Instantaneous dimension of finite metric spaces via magnitude and spread

> Simon Willerton University of Sheffield

Applied Topology in Będlewo 2017

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Applied Category Theory in Będlewo 2017

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

### Measure of size: spread [Willerton]

One parameter family of measures of size  $\{E_q\}_{q=0}^{\infty}$ .

Closely related to magnitude but simpler and better behaved.

$$\mathsf{E}_{q}(X) := \left(\frac{1}{N}\sum_{i} \left(\frac{N}{\sum_{j} \mathrm{e}^{-\mathsf{d}(i,j)}}\right)^{1-q}\right)^{\frac{1}{1-q}}$$

In particular,

$$\mathsf{E}_{0}(X) = \sum_{i}^{N} \frac{1}{\sum_{j} e^{-\mathsf{d}(i,j)}}; \qquad \mathsf{E}_{2}(X) = \frac{N^{2}}{\sum_{i,j} e^{-\mathsf{d}(i,j)}};$$

If X is homogeneous then  $E_q(X) = |X|$  for all q.

#### More on spread



1.  $1 \leq E_q(X) \leq N$ ; 2.  $E_q(tX)$  is increasing in t; 3.  $E_q(tX) \rightarrow 1$  as  $t \rightarrow 0$ ; 4.  $E_q(tX) \rightarrow N$  as  $t \rightarrow \infty$ .

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3 times as big

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### Dimension

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For example, triple the distances:



 $3^1 = 3$  times as big

$$3^2 = 9$$
 times as big

$$3^{\log_3 2} = 2$$
 times as big

### Dimension

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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.

## Instantaneous dimension

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$$S(tX) = a \cdot t^d$$

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$$\dim_{\mathcal{S}}(tX) := \frac{\mathrm{d}\log(\mathcal{S}(tX))}{\mathrm{d}\log t}.$$

Think of  $\dim_S(tX)$  as t varies as the dimension profile of X.

Can you identify these 10 point clouds from just their dimension profile????





Answer: Our little 3 point space.







Answer: 1,000 points in the interval [0,1].







Answer: 100,000 points in the interval [0,1].







Answer:  $270 \times 270$  grid of points.





Answer:  $12 \times 6000$  grid of points.







Answer: 2000 points in a circle in the plane.







Answer: 20000 points in a circle in the plane.







Answer: 10000 points in a 'noisy' circle in the plane.







Answer: 2048 points in the Cantor set.



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Profile 10



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Answer: 16385 points in the Sierpinski gasket.



## Calculate the dimension profiles for some interesting data sets!