# Measuring metric spaces: short sightedness and population diversity

Simon Willerton

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1: Defining the cardinality

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### Definition

A metric space is a set X with a 'distance'  $d_{ij} \in [0, \infty]$  between each pair of points  $i, j \in X$  such that

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- triangle inequality:  $d_{ik} \leq d_{ij} + d_{jk}$
- no self-distance:  $d_{ii} = 0$  for all  $i \in X$
- separation: if  $i \neq j$  then  $d_{ij} \neq 0$
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#### Note

Not every metric space is embeddable in Euclidean space.

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#### Theorem

For any X with n points there exists a  $t_0$  such that for  $t > t_0$  the cardinality |tX| exists, is increasing and tends to n as  $t \to \infty$ .

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$$\sum_{j \in X} e^{-d_{ij}} w_i = 1 \quad \text{for all } i \in X, \qquad \text{i.e.} \quad Zw = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

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The weights do not have to be positive!

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#### Example

There is a useful family  $\sigma_{\alpha}$  of surprise functions.

$$\sigma_{lpha}(oldsymbol{p}):=rac{1-oldsymbol{p}^{lpha-1}}{lpha-1}\qquad lpha\in [0,\infty)$$

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## Diversity measures (definition)

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If (X, p) is a probability metric space then the  $\alpha$ -diversity (or expected  $\alpha$ -surprise) is

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$$|(X,p)|_{0} = \sum \frac{p_{i}}{(Zp)_{i}} \qquad |(X,p)|_{1} = \frac{1}{\prod (Zp)_{i}^{p_{i}}} \\ |(X,p)|_{2} = \frac{1}{p^{T}Zp} \qquad |(X,p)|_{\infty} = \frac{1}{\max\{(Zp)_{i}\}}$$

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#### Note

In the case of a discrete space (all the points infinitely far apart), for all  $\alpha$  the  $\alpha$ -cardinality is maximized by the uniform probability and takes value *n*.

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#### Theorem

If X is a finite metric space with a positive weighting w, then  $\overline{p}_i := \frac{w_i}{|X|}$  is a probability measure and

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So it looks like a weighting on a metric space is analogous to the uniform distribution on a set of points.

3: Cardinality and continuous metric spaces

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Approximate by a set of points: take  $\sum_{1}^{n-1} d_i = a$  and let  $X_d$  be

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So we can define the cardinality of the length a line segment

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Let  $C_a$  be the circle of circumference *a* (with the metric induced from  $\mathbb{R}^2$ ).

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### Theorem (Hadwiger's Theorem)

There is a canonical basis  $\{\mu_m, \ldots, \mu_0\}$  of invariant valuations on subsets of  $\mathbb{R}^m$  and these have the scaling property  $\mu_i(tA) = t^i \mu_i(A)$ .

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#### Theorem

The Wills function  $W(A) := \mu_m(A) + \mu_{m-1}(A) + \cdots + \mu_0(A)$ is multiplicative:  $W(A \times B) = W(A) \times W(B)$ .

## Asymptotic conjecture

A		W(A)
finite collection of points	· · · · · · · · · · · · · · · · · · ·	(number of points)
closed interval	/	(length) + 1
polygon		(perimeter)
filled polygon		$(area) + \frac{1}{2}(perimeter) + 1$
unit ball in $\mathbb{R}^3$		$\frac{4}{3}\pi + 2\pi + 4 + 1$

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#### Conjecture

The cardinality can be defined for any compact subset of  $\mathbb{R}^n$  and

$$|tA| - W(tA) \rightarrow 0$$
 as  $t \rightarrow \infty$ 

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