kids weights estrogen " heights testosterón

Learning:

• Find a model *y*(**x**;**w**) that describes the objects of each class as a function of the features and adaptive parameters (weights) w.



What is the best model: accuracy vs. generalization

 Find a model y(x;w) that avoids overfitting – too high accuracy on the training set may result in poor generalization (classification accuracy on new instances of the data).



Algorithms for supervised learning

- LDA/FDA (Linear/Fisher Discriminate Analysis) (simple linear cuts, kernel non-linear generalizations)
- SVM (Support Vector Machines) (optimal, wide margin linear cuts, kernel non-linear generalizations)
- Decision trees (logical rules)
- k-NN (k-Nearest Neighbors) (simple nonparametric)
- Neural networks (general non-linear models, adaptivity, "artificial brain")

K-means clustering for unsupervised pattern discovery

- Choose the number of clusters (k), choose randomly their centers.
- Compute the mean (or median) vector for all items in each cluster.
- Reassign items to the cluster whose center is closest to the item, iterate the above two steps.
- Problems: spherical clusters, low "noise" tolerance, local minima:



k-NN vs. k-means

or supervised vs. unsupervised pattern discovery.

Critical decisions to be made:

- Do I want to utilize a prior knowledge e.g. about functionally related genes? Yes -> k-NN, No -> k-means
- 2. What similarity measure (metric) to choose?
- 3. Which *k* is best? Try different *k*'s and compare the results!
- 4. When to stop optimization *(k-means)*? Try re-running *k*-means several times!
- 5. Are the results significant? Use cross validation and biological knowledge!