

Magnitude and other measures of metric spaces

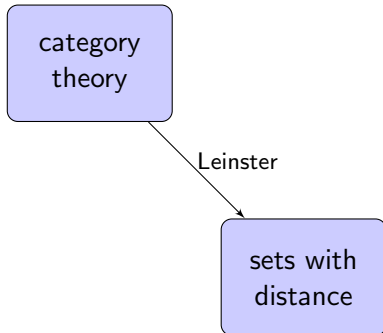
Simon Willerton
University of Sheffield

Exploratory meeting on
the mathematics of biodiversity
CRM Barcelona
July 2012

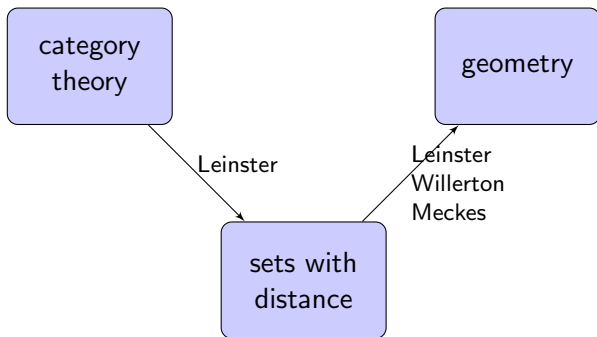
Overview

category
theory

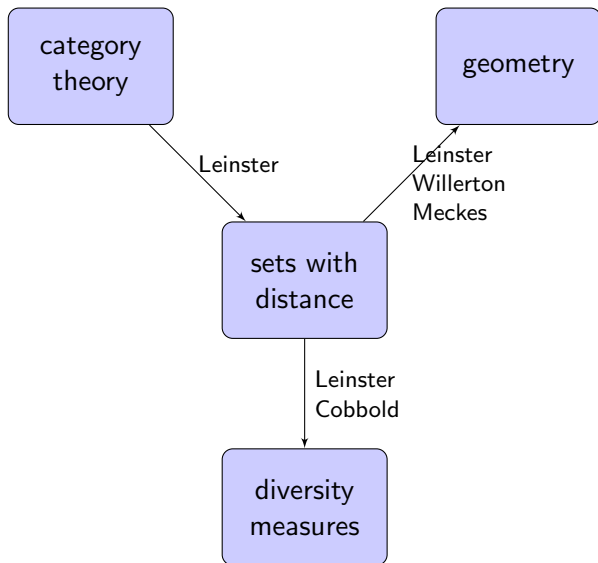
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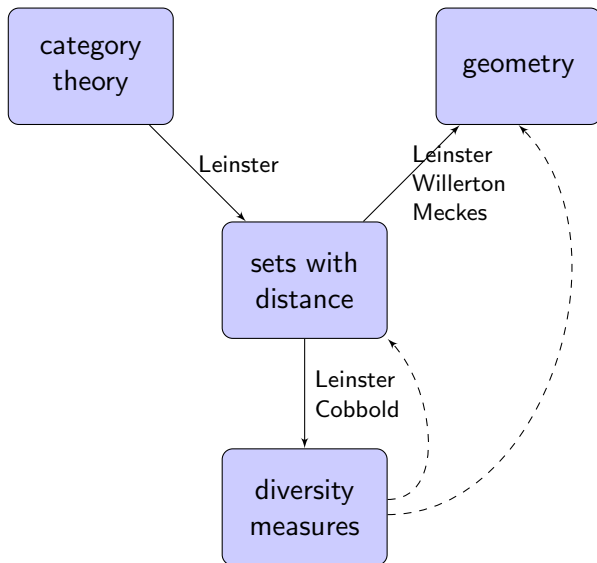
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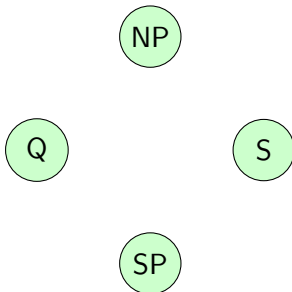
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- ▶ some notion of distance $0 \leq d_{ij} \leq \infty$ between the i th and j th points.

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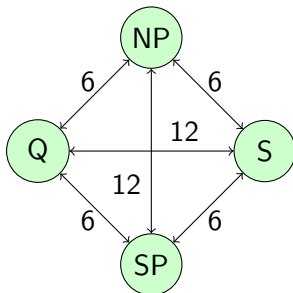


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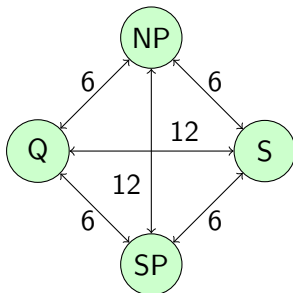


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Note: not every metric space can be thought of as points in Euclidean space.

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Metric space X with similarity matrix $Z_{ij} := e^{-d_{ij}}$.

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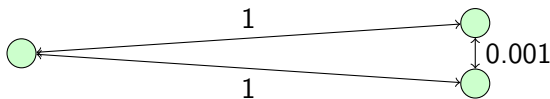
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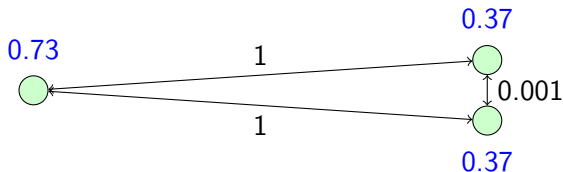


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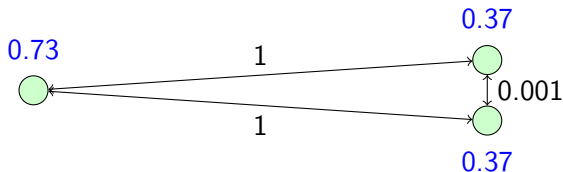


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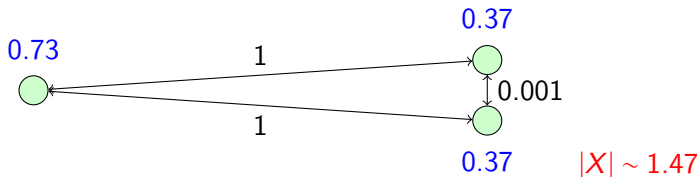
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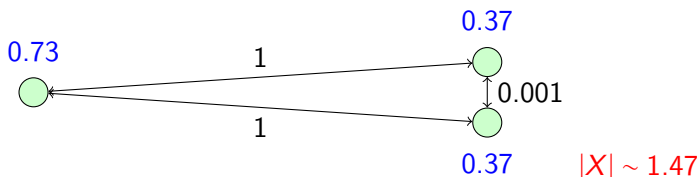
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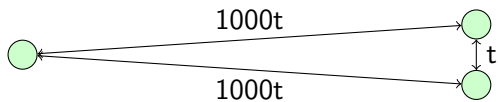


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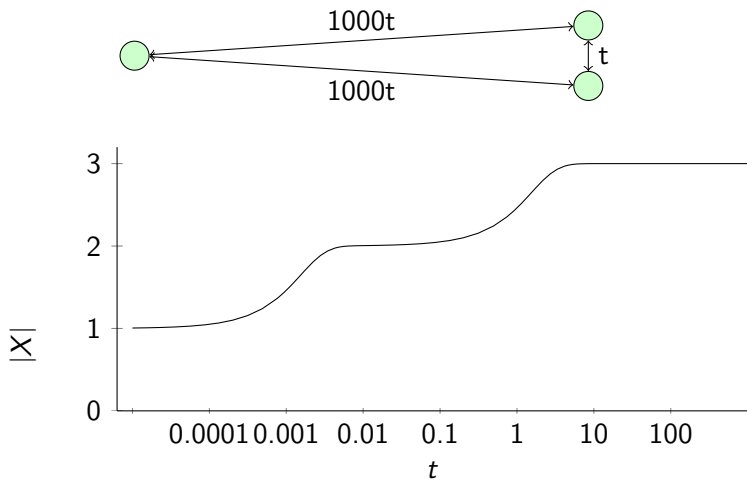
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If Z_{ij} is invertible then $|X| = \sum_{ij} (Z^{-1})_{ij}$.

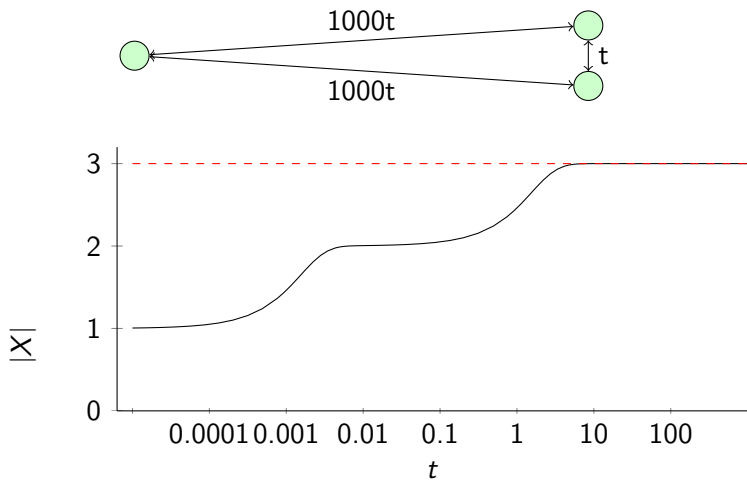
Example of scaling



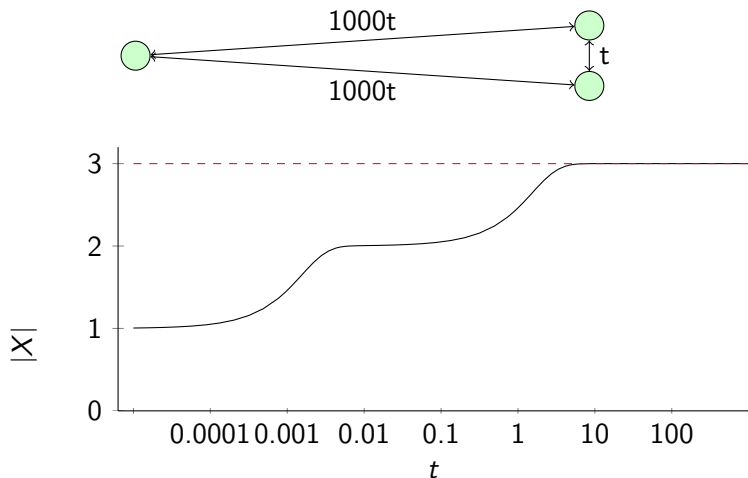
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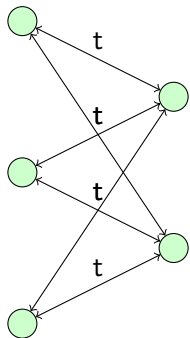


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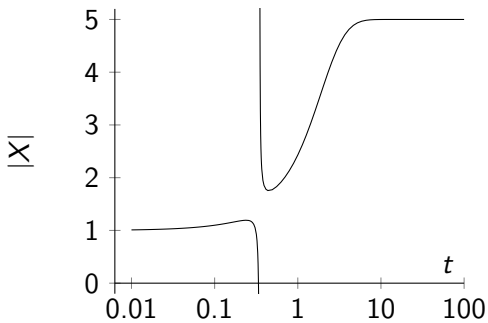
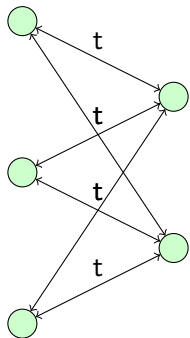


As any space X is scaled bigger and bigger $|X| \rightarrow N$.

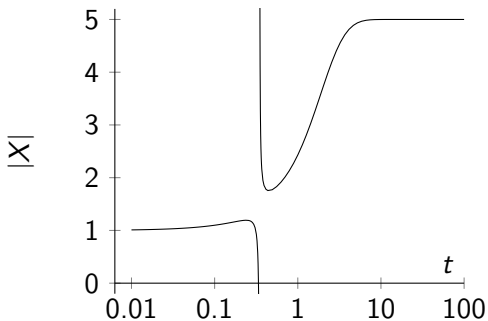
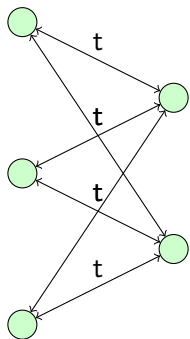
Example of bad metric space



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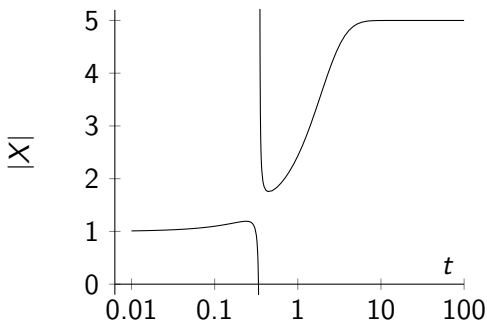
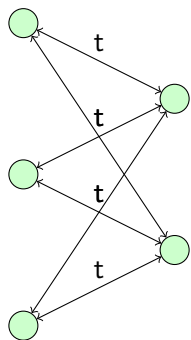


Example of bad metric space



Many metric spaces are better behaved than this.

Example of bad metric space



Many metric spaces are better behaved than this.

If Z is positive definite then $|X|$ is defined.

For example, if X is a subset of Euclidean space then $|X|$ is defined.

Diversity measures [Leinster, Cobbold]

Model our community using

- ▶ a metric space X with similarity matrix Z_{ij}
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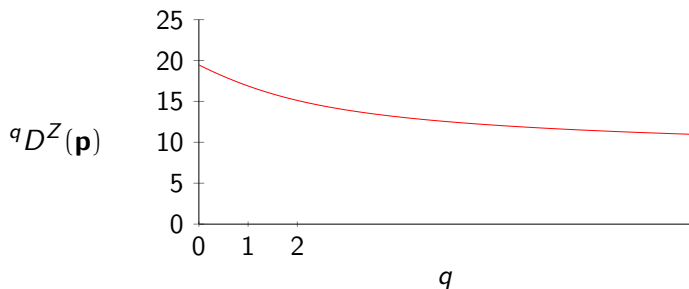
$${}^q D^Z(\mathbf{p}) := \begin{cases} \left(\sum_{i:p_i>0} p_i (Z\mathbf{p})_i^{q-1} \right)^{\frac{1}{1-q}} & q \neq 1, \\ \prod_{i:p_i>0} (Z\mathbf{p})_i^{-p_i} & q = 1, \\ \min_{i:p_i>0} \frac{1}{(Z\mathbf{p})_i} & q = \infty. \end{cases}$$

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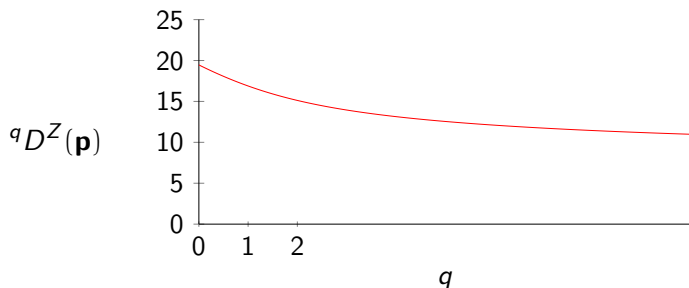


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Effective number of species:



Recover various other measures of diversity using this.

For example, obtain Hill numbers when $d_{ij} = \infty$ (i.e. $Z_{ij} = 0$) for $i \neq j$.

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- ▶ Otherwise

$$D_{\max}(Z) = \max_{Y \subset X \& w_i > 0} |Y|.$$

Summary of magnitude $|X|$

- ▶ Mathematically natural (if mysterious), c.f. category theory.
- ▶ Related to biodiversity.
- ▶ Seemingly related to geometry in Euclidean space.
- ▶ Can behave rather weirdly at times.

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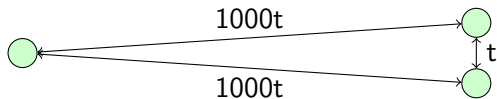
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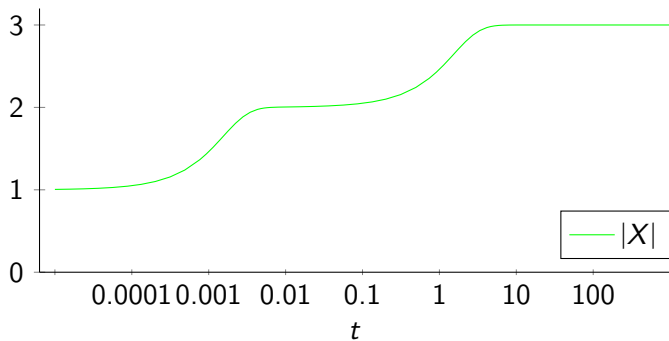
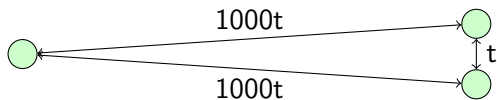
Note: this is **not** the same as

$$|X| = \sum_{i=1}^N \sum_{j=1}^N (Z)_{ij}^{-1}$$

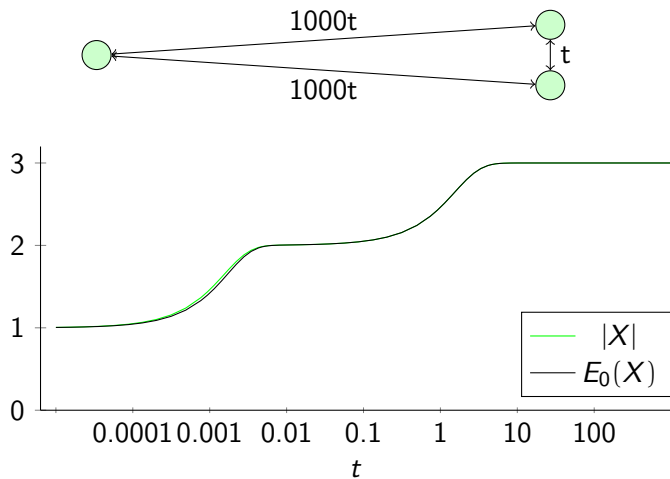
Example of scaling II



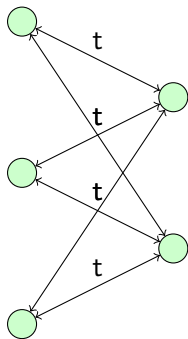
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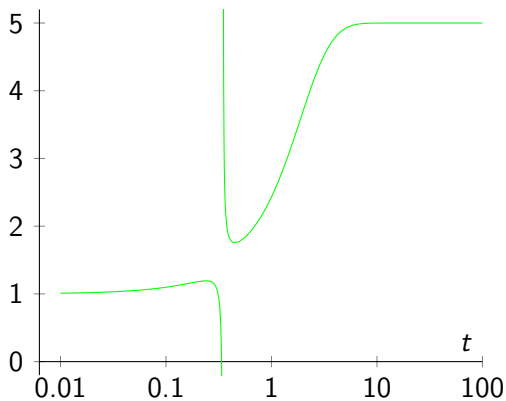
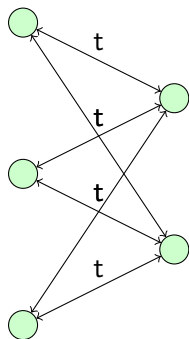
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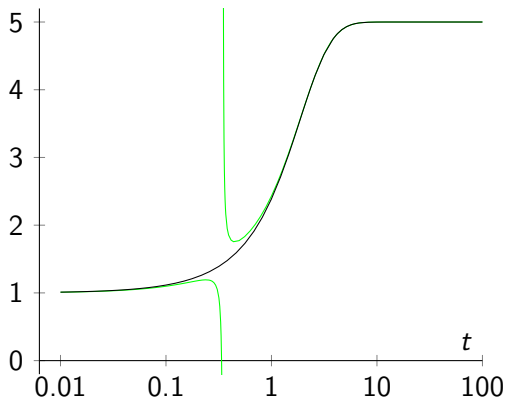
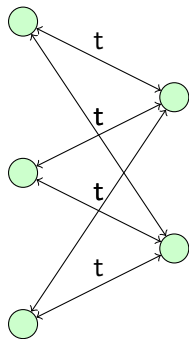
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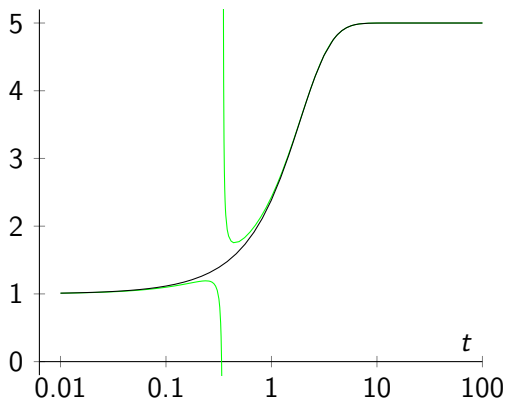
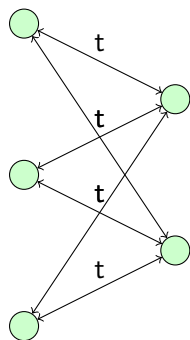
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- ▶ The size ${}^0E(X)$ is defined for all metric spaces.
- ▶ As X is scaled up ${}^0E(X)$ increases from 1 to N .
- ▶ It is much easier to calculate ${}^0E(X)$ than $|X|$.

Zooming in on a space with 6400 points

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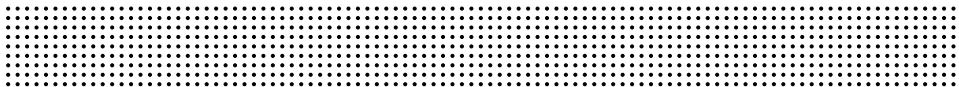
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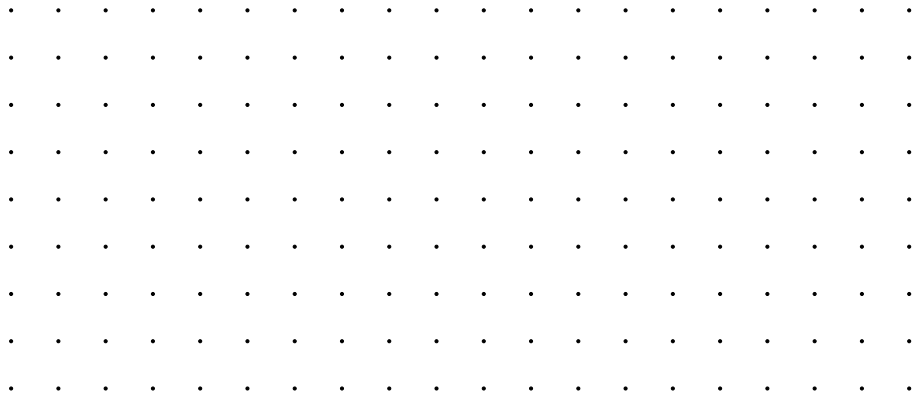
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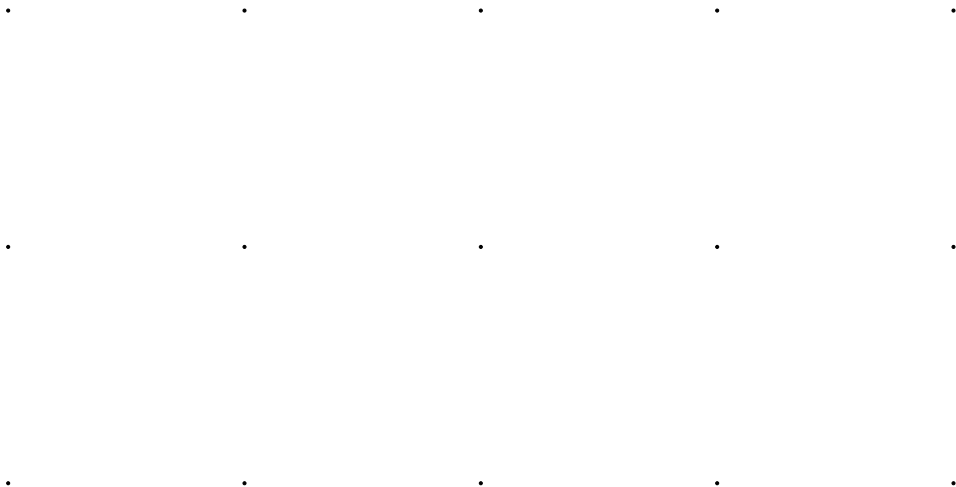
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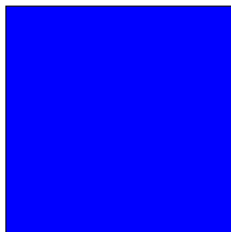
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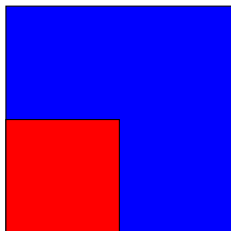
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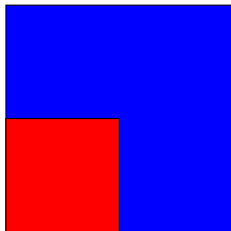
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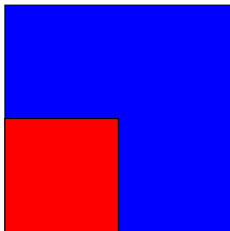
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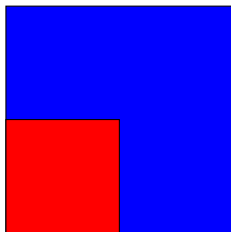
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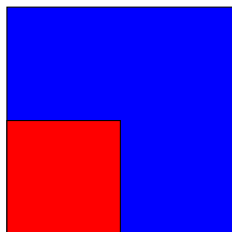
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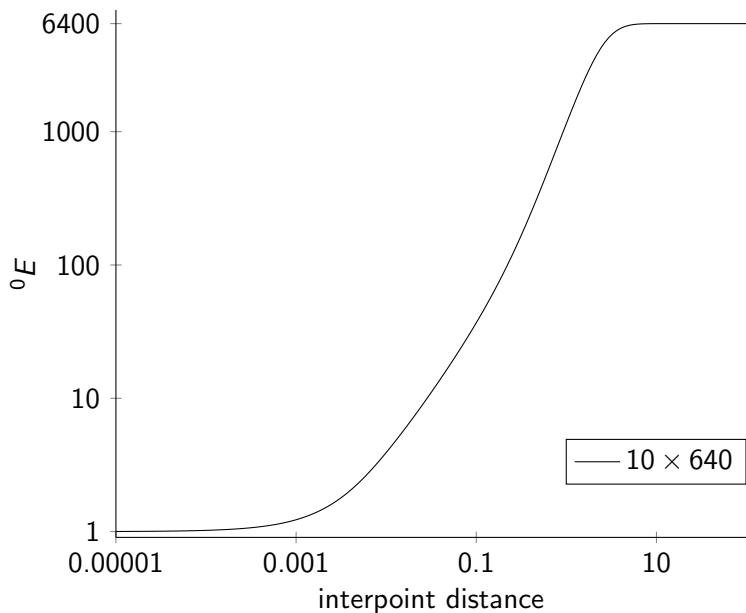
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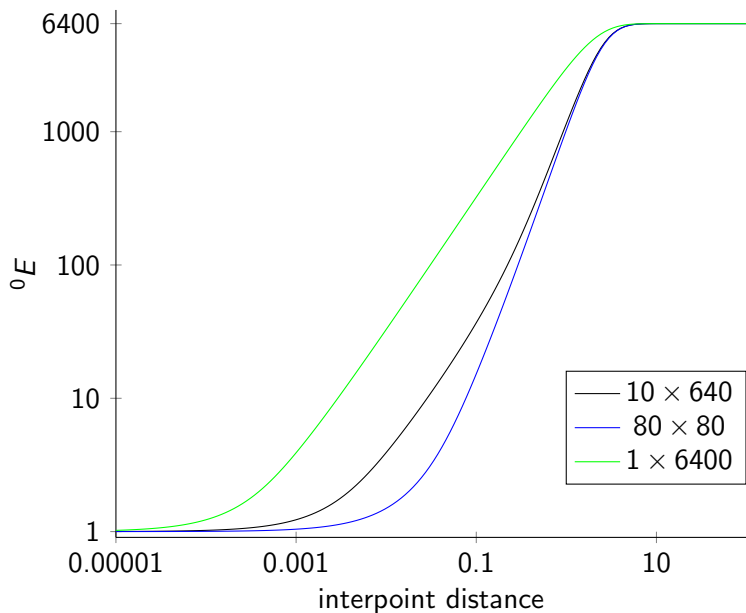
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Think of dimension as how the size changes when the distances are changed.
Given 'size' can see if it gives a good idea of dimension.

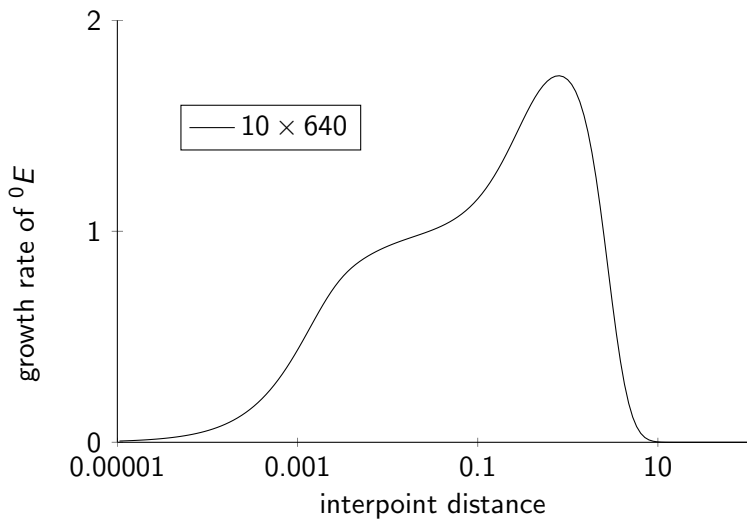
Size of rectangles with 6400 points



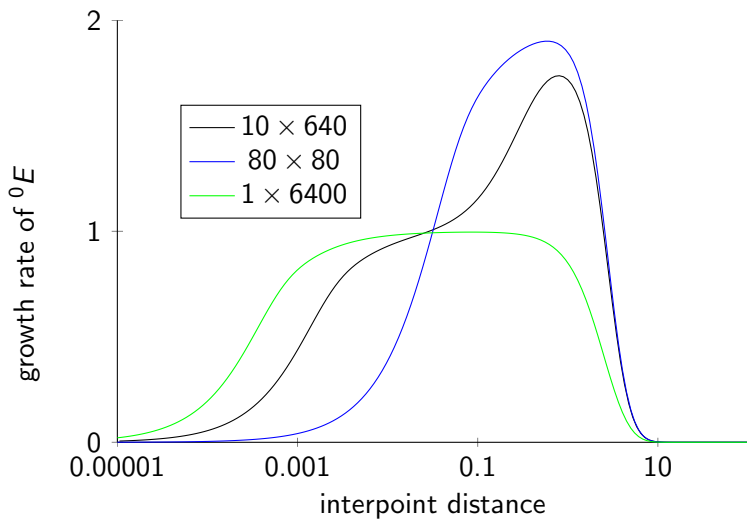
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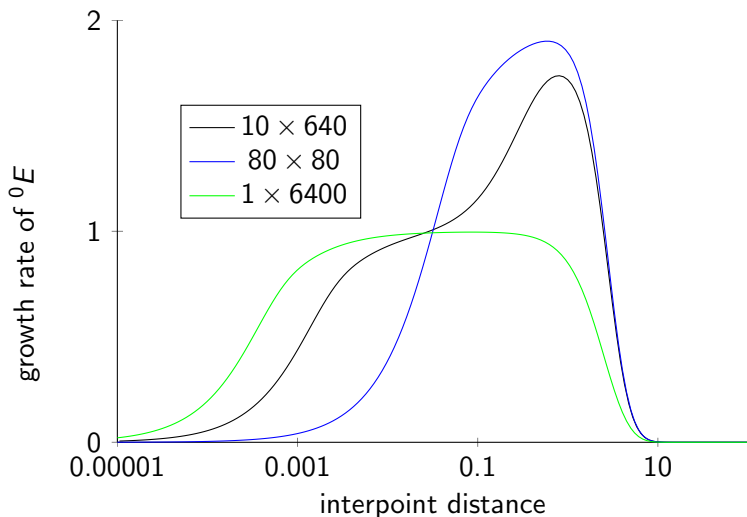
Rectangles with 6400 points and 'dimension'



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There is geometric information is ${}^0E(X)$.