

# Lecture notes - Clustering and Persistence SF2704

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## 1 introduction

The big goal for this lecture series is to understand metrics by searching for a metric between metric spaces. This has been proven to be impossible (ref), but despite the lack of a grand metric of metrics we can still try to search within (common word for ultrametrics, barcodes etc) to at least get a grasp on certain parts.

## 2 notes

### note 1

Skeleton for note 1

### note 2

*Metric space*  $(X, d)$  where  $X$  is a finite set and  $d$  is a distance between the points in this finite set.

$$|X| < \infty \quad (1)$$

$$d(x, x) = 0, d(x, y) = d(y, x), d(x, y) + d(y, z) \leq d(x, z) \forall x, y, z \in X \quad (2)$$

An interesting submetric is the *ultrametric* which has a strong triangle inequality constraint

$$\max\{d(x, y), d(y, z)\} \leq d(x, z) \forall x, y, z \in X \quad (3)$$

also called ultrametric inequality.

Stated in another way, for an ultrametric any 3 points, i.e. a triangle, will have the following property

<image of triangle with the sides (a, a, b <= a)>

$P(X)$  is the set of partitions of  $X$ . A partition of  $X$

$$\sigma = \{u_i\} = \bigsqcup_i u_i \in P(X) \quad (4)$$

where  $u_i \in \sigma$  is called a block.

To be able to order partitionings we define

$$\sigma \leq \tau \Leftrightarrow \forall u \in \sigma \exists v \in \tau : u \subset v \quad (5)$$

A *clustering* is just a function  $\Psi$  in the form of a algorithm or procedure which maps a metric space  $(X, d)$  to a partition of  $X$

$$\Psi : (X, d) \rightarrow P(X) \quad (6)$$

The function  $\Psi(X, d) \in P(X)$  can have 3 important properties

**Scale invariant** Changing the scale for the distance does not change the partitioning

$$\Psi(X, d) = \Psi(X, \alpha d) \forall \alpha \in \mathbb{R}_{++} \quad (7)$$

**Rich**  $\Psi(X, d) \rightarrow P(X)$  i.e. surjective or onto the set of partitions of  $X$ .

$$\forall \sigma \in P(X) \exists d : \Psi(X, d) = \sigma \quad (8)$$

**Consistent**  $d'$  is such that you decrease the intrablock distance and increase the extrablock distance for all blocks. This will correspond to making the clusters more distinct. Let  $x \sim_{\Psi(X, d)} y$  denote that  $x$  and  $y$  belongs to the same block of the clustering  $\Psi(X, d)$ .

$$d'(x, y) : \begin{cases} \leq d(x, y) & x \sim_{\Psi(X, d)} y \\ \geq d(x, y) & x \not\sim_{\Psi(X, d)} y \end{cases} \quad (9)$$

(not in notes; but it should be that  $\Psi(X, d) = \Psi(X, d') \forall d'$  fullfilling the above property)

## note 9

Skeleton for note 9.

## 3 exercises

### exercise 1

## 4 code