

# Metric Spaces, Entropic Spaces and Convexity

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Lawvere: Thermodynamics (1984)

State space  $X$ :  $\bar{\mathbb{R}}^0$ -category (entropic space)

Entropy:  $\bar{\mathbb{R}}^0$ -functor  $X \rightarrow [-\infty, \infty]$

Baez-Lynch-Moeller: Thermostatistics (2021)

State space  $X$ : convex space

Entropy: concave map  $X \rightarrow [-\infty, \infty]$

GOAL: Synthesise these to get a category of convex entropic spaces & concave maps.

Inspiration: Fritz-Perrone approach to convexity monad on metric spaces.

Enrich over a commutative quantale  $\mathbb{R} = (\text{ob } \mathbb{R}, \geq, +, 0, -)$

(ie. (co)complete, skeletal, closed symmetric monoidal thin category)

- Eg. •  $\overline{\mathbb{R}}_+ = ([0, \infty], \geq, +, 0)$       $a - b = \max(0, b - a)$   
•  $\overline{\mathbb{R}} = ([-\infty, +\infty], \geq, +, 0)$       $a - b = b - a$   
•  $\overline{\mathbb{R}}^\circ = ([-\infty, +\infty], \leq, +, 0)$       $a - b = b - a$

$\mathbb{R}$ -category  $X$ :  $X(a, b) \in \text{ob } \mathbb{R}$   
 $\forall a, b, c \in X$       $X(a, b) + X(b, c) \geq X(a, c)$   
                                  $0 \geq X(a, a)$

Eg.  $\overline{\mathbb{R}}_+$ -category is generalization of classical metric space.  
•  $\mathbb{R}$ -category  $\underline{\mathbb{R}}$  with  $\underline{\mathbb{R}}(a, b) = a - b$ .

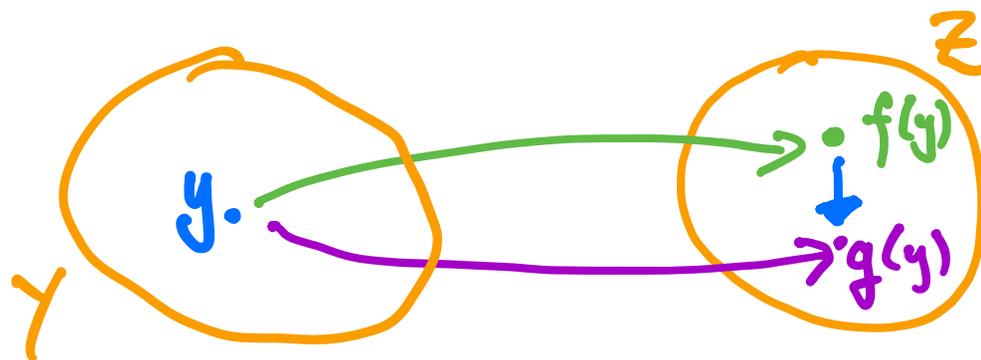
Get a category  $\mathbb{R}\text{-cat}$

Morphisms: short maps - distance non-increasing maps

closed monoidal:

$\text{ob } \llbracket Y, Z \rrbracket = \{\text{short maps } Y \rightarrow Z\}$

$$\llbracket Y, Z \rrbracket(f, g) = \sup_{y \in Y} Z(f(y), g(y))$$



$\llbracket Y, \mathbb{Q} \rrbracket$

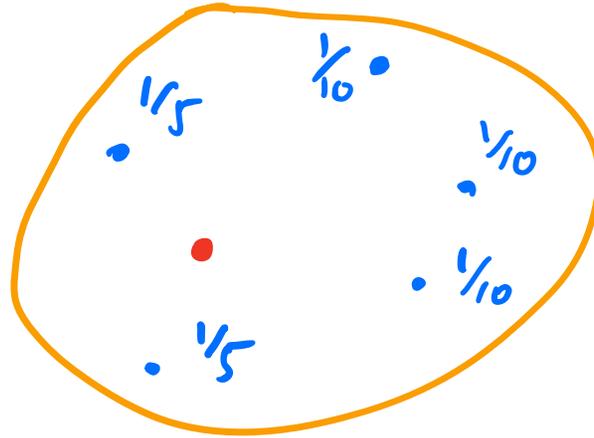
- "scalar-valued functions"  
or "op-presheaves"

# Convex space A

$\forall n \in \mathbb{N}, \alpha_1, \dots, \alpha_n \geq 0, \sum \alpha_i = 1, x_1, \dots, x_n \in A$

$\exists \sum \alpha_i x_i \in A$   
(barycentre)

+ axioms



- Eg.
- Any  $\mathbb{R}$ -vector space or affine space
  - $[0, \infty]$  where  $\alpha \infty + (1-\alpha)a = \infty$  ( $\alpha > 0$ )
  - $[-\infty, +\infty]$  where  $\alpha(+\infty) + (1-\alpha)(-\infty) = +\infty$   
(or can define the other way)

Algebras for convexity monad  $C: \text{Set} \rightarrow \text{Set}$   
 $C(S) = \{ \sum \alpha_i \ulcorner x_i \urcorner \mid \sum \alpha_i = 1 \}$  formal

## Convex quantale:

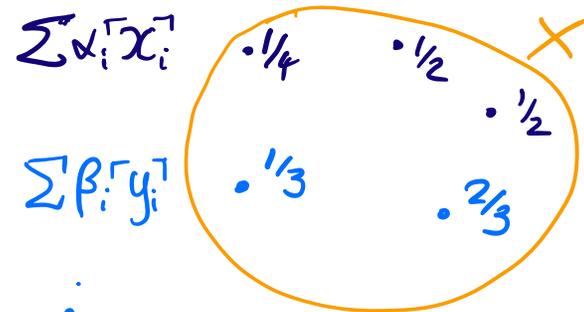
quantale  $\mathbb{R}$  with  $\text{ob}\mathbb{R}$  a convex set and

- $a_i \geq b_i \Rightarrow \sum \alpha_i a_i \geq \sum \alpha_i b_i$
- $\sum \alpha_i (a_i - b_i) \geq (\sum \alpha_i a_i) - (\sum \alpha_i b_i)$

- Eg
- $\overline{\mathbb{R}}_+ = ([0, \infty], \geq)$
  - $\overline{\mathbb{R}} = ([-\infty, +\infty], \geq)$   $+\infty$  dominates
  - $\overline{\mathbb{R}}^\circ = ([-\infty, +\infty], \leq)$   $-\infty$  dominates

Monad  $C: \text{Set} \rightarrow \text{Set}$ , want  $\mathbb{R}\text{-cat} \rightarrow \mathbb{R}\text{-cat}$

$$C^{\text{DD}}(X)(\sum \alpha_i \cdot x_i, \sum \beta_j \cdot y_j) = ??$$



Consider  $\sum \alpha_i \cdot x_i$  as a finitely supported probability measure.  
Get a set map

$$C(X) \times \text{ob}[\mathbb{X}, \mathbb{R}] \rightarrow \text{ob} \mathbb{R}; \quad (\sum \alpha_i \cdot x_i, f) \mapsto \underbrace{\sum \alpha_i \cdot f(x_i)}_{\mathbb{R} \text{ convex}}$$

Gives

$$C(X) \rightarrow \text{ob}[[\mathbb{X}, \mathbb{R}], \mathbb{R}] \leftarrow \text{short maps}$$

Pull back the metric

$$C^{\text{DD}}(X)(\sum \alpha_i \cdot x_i, \sum \beta_j \cdot y_j) = \sup_{f \in [\mathbb{X}, \mathbb{R}]} \sum \alpha_i \cdot f(x_i) - \sum \beta_j \cdot f(y_j)$$

This has an optimal transport interpretation in terms of costs. 6

We get a category of algebras and algebra maps for  $C^\infty$ .

- Algebras are **convex  $\mathbb{R}$ -categories** : convex spaces which are  $\mathbb{R}$ -categories in a compatible way.

[Eg.  $\sum \alpha_i X(x_i, y_i) \geq X(\sum \alpha_i x_i, \sum \alpha_i y_i)$   
This is sometimes sufficient.]

- Algebra maps are **convex linear maps**  
 $f: X \rightarrow Y$  s.t.  $f(\sum \alpha_i x_i) = \sum \alpha_i f(x_i)$

These are the wrong maps!  
We want concave/convex maps.

We can fix this.

$C^{\text{DD}}$  actually gives an  $\mathbb{R}$ -cat-enriched monad

$$C^{\text{DD}}: \underline{\mathbb{R}\text{-CAT}} \rightarrow \underline{\mathbb{R}\text{-CAT}}$$

The "underlying category" of any  $\mathbb{R}$ -category is a poset

$$x \geq_x y \iff 0 \geq X(x, y)$$

This gives a 2-monad

$$C^{\text{DD}}: \mathbb{R}\text{-CAT} \rightarrow \mathbb{R}\text{-CAT}$$

objects:  $\mathbb{R}$ -categories  
 morphisms: short maps  
 2-morphisms:  $\geq$

- Strict algebras: convex  $\mathbb{R}$ -categories
- Lax & colax algebra maps: convex / concave maps

$$\begin{array}{ccc} X & C^{\text{DD}}(X) & \rightarrow X \\ \downarrow f & C^{\text{DD}}(f) \downarrow & \cong \downarrow f \\ Y & C^{\text{DD}}(Y) & \rightarrow Y \end{array}$$

$$\sum \alpha_i f(x_i) \geq f(\sum \alpha_i x_i)$$

convex

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Entropy:  $\bar{\mathbb{R}}^0$ -functor  $X \rightarrow [-\infty, \infty]$

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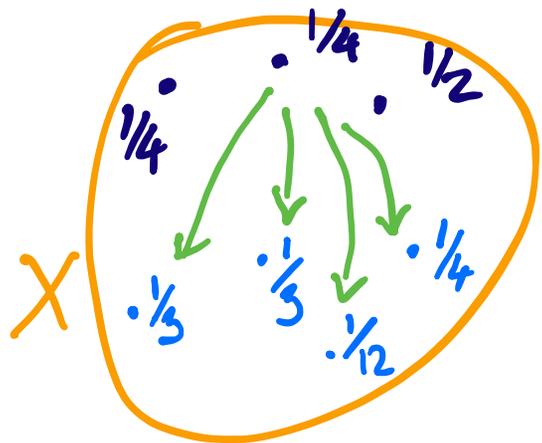
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There's another way to define a distance between convex linear combinations.  
 Optimal transport plan of goods: minimize cost.

$$C^{opt}(X)(\sum \alpha_i \cdot x_i, \sum \beta_j \cdot y_j) = \inf \left\{ \sum \delta_{ij} X(x_i, y_j) \mid \sum_j \delta_{ij} = \alpha_i, \sum_i \delta_{ij} = \beta_j \right\}$$



$\delta_{ij}$  is amount of goods  $x_i \rightsquigarrow y_j$   
 $\delta_{ij} X(x_i, y_j) = \text{cost of transportation}$

(aka Wasserstein or Kantorovich-Rubinstein metric)  
 Not clear to me how to interpret this in category theory terms.

"Weak duality" holds:

$$C^{opt}(X)(\sum \alpha_i \cdot x_i, \sum \beta_j \cdot y_j) \geq C^{DD}(X)(\sum \alpha_i \cdot x_i, \sum \beta_j \cdot y_j)$$