GENIOMHE

Data-mining and Machine Learning

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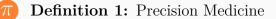
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1 Unsupervised Learning



Design of treatment for a given patient, based on genomic data.

Definition 2: Hierarchical clustering

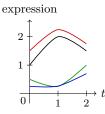
Gene expression time series: look for genes with similar expression footprint.

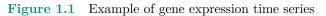
Representation of data

- Tables;
- Trees / Graphs;
- Time series...

1.1 Distances and Similarities

Property 1 (Distance). *non-negativity* $d(i, j) \ge 0$





1 Unsupervised Learning

isolation d(i,i) = 0symmetry d(i,j) = d(j,i)triangular inequality $d(i,j) \le d(i,h) + d(h,j)$

Definition 3: Dissimilarity

Distance without triangular inequality.

Definition 4: Similarity

Function s from $X \times X$ to \mathbb{R}_+ such that:

- 1. s is symmetric: $(x, y) \in X \times X$; s(x, y) = s(y, x)
- 2. $(x, y) \in X \times X; s(x, x) = s(y, y) > s(x, y).$

Exercise 1:

Let d(x, y) be the distance, $d(x, y) \in [0, +\infty[$. What should be the similarity measure S(x, y) = f(d(x, y)) that satisfies the following property:

 $(x,y) \in X \times X \mid S(x,y) > S(x,y)$

having $S(x,y) \leq M, \ S(x,y) \in]0,M]$. $d(x,y) \geq 0 \ \forall (x,y)$

$$S(x,y) = \frac{M}{d(x,y) + 1}$$
(1.1)

In eq. (1.1), S(x, y) ranges from 0 to M.

$$\lim_{n \to \infty} \frac{M}{n+1} = 0 \qquad \lim_{n \to 0} \frac{M}{n+1} = M$$
(1.2)

1.2 Data Representation

Data matrix

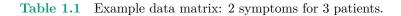
Distance matrix

 $\begin{bmatrix} 0 & & & \\ d(2,1) & 0 & & \\ d(3,1) & d(3,2) & 0 & \\ \vdots & \vdots & \ddots & \\ d(n,1) & d(n,2) & \dots & \dots & 0 \end{bmatrix}$

Definition 5: Minkowski distance

$$L_p(x,y) = (|x_1 - y_1|^p + |x_2 - y_2|^p + \ldots + |x_d - y_d|^p)^{1/p} = \left(\sum_{i=1}^d (x_i - y_i)^p\right)^{1/p}$$

	s_1	s_2
p_1	0	1
p_2	1	0
p_3	3	2



 π Definition 5 continued

where p is a positive integer.

Definition 6: Manhattan distance

$$L_1(x,y) = \sum_{i=1}^{d} |x_i - y_i|$$

Definition 7: Euclidian distance

Let A and B be two points, with (x_A, y_A) and (x_B, y_B) their respective coordinates,

If $p = 2, L_2$ is the Euclidian distance:

 π **Definition 8:** Euclidian distance

 $d(x,y) = \sqrt{|x_1 - y_1|^2 + |x_2 - y_2| + \ldots + |x_d - y_d|^2}$

We can add weights

1.2.1 K-means

The cost function is minimized:

$$Cost(C)\sum_{i=1}^k \dots$$

Algorithm 1 K-means algorithm

Choose the number of clusters k.

Choose randomly k means.

For each point, compute the distance between the point and each means. We allocate the point to the cluster represented by the clostest center.

We set each means to the center of the cluster, and reiterate.

Exercise 2: We have six genes:



Figure 1.2 k-means states at each of the 3 steps

	g_1	g_2	g_3	g_4	g_5	g_6
$\times 10^{-2}$	10	12	9	15	17	18

Table 1.2Sample values for six gene expressions.

With k = 2 and $m_1 = 10 \cdot 10^{-2}$ and $m_2 = 9 \cdot 10^{-2}$ the two initial randomly chosen means, run the k-means algorithm.