

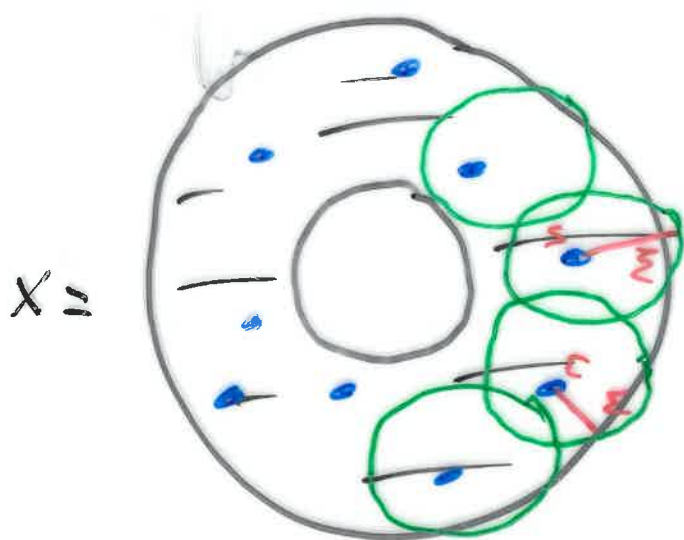
Suppose  $X$  is an unknown metric space. Suppose given a finite sample  $S \subset X$  and the distance

$$d_X(x, y)$$

for all  $x, y \in S$ .

Problem We'd like to infer something about the topology of  $X$  using the distance  $d_X(x, y)$  for  $x, y \in S$ .

Example A sample  $S$  is taken from an unknown metric space  $X \subseteq \mathbb{R}^2$ .



$S$  finite sample.

know

$$d_X(x, y), x, y \in S$$

We could construct the ball

$$B(s, \varepsilon) = \{x \in X : d_X(x, s) < \varepsilon\}.$$

We could hope that the collection

$$\mathcal{U}_\varepsilon = \{B(s, \varepsilon)\}_{s \in X}$$

might be an open cover of  $X$   
for suitable  $\varepsilon > 0$ .

If  $\mathcal{U}_\varepsilon$  is an open cover of  $X$   
then Leray's theorem says  
that the nerve  $N\mathcal{U}_\varepsilon$  is  
homotopy equivalent to  $X$ .

The hypothesis of Leray's  
theorem is satisfied because  
a ball is contractible, and  
so is any intersection of balls.

N.B. This means that  $N\mathcal{U}_\varepsilon$  has the same Euler characteristic as  $X$ .

However, we can't construct the balls as we don't know  $X$ . Also, even if we could construct the balls, it would be expensive to compute their intersections.

Recall:  $N\mathcal{U}_\varepsilon$  is a simplicial complex with one vertex  $s$  for each ball  $B(s, \varepsilon)$ ,  $s \in S$ .

$N\mathcal{U}_\varepsilon$  has an edge



if  $B(s, \varepsilon) \cap B(t, \varepsilon) \neq \emptyset$ .

This will be non-empty if and only if  $d_X(s, t) < 2\varepsilon$ .

So we can construct the vertices and edges of  $N_{2\varepsilon}$  just from the distances between points in  $S$ .

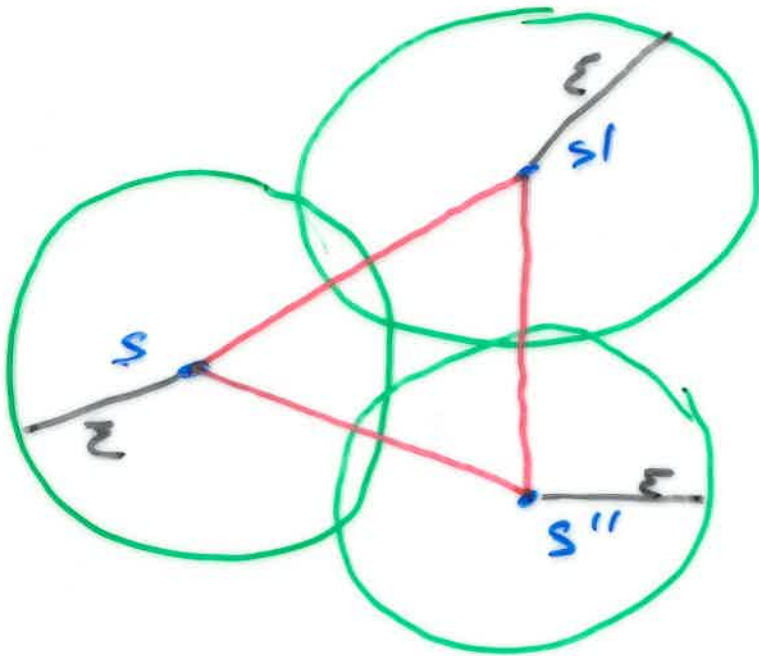
$N_{2\varepsilon}$  has a 2-simplex for

$$B(s', \varepsilon) \cap B(s'', \varepsilon) \cap B(s''', \varepsilon) \neq \emptyset$$

$s, s', s'' \in S$ .

If we happened to know  $X$  we could compute these balls and their intersections. However, this is a length computation,

We could approximate  $N\mathcal{U}_\varepsilon$  by considering the picture:



We could approximate the nerve  $N\mathcal{U}_\varepsilon$  by the clique complex

$K_{2\varepsilon} :=$

$K_{2\varepsilon}$  has a  $k$ -simplex

$$\sigma = \{s_0, s_1, \dots, s_k\}$$

whenever

$$d(s_i, s_j) < 2\varepsilon \text{ for all } s_i, s_j \in \sigma.$$

## Proposition

$$NU_{\Sigma} \subseteq K_{2\Sigma} \subseteq NU_{2\Sigma}$$

Conclusion:

$K_{\Sigma}$  is a good approximation  
to  $NU_{\Sigma}$  since we consider  
a range of  $\Sigma > 0$ .

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The above ideas combine  
to form the Mapper  
Clustering algorithm.