# Computational topology: Image classification using Vietoris-Rips complex

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### Obtaining and preprocessing data

```
class Scaler(BaseEstimator, TransformerMixin):
       """ Class used for scaling the dataset x.
           Removes the mean and than scales to values from [-1, 1].
           Implements inverse_transform method.
           axis: if 1, standardize samples (rows)
                 if 0, standardize features (columns)
10
       def __init__(self, axis=1):
           assert axis in [0,1]
           self.axis = axis
       def fit(self, x):
15
           shape = (len(x), 1) if self.axis == 1 else (1, len(x[0]))
           self.means = np.mean(x, self.axis).reshape(shape)
           self.maxs = np.max(np.abs(x - self.means), self.axis).reshape(shape)
       def transform(self, x):
           return np.nan_to_num((x - self.means)/self.maxs)
       def fit_transform(self, x):
           self.fit(x)
           return self.transform(x)
       def inverse_transform(self, x):
           return (x*self.maxs) + self.means
```

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#### The classification model

**Definition 1.** [Vietoris-Rips complex] Let X be a set of m-dimensional points  $X \in \mathbb{R}^m$  and let d be a metric. Pick a parameter r > 0. Construct a simplicial complex as follows:

- Add a 0-simplex for each point in X.
- For  $x_1, x_2 \in X$  add a 1-simplex between  $x_1, x_2$  if  $d(x_1, x_2) \leq r$ .
- For  $x_1, x_2, x_3 \in X$  add a 2-simplex with vertices  $x_1, x_2, x_3$  if  $d(x_1, x_2), d(x_1, x_3), d(x_2, x_3) \leq r$ .
- ...

• For  $x_1, x_2, ..., x_m \in X$ , add a (m-1)-simplex with vertices  $x_1, x_2, ..., x_m$  if  $d(x_i, x_j) \le r$  for  $0 \le i, j \le m$ ; that is, if all the points are within a distance of r from each other.

The simplicial complex is called the Vietoris-Rips complex and is denoted  $V_r(X)$ .

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#### Relation to single linkage clustering algorithm

Our intuition tells us, that the model we build using the Vietoris-Rips complex to classify the images, produces the same results as the well known single linkage clustering algorithm. In this section we aim to prove or at least give a strong intuition that this is indeed the case.

**Definition 2.** A connected component in a graph G(V, E) (V is a set of vertices, E a set of edges) is a set of vertices C such that between every two points  $a, b \in C$  there exists a path in the graph G between them.

Both algorithms take a set X of n samples with m features as input. We can think of a sample x in the set X as a point in the m-dimensional space  $x \in \mathbb{R}^m$ . The algorithms then constructs a graph with points X as the vertices (V) and edges E. The k connected components in the constructed graph corespond to the classes of samples.

Single linkage algorithm. The algorithm starts with n connected components (no edges in the graph). n each step the algorithm chooses the two connected components that are closest to each according to some distance metric d (in our case the euclidean distance) and joins them into one by adding an edge between their closest two vertices.

**Definition 3.** The distance D between two connected components A and B is defined as the distance of the pair of vertices (one from A and one from B) that are closest to each other. More formally

$$D(A,B) = \min_{a \in A, b \in B} d(a,b).$$

The algorithm stops when there are only k connected components left.

**Vietoris-Rips classification algorithm.** The algorithm builds a (1-dimensional) Vietoris-Rips complex  $V_r(X)$  with parameter r. We choose the biggest r such that the Vietoris-Rips complex  $V_r(X)$  has k connected components.

To prove that the two algorithms indeed produce the same connected components we will first prove the next claim.

Claim 1. Let  $G_{sl}(V_{sl}, E_{sl})$  be a graph produced by the single linkage algorithm for finding k clusters and let  $d_{max}$  denote the distance between vertices in graph  $G_{sl}$  that were connected in the last iteration of the algorithm. Graph  $G_{sl}$  has connected components  $A_1...A_k$ . The graph  $G_{vr}(V_{vr}, E_{vr})$  induced by the Vietoris-Rips complex  $V_{d_{max}}(V_{sl})$  has the same connected components  $A_1...A_k$ .

*Proof.* To prove that we have the exact same connected components in  $G_{vr}$  as in  $G_{sl}$  the following must be true for  $G_{vr}$ :

- 1. The vertices inside  $A_i$  are connected i.e. there exists a path between arbitrary two vertices  $a, b \in A_i$ , for every  $i \in 1...k$ .
- 2. The sets of vertices  $A_1, ..., A_k$  are disconnected i.e. there isn't an edges e(a, b) between a pair of two points (a, b) such that  $a \in A_i, b \in A_j, i \neq j$ .

From the way the algorithm for single linkage works (it is adding edges of increasing length), we can easily see that  $d_{max}$  is the largest edge in the graph  $G_{sl}$ . From this and the definition of Vietoris-Rips (Definition 1) it follows that the set of edges in  $G_{sl}$  is a subset of edges in  $G_{vr}$  i.e.  $E_{sl} \subseteq E_{vr}$ . This means that all the edges found in  $G_{sl}$  will also appear in  $G_{vr}$  therefore the vertices inside individual subsets  $A_i$  will indeed be connected.

We prove the second part by contradiction. By definition of Vietoris-Rips complex  $V_{d_{max}}(V_{sl})$ , no edge  $e \in E_{vr}$  will be greater than  $d_{max}$ . Let's now assume that there exists an edge e with length  $d_e$  between two vertices a, b that are in separate subsets  $a \in A_i$  and  $b \in A_j$ ,  $i \neq j$ . Now we have two options. First one  $d_e < d_{max}$ . If this was the case than  $d_e$  would necessarily have to connect two points from the same subset  $A_i$ . This follows from the fact that single linkage algorithm always connects two of the closest connected components so any edge with a length smaller than  $d_{max}$  is an edge inside one of the subsets  $A_1, ... A_k$ . TODO: further explanation needed for the first case??? The second option is that  $d_e > d_{max}$ . This also immediately leads to a contradiction, this time with the Vietoris-Rips definition.

Using Claim 1 we see that the connected components in  $G_{vr}$  and  $G_{sl}$  are indeed the same. We need to take into account that the Vietrois-Rips algorithm takes the biggest such r so that the graph has k connected components, so  $r > d_{max}$ . But we can quickly see that the extra edges in the graph induced by  $V_r(V_{sl})$  will not change the connected components. After all we allready have k connected components in  $G_{vr}$ . To join two together would mean a violation of a fundamental rule of the algorithm.

## Testing

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#### References

[1] Leslie Lamport, LaTeX: a document preparation system, Addison Wesley, Massachusetts, 2nd edition, 1994.

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