MA-INF 4306 Lab Development and Application of Data Mining and Learning Systems: Big Data A measure for (weak) convexity Second Progress Meeting

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The top-level road-map

Define a measure for (weak) convexity



The top-level road-map

- 1 Define a measure for (weak) convexity
- 2 Test that measure on artificial data



The top-level road-map

- Define a measure for (weak) convexity
- 2 Test that measure on artificial data
- 3 Test the measure on real-world data



Questions addressed since the last meeting

1 Sanity check for the convex hull Algorithm



Questions addressed since the last meeting

- Sanity check for the convex hull Algorithm
- **2** Can ϵ and θ be modified as a measure?

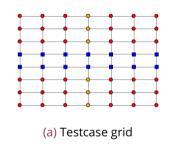


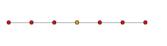
Questions addressed since the last meeting

- Sanity check for the convex hull Algorithm
- **2** Can ϵ and θ be modified as a measure?
- Idea for measures.



Sanity check for the convex hull Algorithm





(b) Testcase line

Figure: Application with $\theta=3$, $\epsilon=0.0$ and step distance; class 0 is red; class 1 is blue; class 1 assigned as the convex hull of class 0 is orange

$$A \subset X, z \in X, \epsilon \geq 0, \theta \geq 0.$$

$$z \in A \Leftrightarrow \exists x, y \in A$$
:

$$d(x,z) + d(z,y) \le d(x,y) + \epsilon \wedge d(x,z) + d(z,y) \ge d(x,y) - \epsilon \wedge d(x,y) \le \theta$$





In theory we can combine the notion of (θ, ϵ) -convexity as follows:

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① For $\epsilon = 0$ and $\theta \ge argmax \ d(x, y)$ we get the standard notion for convexity.

```
(z \in A \Leftrightarrow \exists x, y \in A : d(x, z) + d(z, y) = d(x, y))
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- ① For $\epsilon = 0$ and $\theta \ge \underset{(z \in A \Leftrightarrow \exists x, y \in A: d(x, z) + d(z, y))}{argmax} d(x, y)$ we get the standard notion for convexity.
- ② For $\epsilon = 0$ and $\theta \in [0, ... argmax d(x, y))$ we get θ -convexity [SHW21]. $x, y \in A$ $(z \in A \Leftrightarrow \exists x, y \in A : d(x, z) + d(z, y) = d(x, y) \land d(x, y) \leq \theta)$



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- **3** ϵ "smears" the boundary of the convex hull. For a big enough ϵ every point $z \in X$ is contained in the convex hull.



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- 1 For $\epsilon = 0$ and $\theta \ge \underset{x,y \in A}{\operatorname{argmax}} d(x,y)$ we get the standard notion for convexity. $x,y \in A$ $(z \in A \Leftrightarrow \exists x,y \in A: d(x,z) + d(z,y) = d(x,y))$
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- **3** ϵ "smears" the boundary of the convex hull. For a big enough ϵ every point $z \in X$ is contained in the convex hull.
- 4 If we find a "core" object $C \subset A$ which is the biggest convex subset, for what parameter ϵ and θ as inputs to a convex hull algorithm do we get A as result?

How does the number of components and the number of vertices in the convex hull develop for different θ s?

Figure: Application of the Algorithm with different θ , $\epsilon = 0.0$, n = 1000 vertices, 4 clusters and step distance;

class 0 is red; class 1 is blue; class 1 assigned as the convex hull of class 0 is orange

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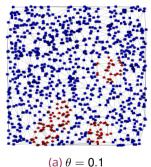


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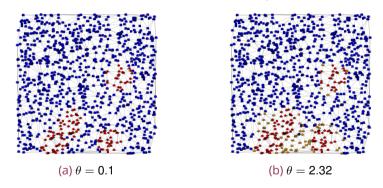


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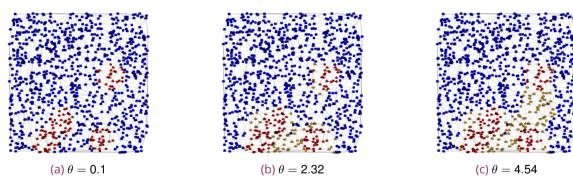


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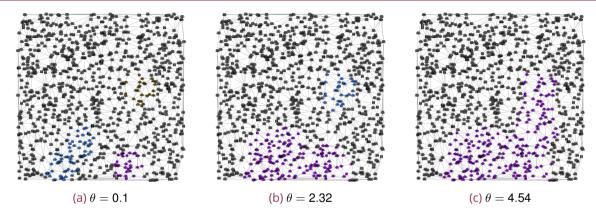
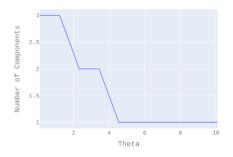
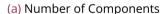


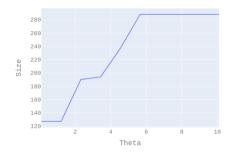
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vertices not part of a component and not part of the convex hull are grey, all other vertices are brown, blue or purple depending on their component.

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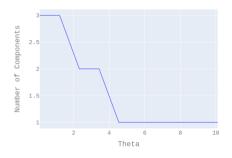


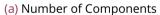


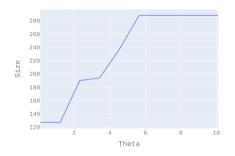


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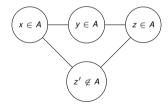
No answers for now. Questions for the future: Does a core object exists? Is it unique?

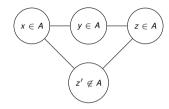
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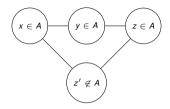
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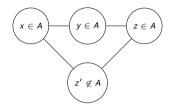






$$A = \{x, y, z\} \quad sp(x, y) = \{\{x, y\}\}, \ sp(y, z) = \{\{y, z\}\}, \ sp(x, z) = \{\{x, y, z\}, \{x, y, z'\}\}$$

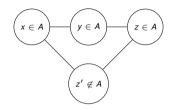




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Polygonal Entropy: a convexity measure [Ste89]

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Convexity of a polygon P is $C(P) = E(P)/E_{max}(P)$ with $E_{max}(P) = -ln(1/A(P))$



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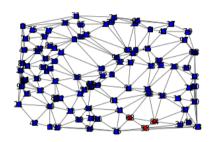
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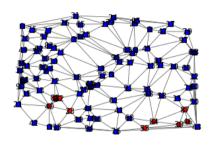


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Entropy of a vertex set E(P) = -\sum f(p)ln(f(p))
Convexity of a set of vertices is C_2(P) = E(P)/E_{max}(P) with E_{max}(P) = -\ln(1/A(P))
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First Results (I)



(a)
$$C_1(P) = 0$$
, $C_2(P) = 1$

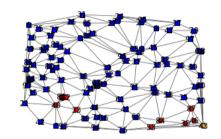


(b) $C_1(P) \approx 1.3$, $C_2(P) \approx 0.65$

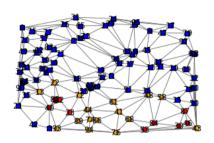
Figure: Step distance



First Results (II)



(a)
$$\theta = 2$$
, $C_1(P) \approx 1.090$, $C_2(P) \approx 0.65$, convex



(b)
$$\theta = 3$$
, $C_1(P) = 0$, $C_2(P) = 1$

Figure: Step distance; Results for measures after application of the convex hull algorithm



• Can ϵ and θ be modified as a measure?



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Thank you for your attention!

What are your questions?



References



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