

Visualization

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1 DISTANCES

Distances in two dimensions

Two points M_1 and M_2 in the 2D space with coordinates (x_1, y_1) , and (x_2, y_2) , respectively.

L_2

$$d(M_1, M_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2} \quad (1)$$

L_1

$$d(M_1, M_2) = |x_1 - x_2| + |y_1 - y_2| \quad (2)$$

L_∞

$$d(M_1, M_2) = \max(|x_1 - x_2|, |y_1 - y_2|) \quad (3)$$

weighted L_1

$$d(M_1, M_2) = \alpha_1 |x_1 - x_2| + \alpha_2 |y_1 - y_2| \quad (4)$$

Distances in three dimensions

Two points M_1 and M_2 in the 3D space with coordinates (x_1, y_1, z_1) , and (x_2, y_2, z_2) , respectively.

L_2

$$d(M_1, M_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2} \quad (5)$$

L_1

$$d(M_1, M_2) = |x_1 - x_2| + |y_1 - y_2| + |z_1 - z_2| \quad (6)$$

L_∞

$$d(M_1, M_2) = \max(|x_1 - x_2|, |y_1 - y_2|, |z_1 - z_2|) \quad (7)$$

weighted L_1

$$d(M_1, M_2) = \alpha_1 |x_1 - x_2| + \alpha_2 |y_1 - y_2| + \alpha_3 |z_1 - z_2| \quad (8)$$

2 LIKELIHOOD

- Observations : (x_1, \dots, x_n)
- Model : p (for instance a normal law)
- Parameters : θ (for instance (μ, σ) , the mean and the standard deviation of the normal law)

$$L(\theta) = p(x_1, \dots, x_n | \theta) \quad (9)$$

2.1 Exercise 5

Here p is a **parameter** (that of a Bernoulli law), not to be mixed with the letter p that stands for a **model** of a probability distribution.

$$L(p) = p \times (1 - p) \quad (10)$$

$$L(p) = p - p^2 \quad (11)$$

$$L'(p) = 1 - 2p \quad (12)$$

$L'(p) = 0$ if and only if $p = \frac{1}{2}$

Studying the variations of L as a function of p , we see that the maximum is at $p = \frac{1}{2}$.

3 DERIVATIVE

$$f : x \rightarrow f(x)$$

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} \quad (13)$$

$$g : x \rightarrow 3x$$

$$\forall x \in \mathbb{R}, g'(x) = 3$$

$$h : x \rightarrow x^2$$

$$h' = ?$$

4 EXPECTED VALUE

X constant random variable : $X = \alpha$

$$\sum_{i=1}^n p_i x_i = \sum_{i=1}^n p_i \alpha = \alpha \sum_{i=1}^n p_i \quad (14)$$

5 K-MEANS

- Datapoints (x_1, \dots, x_n)
- Centroids (c_1, \dots, c_n) (one centroid per point, however the number of different centroids is smaller than the number of datapoints)

The inertia I is given by :

$$I = \sum_{i=1}^n d(x_i, c_i)^2 \quad (15)$$

6 ENTROPY

Entropy of certain distribution.

$$H = 0 \quad (16)$$

Entropy of uniform distribution with n values :

$$\begin{aligned} H &= - \sum_{i=1}^n \frac{1}{n} \log \frac{1}{n} \\ &= -n \times \frac{1}{n} \times \log \frac{1}{n} \\ &= \log n \end{aligned} \quad (17)$$

7 COMPLEXITY

Let n be the size of the problem.

Polynomial complexity :

$$a_k n^k + a_{k-1} n^{k-1} + \dots + n \quad (18)$$

Exponential complexity :

$$k^n \quad (19)$$

avec $k > 1$

8 EXERCISE 2 (TIDE LEVEL)

Hypothèse :

L : tide level in meters

t : time in hours

A : amplitude

ϕ : phase

f : frequency ($\text{Hz} = \text{s}^{-1}$)

$\lambda = \frac{1}{f}$: periode en secondes

$\omega = 2\pi f$: pulsation (radian par seconde)

$$L = A \sin(\omega t + \phi) + c \quad (20)$$

error E :

$$\sum_{\text{samples}} (\text{prediction} - \text{truth})^2 \quad (21)$$