

# COMP229: Introduction to Data Science

## Lecture 5: isometric point clouds

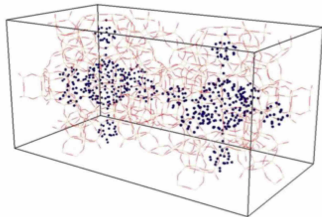
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Use your card with the scanner for registration

# Point clouds in real data

A real object is often represented (or described) by a cloud of points, for example

- *edgels* (edge pixels where intensity suddenly changes) are simple descriptors of images
- noisy (low quality) scans of visual markers
- point samples on surfaces around molecules



# Cloud classification problem

**Example problem.** Decide if any two clouds in  $\mathbb{R}^m$  (containing the same number of points) are isometric, i.e. can be obtained from each other by translations, rotations, reflections.

For 3-point clouds (triangles, also degenerate ones with 3 vertices in the same line), a complete invariant consists of 3 edge-lengths by Claim 4.8.

How many pairwise distances between  $n$  points (distinct triangles on  $n$  points) can we count?

# How many pairwise distances?

**Claim 5.2.** For  $n$  distinct points in any metric space, there are  $\frac{n(n-1)}{2}$  pairwise distances, which can be computed in time  $O(n^2)$ , i.e. we need at most  $Bn^2$  operations for a constant  $B$ .

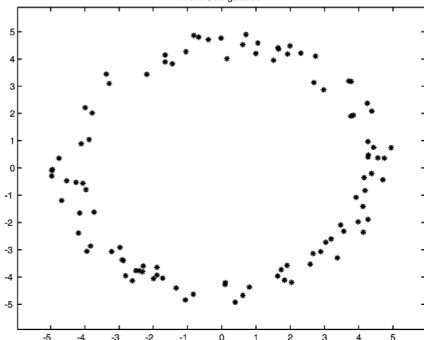
*Proof.* Any of  $n$  points has  $n-1$  distances to the remaining points. By symmetry these  $n(n-1)$  distances split into  $\frac{n(n-1)}{2}$  pairs. Each distance needs only a constant number of operations.  $\square$

Can all distances distinguish 4-point clouds?

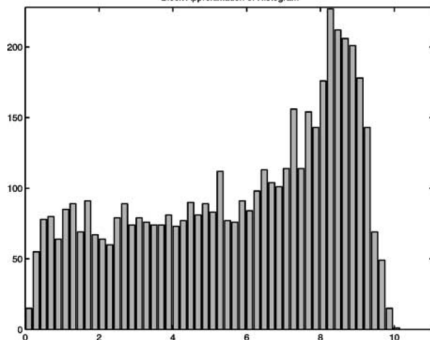
# Circular cloud and its distribution

The histogram on the right contains vertical bars. Each bar counts distances within a short range.

Point Configuration

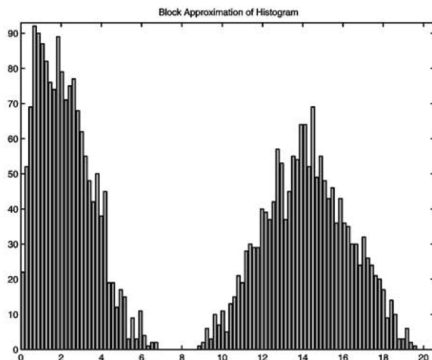
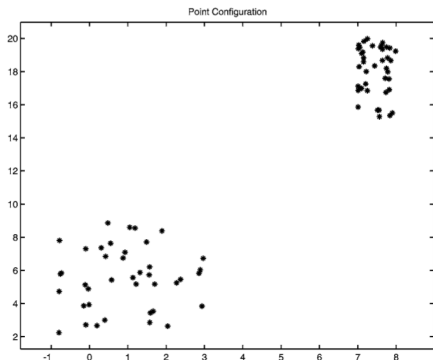


Block Approximation of Histogram



The histogram contains distances of all lengths from short to long, more longer distances.

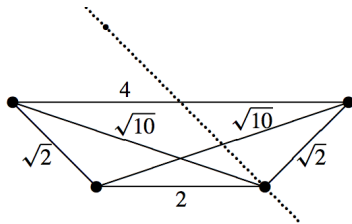
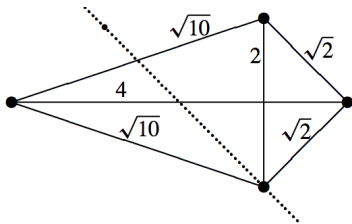
## 2-cluster cloud and its distribution



The histogram contains many short distances (within clusters), many long distances (between clusters) and very few mid-range distances.

# Interesting 4-point clouds

The 4-point clouds below have the exactly same distribution of 6 pairwise distances:  $\sqrt{2}$ ,  $\sqrt{2}$ , 2,  $\sqrt{10}$ ,  $\sqrt{10}$ , 4. Add the 5th marked point to each cloud to see how these clouds are related.



The 4-point clouds are not isometric, because their quadrilaterals have different areas.

# All distances in one polynomial

Add more points at corresponding positions in the dotted line gives larger clouds with the same distribution of distances. In general, a cloud cannot be uniquely reconstructed from distances.

**Definition 5.2.** Label points in a cloud  $C$  by  $1, 2, \dots, n$ . Let  $d_{ij}$  be the distance between the  $i$ -th and  $j$ -th points. The *distance polynomial* is

$F_C(x) = \prod_{1 \leq i < j \leq n} (x - d_{ij})$ , the product of linear

factors  $x - d_{ij}$ , where  $x$  is a real variable, e.g.

$F_C(x) = (x - d_{12})(x - d_{23})(x - d_{13})$  for 3 points.



# Reconstructible configurations

**Definition 5.3.** A cloud  $C$  is *reconstructible from distances* if for any other cloud  $C'$  such that their distance polynomials are equal ( $F_C(x) = F_{C'}(x)$  for all  $x$ ) there is an isometry of  $\mathbb{R}^m$  sending  $C$  to  $C'$ .

The 4-point clouds  $C, C'$  from slide 7 are not reconstructible from distances, because  $C, C'$  are not isometric, but their polynomials are equal:  
$$F_C(x) = (x - \sqrt{2})^2(x - 2)(x - \sqrt{10})^2(x - 4) = F_{C'}(x).$$
Luckily these are "almost all" exceptions.

# Reconstructible configurations

**Claim 5.4.** [no proof needed, Theorem 2.6 in  
Boutin, Kemper, Advances in Appl. Maths (2004)]

For any  $n \geq m + 2$ , there is a non-zero polynomial  $f(C)$  depending on (all coordinates of)  $n$  points of cloud  $C \subset \mathbb{R}^m$  such that if  $f(C) \neq 0$  then the cloud  $C$  is reconstructible from distances.

For any non-zero polynomial  $f$  depending on  $mn$  coordinates of  $n$  points from  $C$ , a random cloud  $C$  satisfies  $f(C) \neq 0$  with a high probability. Briefly, "almost any"  $C$  is reconstructible from distances.

# Your questions and the quiz

To benefit from the lecture, now you could

- ask or submit your anonymous questions to the COMP229 folder after the lecture;
- write down your summary in 2-3 phrases, e.g. list key concepts you have learned;
- talk to your classmates to revise the lecture.

**Question.** How many distinct triangles on  $n$  points can we count?

# Answer to the quiz and summary

**Answer.** We have  $n$  choices for a 1st vertex,  $n - 1$  for a 2nd vertex,  $n - 2$  for a 3rd vertex. The resulting  $n(n - 1)(n - 2)$  triangles split into 6-tuples related by 6 permutations of 3 vertices.

- $\frac{n(n - 1)}{2}$  pairwise distances between  $n$  points.
- $\frac{n(n - 1)(n - 2)}{6}$  distinct triangles on  $n$  points.
- "Almost all" clouds in  $\mathbb{R}^m$  are reconstructible from the distribution of pairwise distances.