

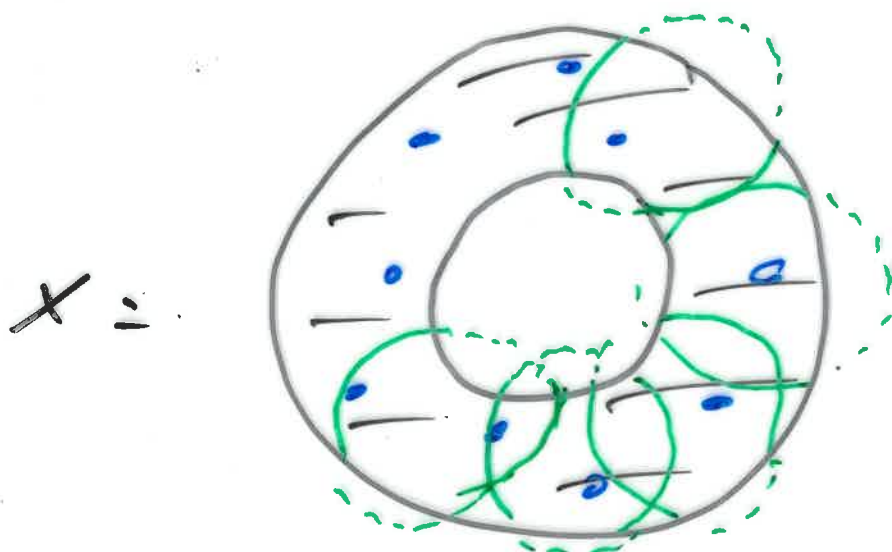
Suppose  $X$  is an unknown metric space. Suppose given a finite sample  $S \subset X$  and the distance

$$d_X(x, y)$$

for all  $x, y \in S$ .

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

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



$S$  finite  
sample

If 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

the nerve  $N\mathcal{U}_\varepsilon$  is homotopy

equivalent to  $X$  (so has the

same Euler characteristic etc.  
as  $X$ ).

However, we can't construct the balls  $B(s, \varepsilon)$  as we don't know  $X$ .

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  $d_X(s, t) < 2\varepsilon$ .

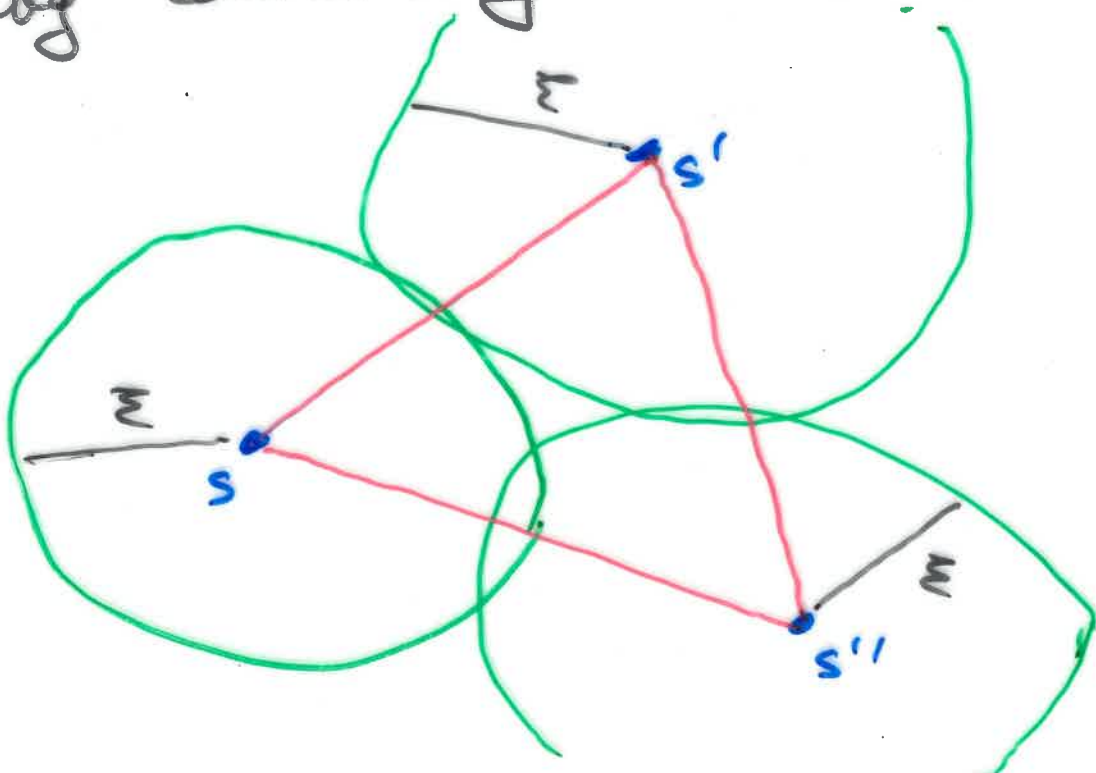
So we can construct the vertices and edges of  $N\mathcal{U}_\varepsilon$ .

$N\mathcal{U}_\varepsilon$  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 happen to know that  $X \subseteq \mathbb{R}^n$  then we could compute these balls and their intersections. However, this might be a lengthy 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

$$N\mathcal{U}_\varepsilon \subseteq K_{2\varepsilon} \subseteq N\mathcal{U}_{2\varepsilon}$$

Conclusion:

$K_\varepsilon$  is a good approximation to  $N\mathcal{U}_\varepsilon$  since we consider a range of  $\varepsilon > 0$ .

## Illustration: Natural Images

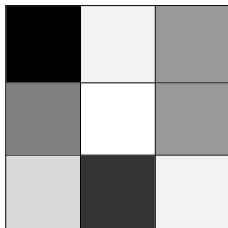
van Hateren & van der Schaaf: 4167 digital photos of random outdoor scenes.



## Illustration: Natural Images

van Hateren & van der Schaaf: 4167 digital photos of random outdoor scenes.

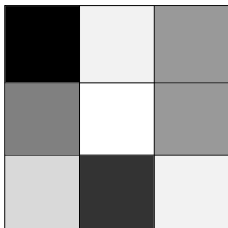
Mumford, Lee, Pedersen: 8 000 000 random high-contrast  $3 \times 3$  patches in  $\mathbb{E}^9$ .



## Illustration: Natural Images

van Hateren & van der Schaaf: 4167 digital photos of random outdoor scenes.

Mumford, Lee, Pedersen: 8 000 000 random high-contrast  $3 \times 3$  patches in  $\mathbb{E}^9$ .



By normalizing with respect to brightness and contrast the patches are projected onto a set  $\mathcal{S}$  of points in a topological seven-sphere  $\mathbb{S}^7 \subset \mathbb{E}^8$ .



Carlsson, Ishkanov, de Silva, Zomorodian: consider high-density subsets  $\mathcal{S}(k) \subset \mathcal{S}$ .

Carlsson, Ishkanov, de Silva, Zomorodian: consider high-density subsets  $\mathcal{S}(k) \subset \mathcal{S}$ .

**Codensity function:**  $\delta_k(x)$  is the Euclidean distance from  $x \in \mathcal{S}$  to its  $k$ th nearest neighbour in  $\mathcal{S}$ .

Carlsson, Ishkanov, de Silva, Zomorodian: consider high-density subsets  $\mathcal{S}(k) \subset \mathcal{S}$ .

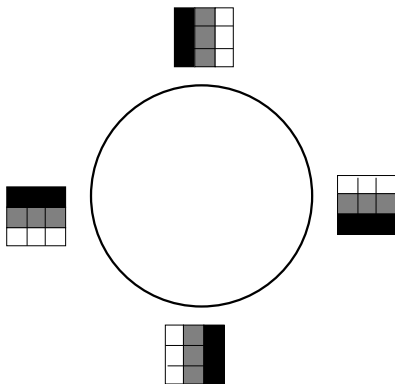
**Codensity function:**  $\delta_k(x)$  is the Euclidean distance from  $x \in \mathcal{S}$  to its  $k$ th nearest neighbour in  $\mathcal{S}$ .

$\mathcal{S}(k)$  contains the 25% of patches with least  $\delta_k(x)$ .

The simplicial complex  $K_\epsilon$  for  $\mathcal{S}(300)$  'look like' a single circle.

The simplicial complex  $K_\epsilon$  for  $\mathcal{S}(300)$  'look like' a single circle.  
(Later we'll be more precise: The  $\beta_1$  bar code for a random sampling of  $\mathcal{S}(300)$  yields a single persistent homology class.)

The simplicial complex  $K_\epsilon$  for  $\mathcal{S}(300)$  'look like' a single circle.  
 (Later we'll be more precise: The  $\beta_1$  bar code for a random sampling of  $\mathcal{S}(300)$  yields a single persistent homology class.)

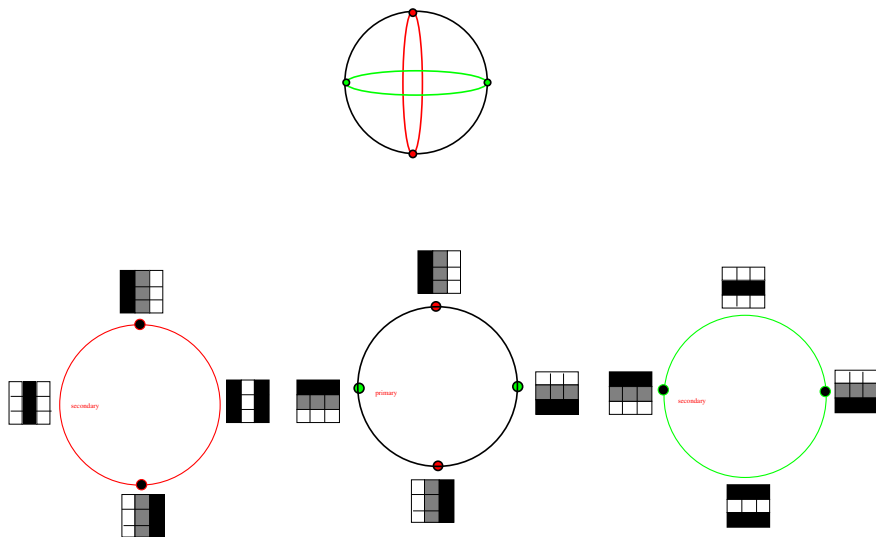


The simplicial complex  $K_\epsilon$  for  $\mathcal{S}(300)$  'comprises' five circles.



The simplicial complex  $K_\epsilon$  for  $\mathcal{S}(300)$  'comprises' five circles.  
(Later we'll be more precise: The  $\beta_1$  bar code for a random sampling of  $\mathcal{S}(15)$  yields five persistent homology classes.)

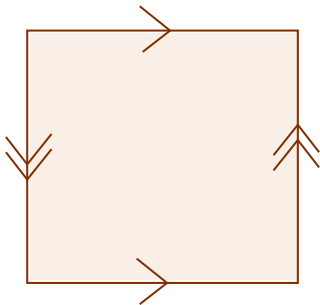
The simplicial complex  $K_\epsilon$  for  $\mathcal{S}(300)$  'comprises' five circles.  
 (Later we'll be more precise: The  $\beta_1$  bar code for a random sampling of  $\mathcal{S}(15)$  yields five persistent homology classes.)



Further inspection suggests that  $K_\epsilon$  'looks like' a Klein bottle.  
(Later we'll be more precise: The  $\beta_2$  bar codes are less robust, but seem to indicate a single homology class with mod 2 coefficients.)

Further inspection suggests that  $K_\epsilon$  'looks like' a Klein bottle.  
(Later we'll be more precise: The  $\beta_2$  bar codes are less robust, but seem to indicate a single homology class with mod 2 coefficients.)

$$H_2(\text{Klein bottle}, \mathbb{Z}_2) = \mathbb{Z}_2$$



Further inspection suggests that  $K_\epsilon$  'looks like' a Klein bottle.  
(Later we'll be more precise: The  $\beta_2$  bar codes are less robust, but seem to indicate a single homology class with mod 2 coefficients.)

$$H_2(\text{Klein bottle}, \mathbb{Z}_2) = \mathbb{Z}_2$$

