

An Enriched Category Theory of Language

Tai-Danae Bradley

**a mathematical framework for language
(inspired by LLMs)**

category theory

Why? It allows us to bring in *few* assumptions.

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category theory

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language

It's a little different than fitting a model to language.
Instead, we're just seeing "what's there."

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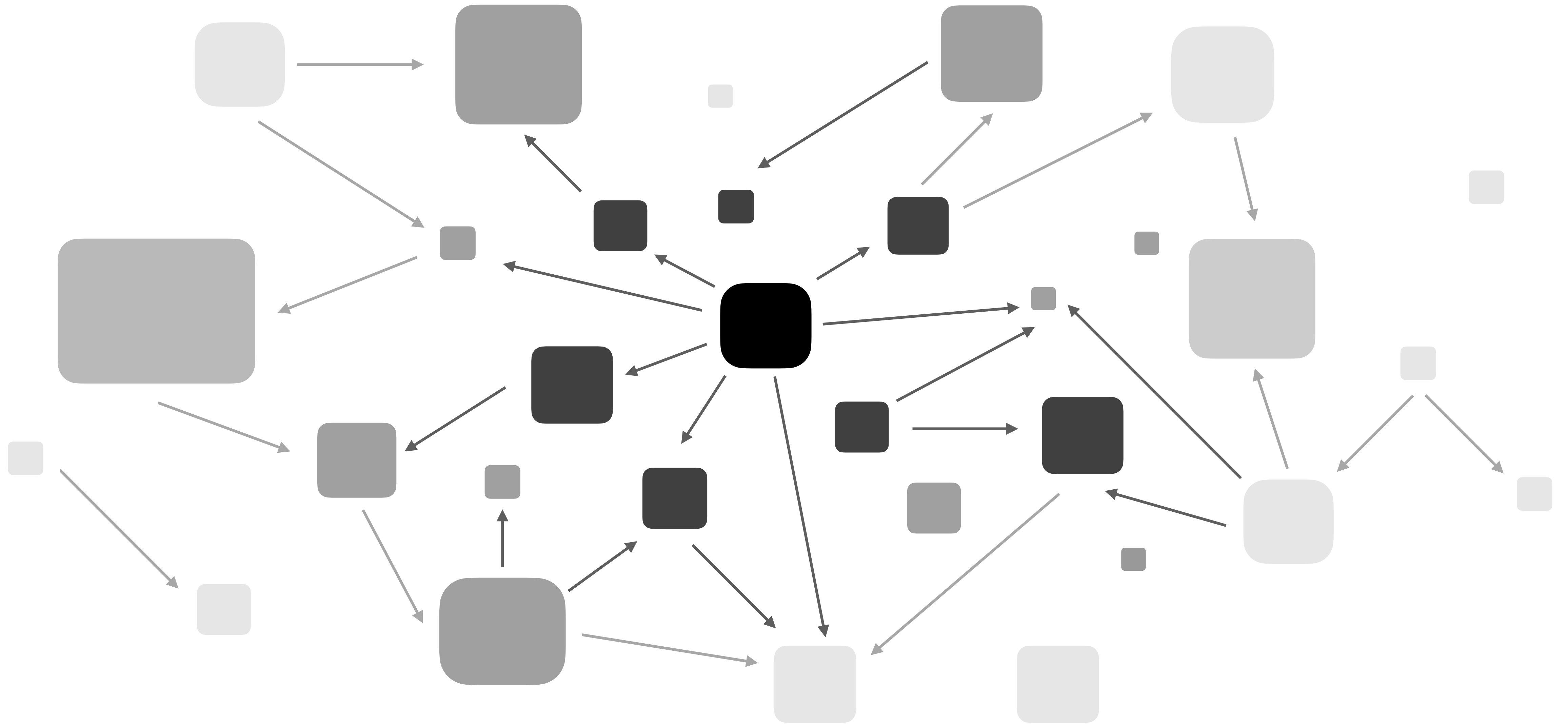
**AN ENRICHED CATEGORY THEORY OF LANGUAGE: FROM
SYNTAX TO SEMANTICS**

TAI-DANAE BRADLEY¹, JOHN TERILLA², AND YIANNIS VLASSOPOULOS³

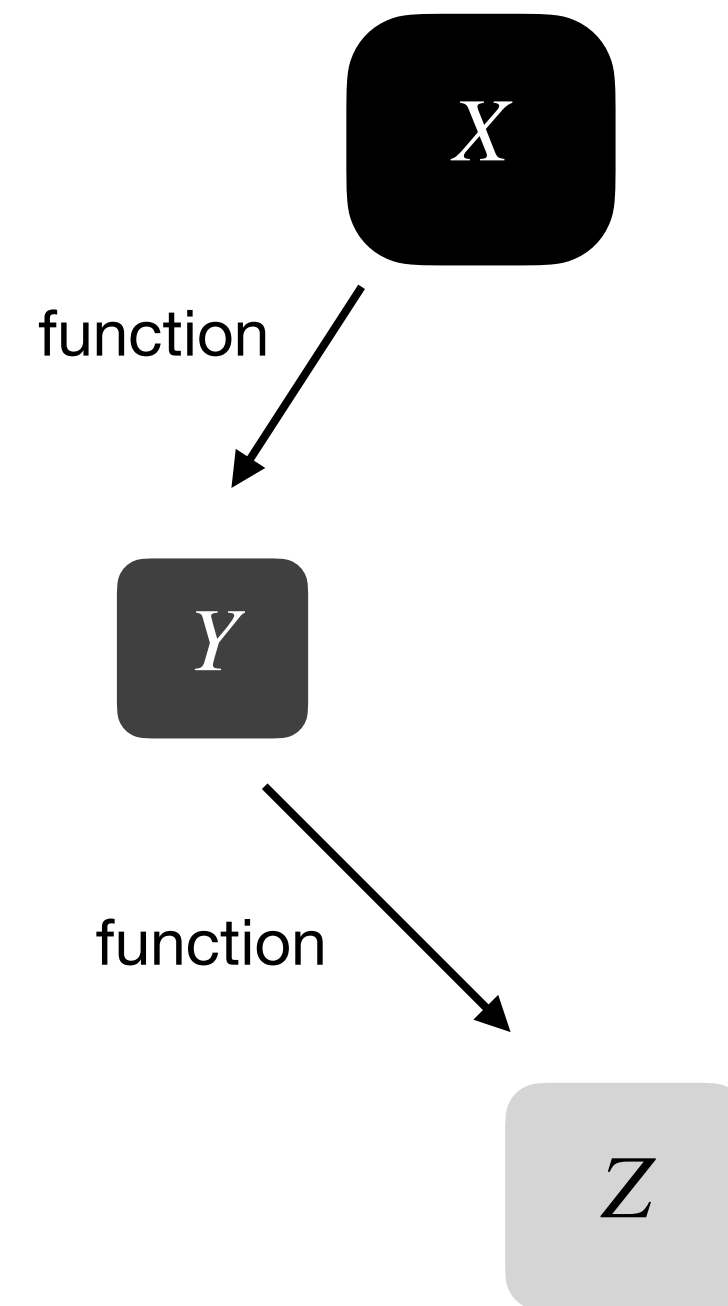
enriched

The statistics of texts can be captured neatly using
a "richer" version of category theory.

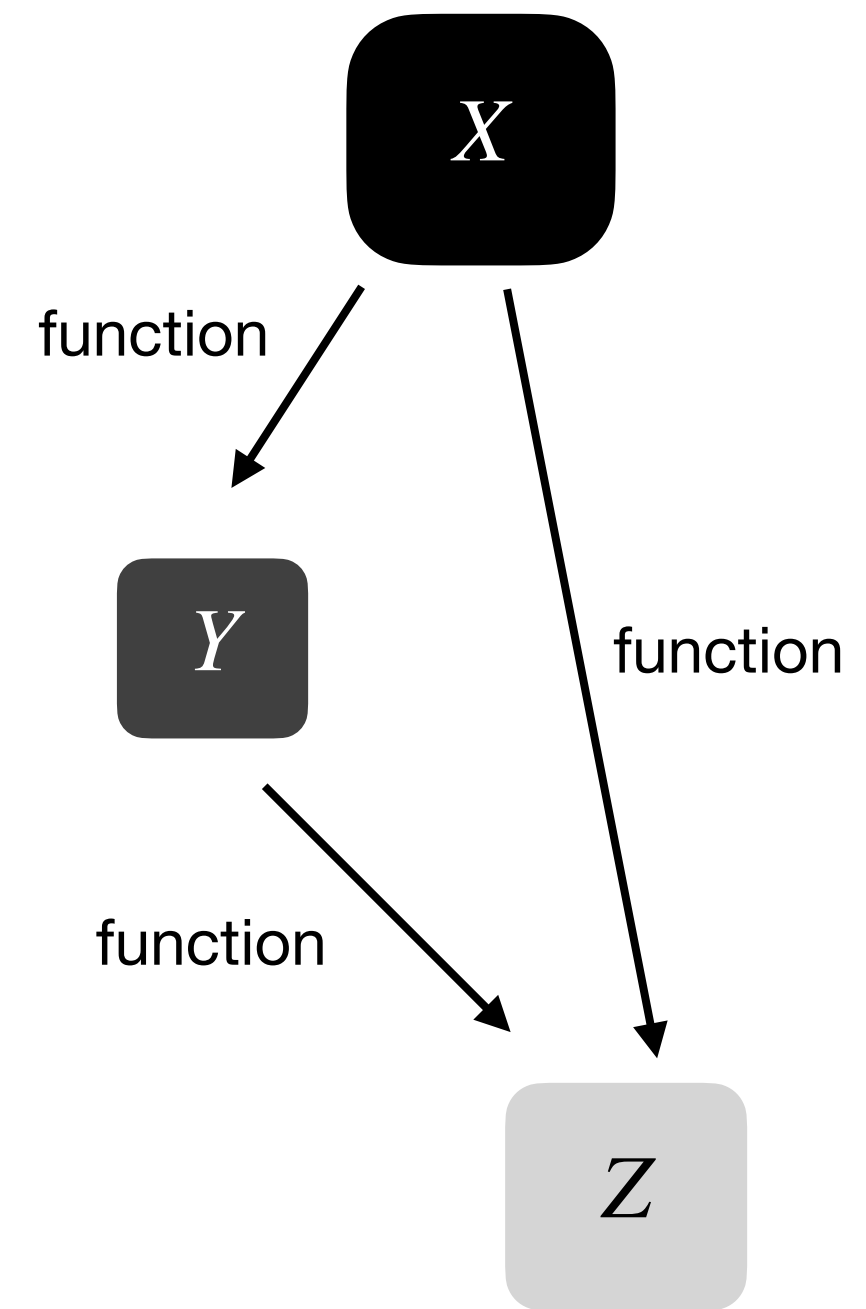
1. introduce **category theory**
2. describe **language** as a category
3. see **what we can do**



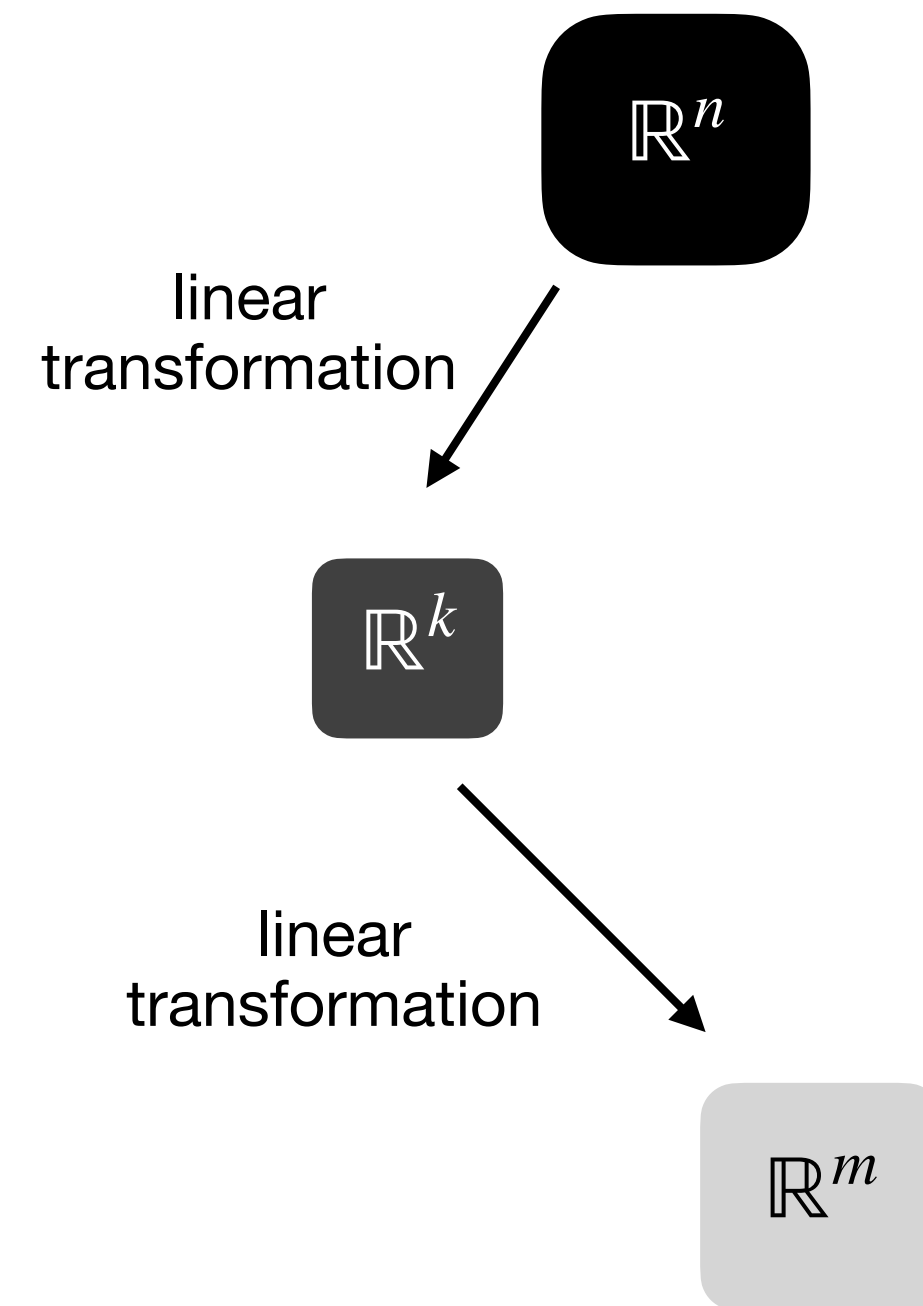
sets



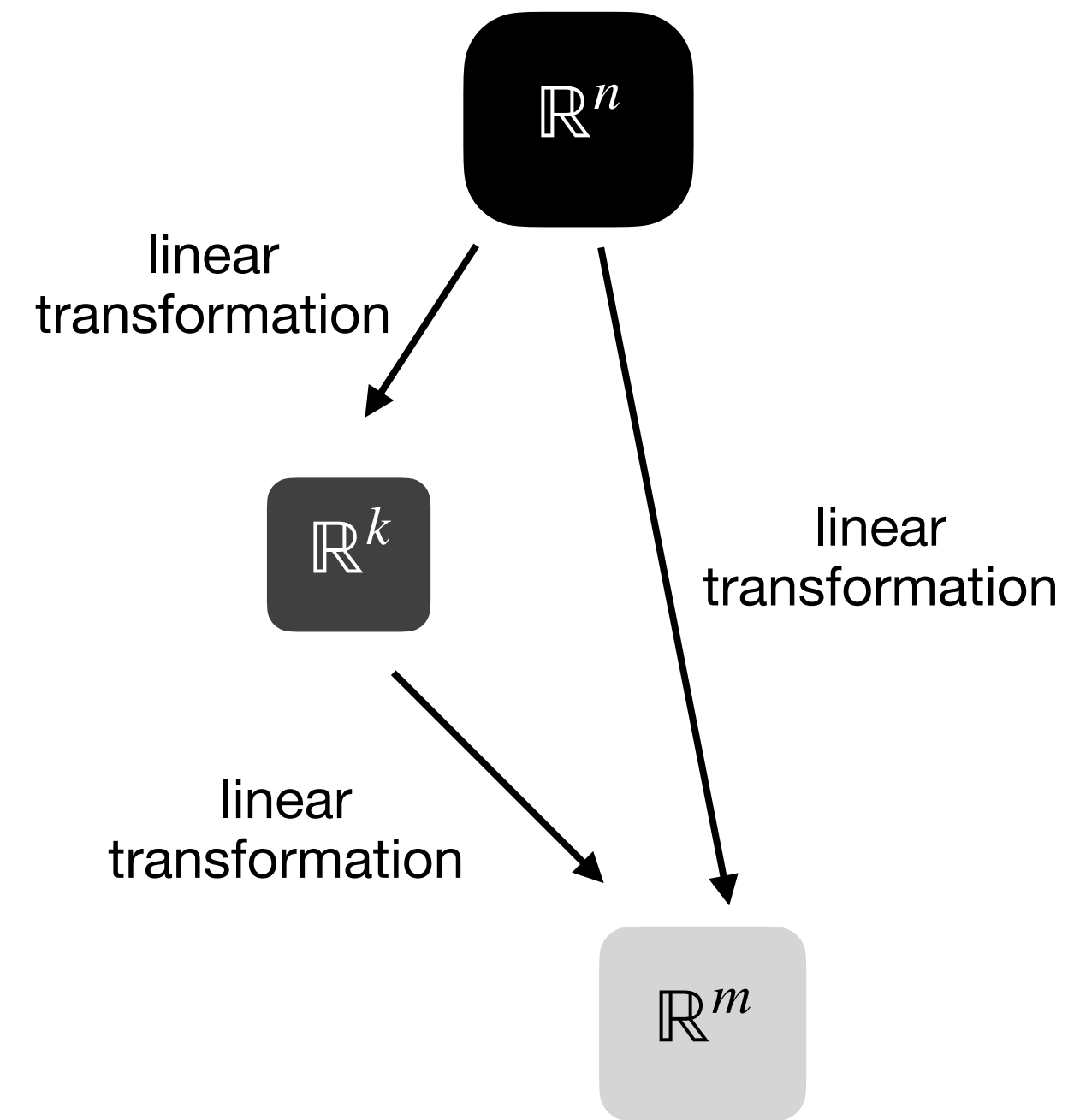
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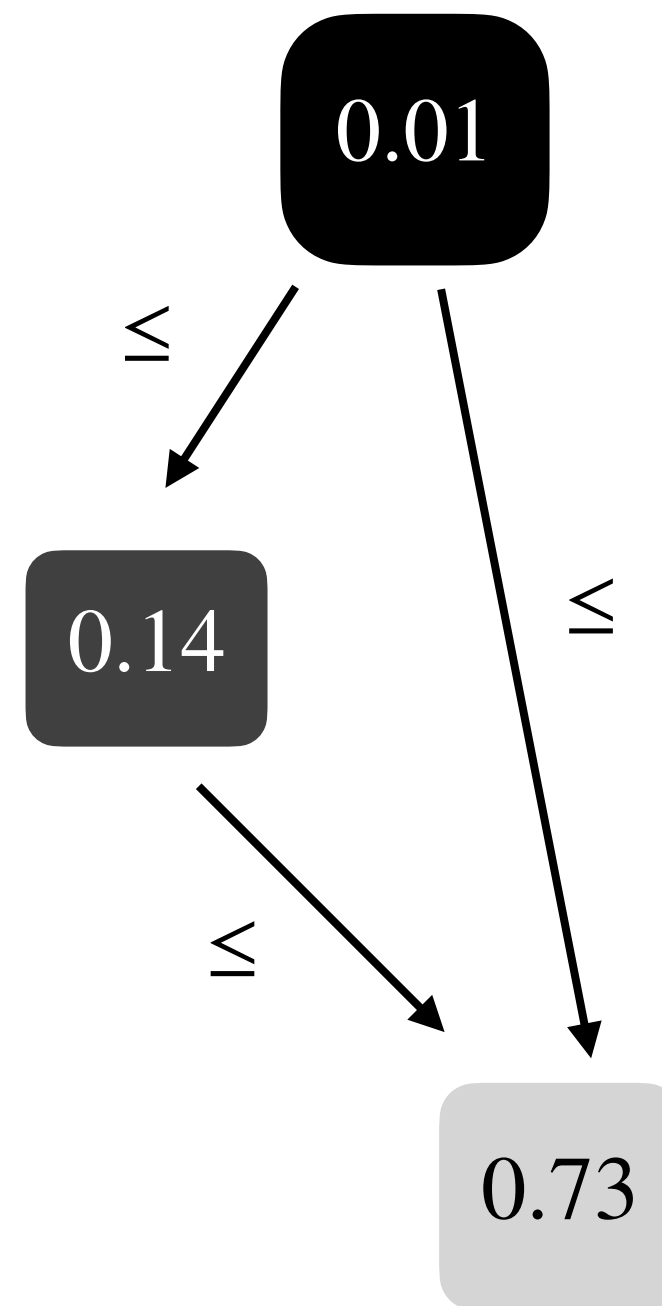
vector spaces



vector spaces



real numbers in $[0,1]$



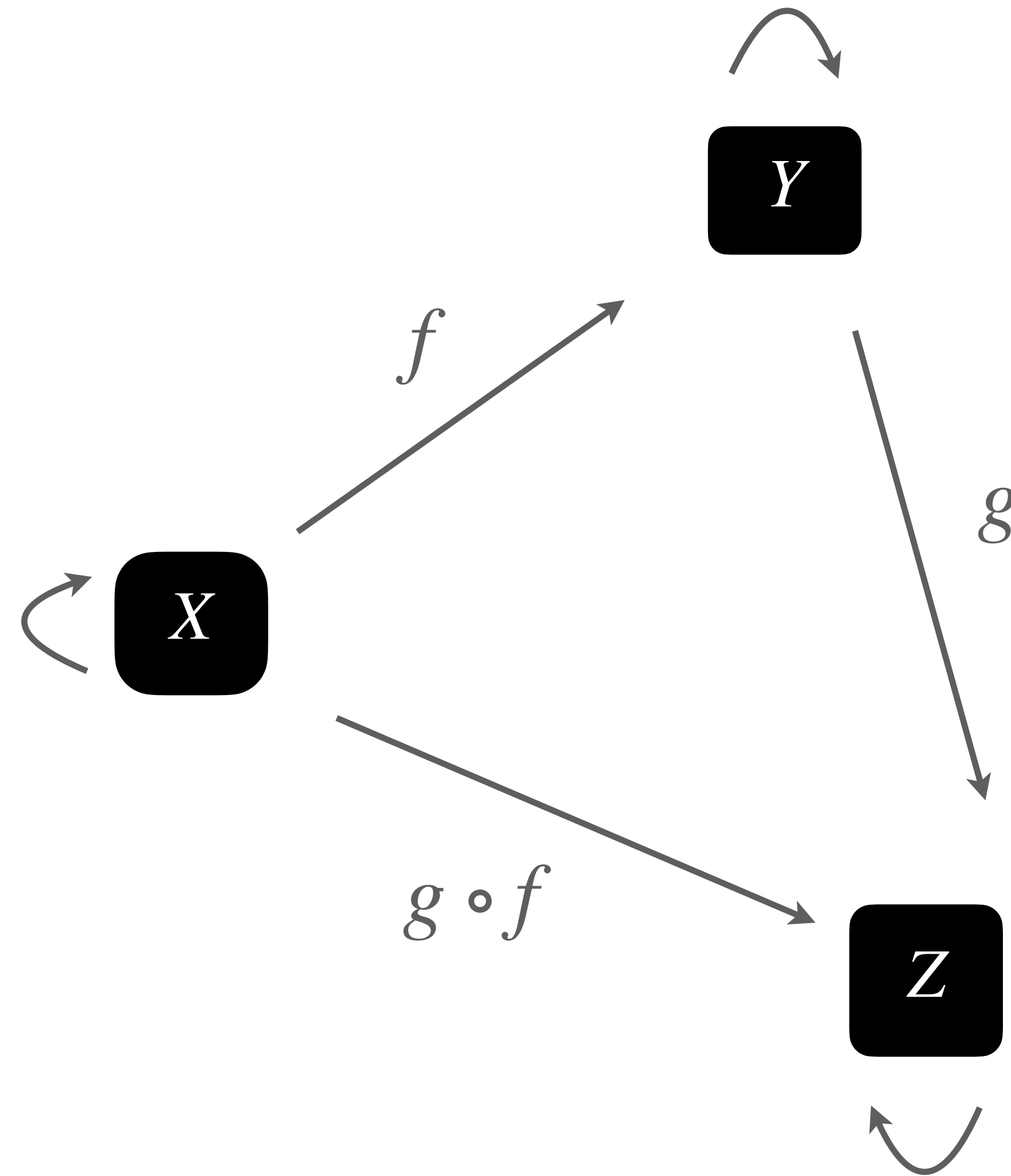
A Category

Loose Definition

A category \mathcal{C} consists of

- **objects** X, Y, \dots
- **morphisms** (i.e. arrows) between them
- a **composition rule**

that satisfy some reasonable axioms.



categories

language

categories

OBJECTS

X, Y, \dots

language

categories

OBJECTS

X, Y, \dots

language

STRINGS

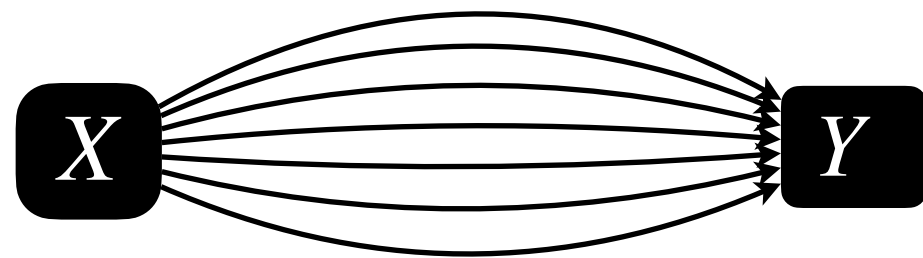
x, y, \dots

categories

OBJECTS

X, Y, \dots

A SET



language

STRINGS

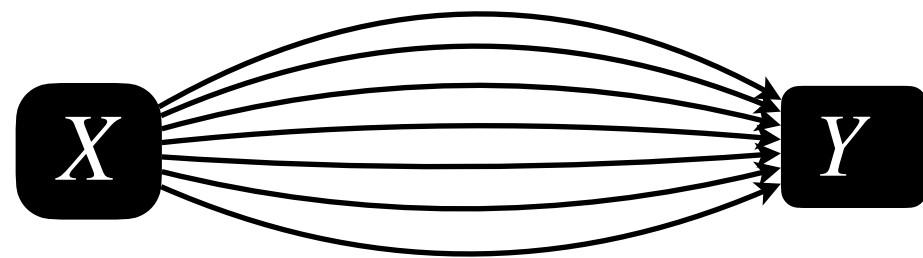
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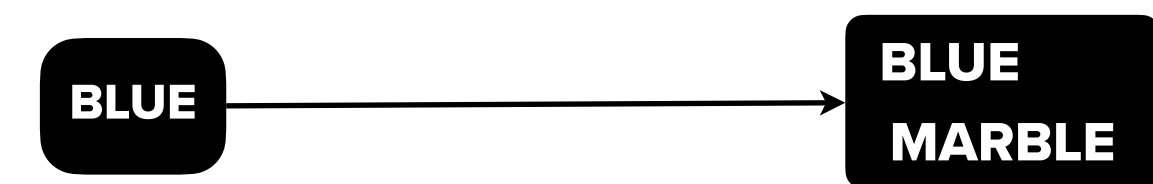


language

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x, y, \dots

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categories

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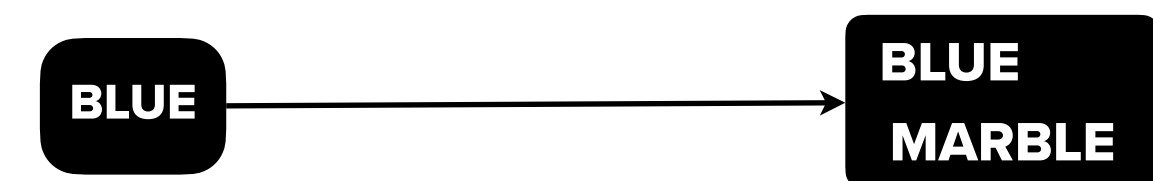
$C(X, Y)$

language

STRINGS

x, y, \dots

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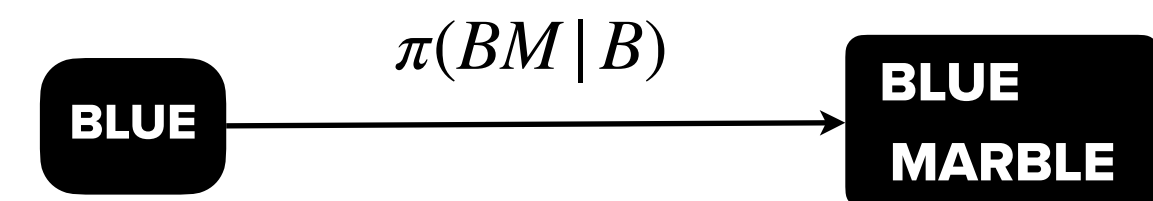
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$\pi(y | x)$

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A COMPOSITION RULE

$$C(X, Y) \times C(Y, Z) \rightarrow C(X, Z)$$

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Language as a Category, L

Summary so far...

Consider all strings x, y, \dots from a finite set of atomic symbols. (*Think: expressions in a language.*)

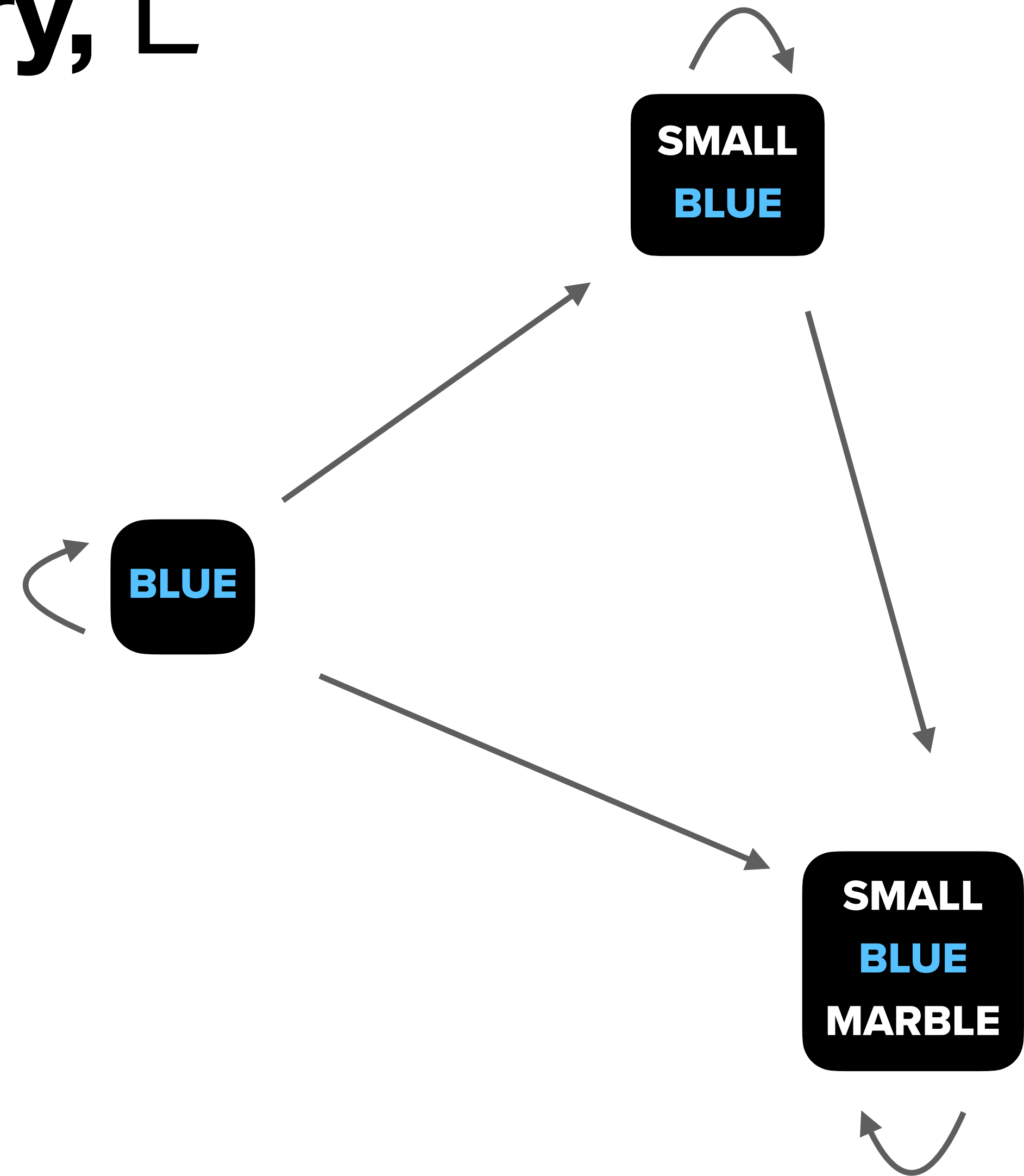
Write $x \rightarrow y$ to indicate **substring containment**.

- Arrows compose:

if $x \rightarrow y \rightarrow z$, then $x \rightarrow z$

- Each string contains itself:

$x \rightarrow x$



1. introduce **category theory**
2. describe **language** as a category
- 3. see what we can do**

1.
Represent
Meanings

2.
Operate on
Representations

3.
Incorporate
Probabilities

4.
Represent
Enriched Meanings

5.
Operate on
Enriched Representations

6.
Adopt a
Geometric Perspective

1.
Represent
Meanings

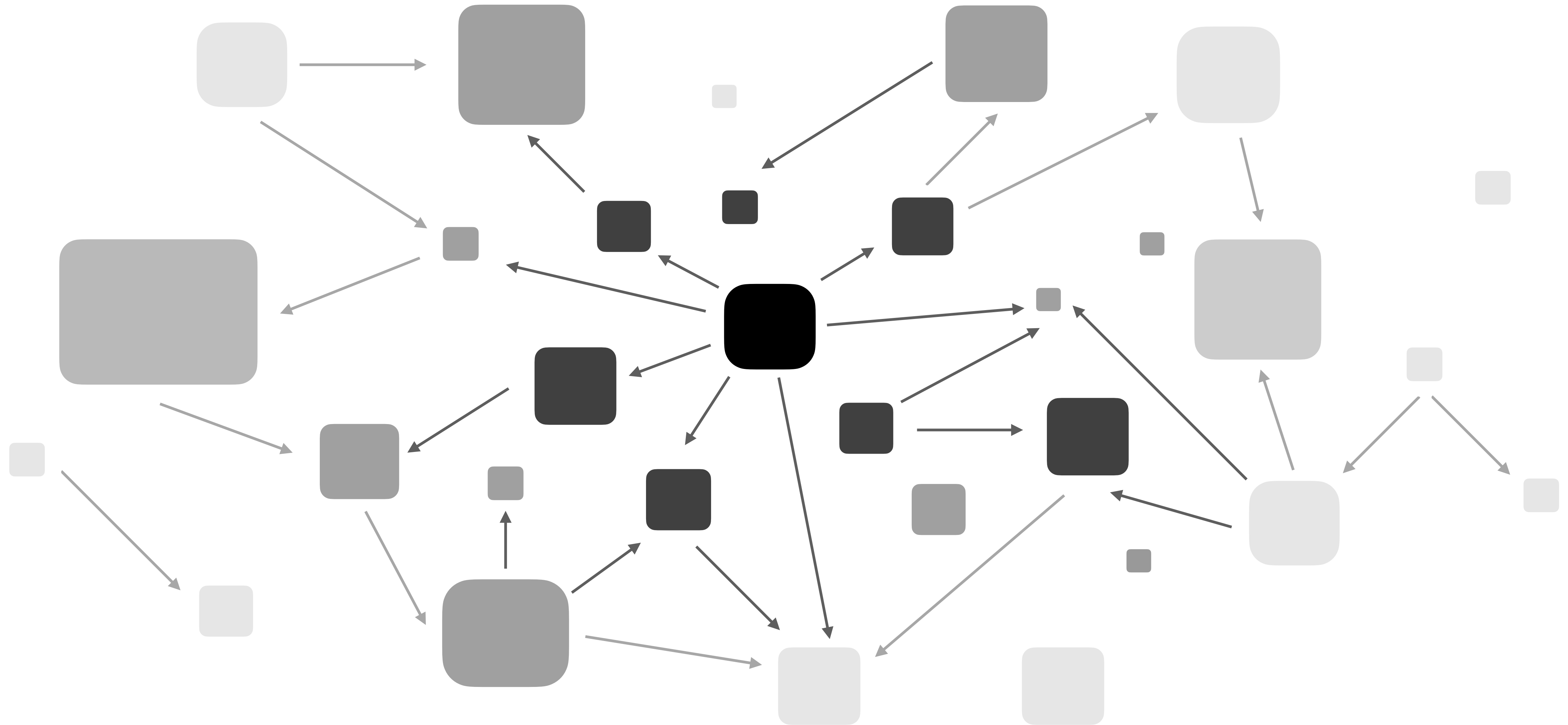
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$C(X, -)$

$C(X, -)$

C

$C(X, -)$
→
functor

Set

$C(X, -)$ C $C(X, -)$
 \longrightarrow
functor

Set

 $X \cong Y$

iff

 $C(X, -) \cong C(Y, -)$ **Yoneda Lemma**

(or rather, a corollary of it)

Idea: Given a “prompt” (i.e. expression) x , consider the network of ways it fits into the language category L .

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Ask: “Is x is contained in a given expression y ?”

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$$x \mapsto L(x, -)$$

$$L(\text{blue, blue marble}) = \{ \rightarrow \}$$

$$L(\text{blue, curiosity killed the cat}) = \emptyset$$

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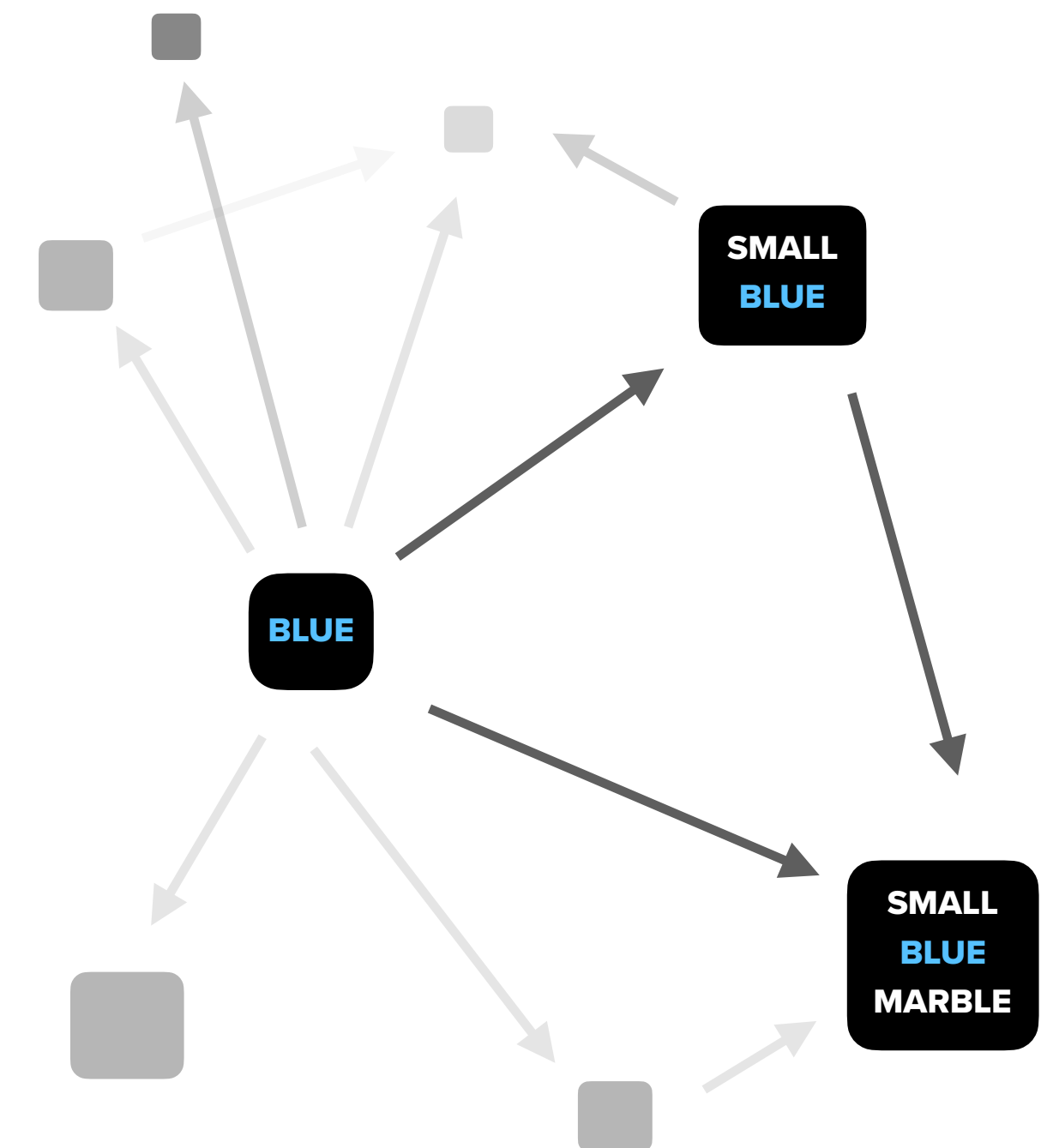
Upshot: The Yoneda Lemma motivates us to consider the functor

$$L(x, -) : L \rightarrow \text{Set}$$

which “represents” (or is a rough approximation to) the meaning of x .

Example:

$$L(\text{blue}, y) = \begin{cases} \{ \rightarrow \} & \text{if blue} \leq y \\ \emptyset & \text{otherwise} \end{cases}$$



$L(\text{blue}, -) \stackrel{\text{SORT OF}}{=} \begin{bmatrix} \emptyset \\ * \\ * \\ \emptyset \\ * \\ \emptyset \\ \vdots \end{bmatrix}$

\emptyset	deep red Bing cherries
*	small blue marble
*	beautiful blue ocean
\emptyset	did you put the kettle on
*	red and blue fireworks
\emptyset	Sencha green tea
\vdots	\vdots

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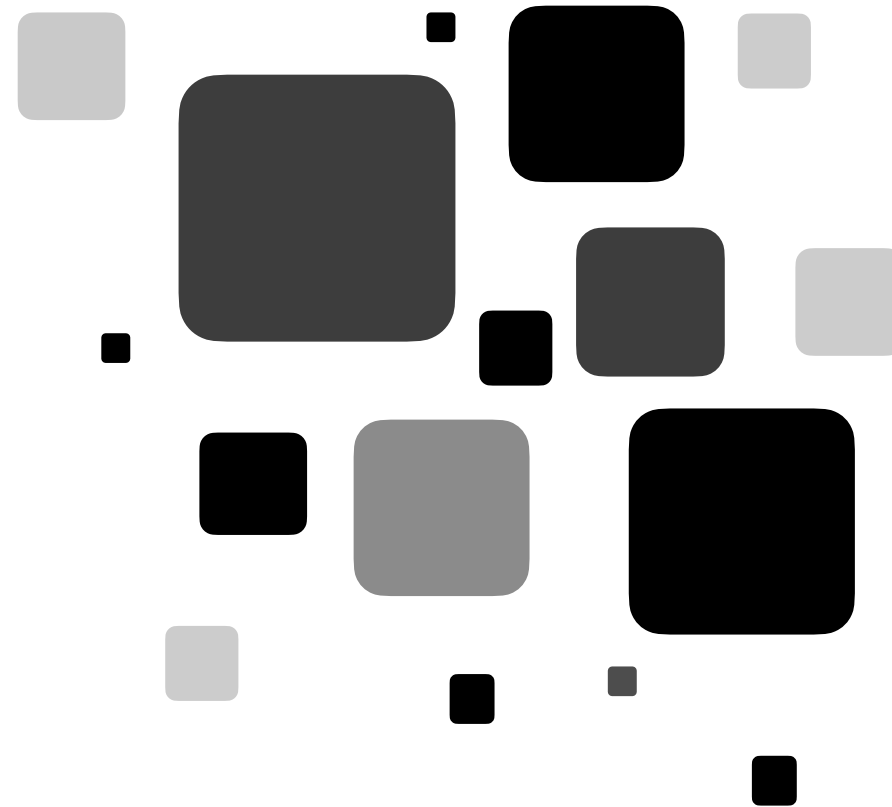
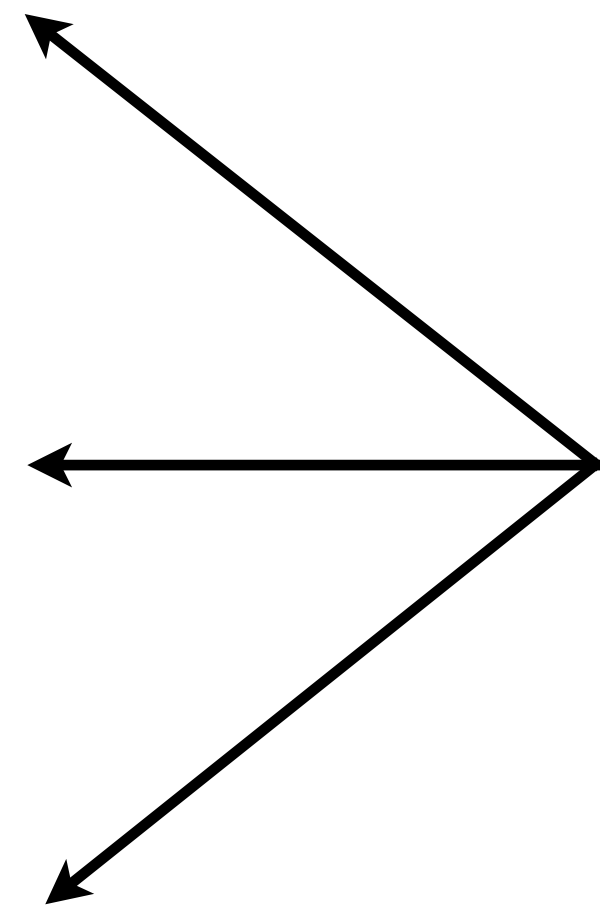
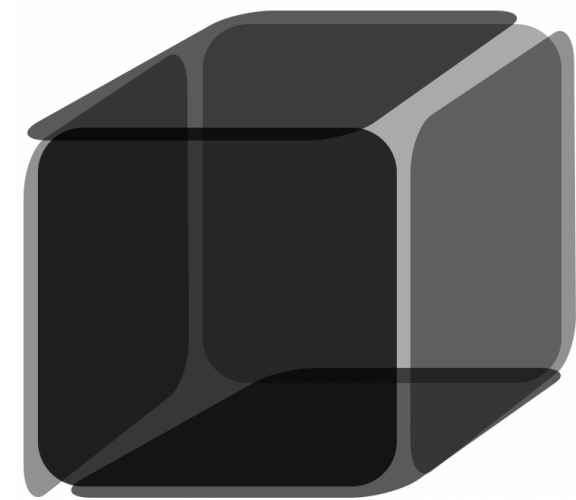
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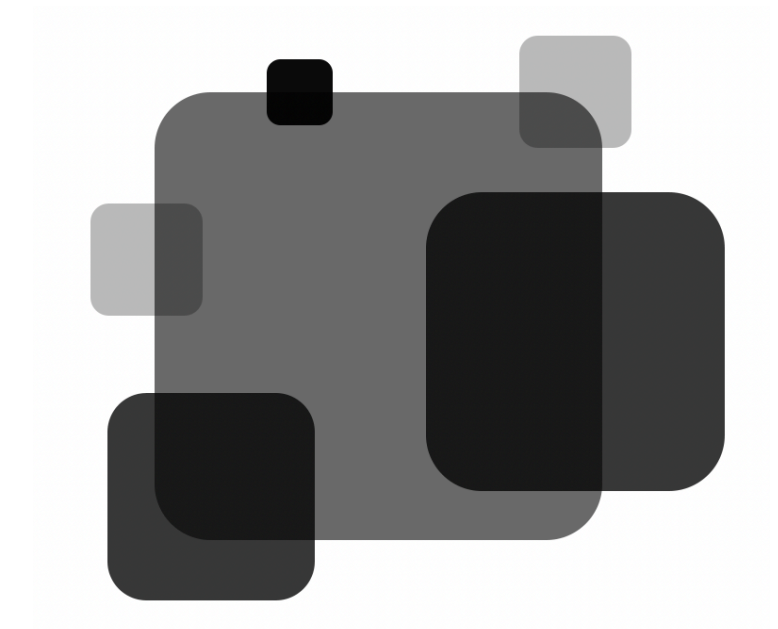
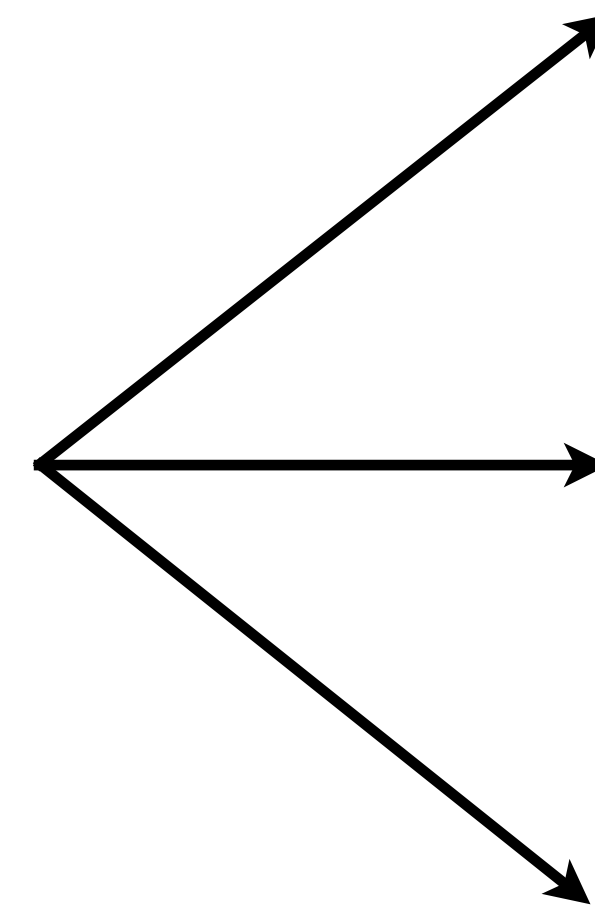
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limits



intersections, products, direct sums, meets, greatest common divisors, kernels,...

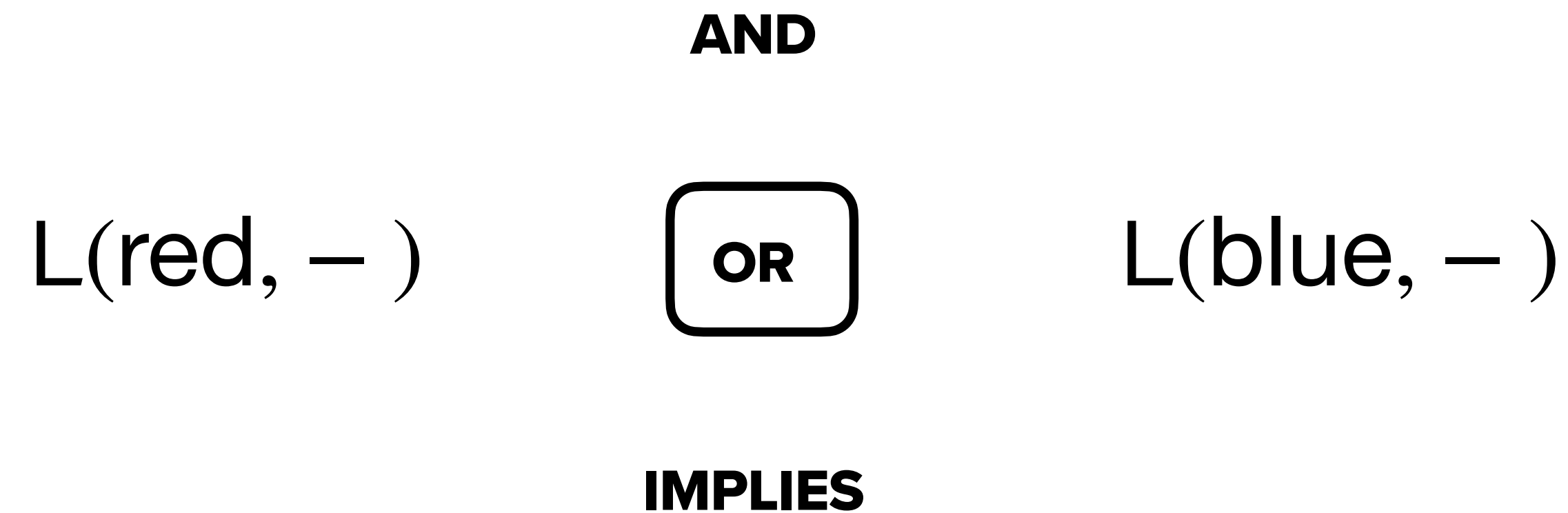
colimits



unions, coproducts, direct sums, joins, least common multiples, cokernels,...

All functors $L \rightarrow \text{Set}$ form a new category Set^L that has *lots* of structure, which it inherits from the category of sets. Just as we can combine sets in many ways (intersections, unions, etc.) we can now combine *functors* in many ways.

Practically speaking, this means we have notions of conjunction, **disjunction**, and implication. (Formally speaking, Set^L has “all limits, colimits, and is Cartesian closed.”)



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$$L(\mathbf{red}, -) \sqcup L(\mathbf{blue}, -) = \begin{array}{c} \text{SORT OF} \\ \left[\begin{array}{l} * \\ * \\ * \\ \emptyset \\ * * \\ \emptyset \\ \vdots \end{array} \right] \begin{array}{l} \text{deep } \mathbf{red} \text{ Bing cherries} \\ \text{small } \mathbf{blue} \text{ marble} \\ \text{beautiful } \mathbf{blue} \text{ ocean} \\ \text{did you put the kettle on} \\ \mathbf{red} \text{ and } \mathbf{blue} \text{ fireworks} \\ \text{Sencha green tea} \\ \vdots \end{array} \end{array}$$

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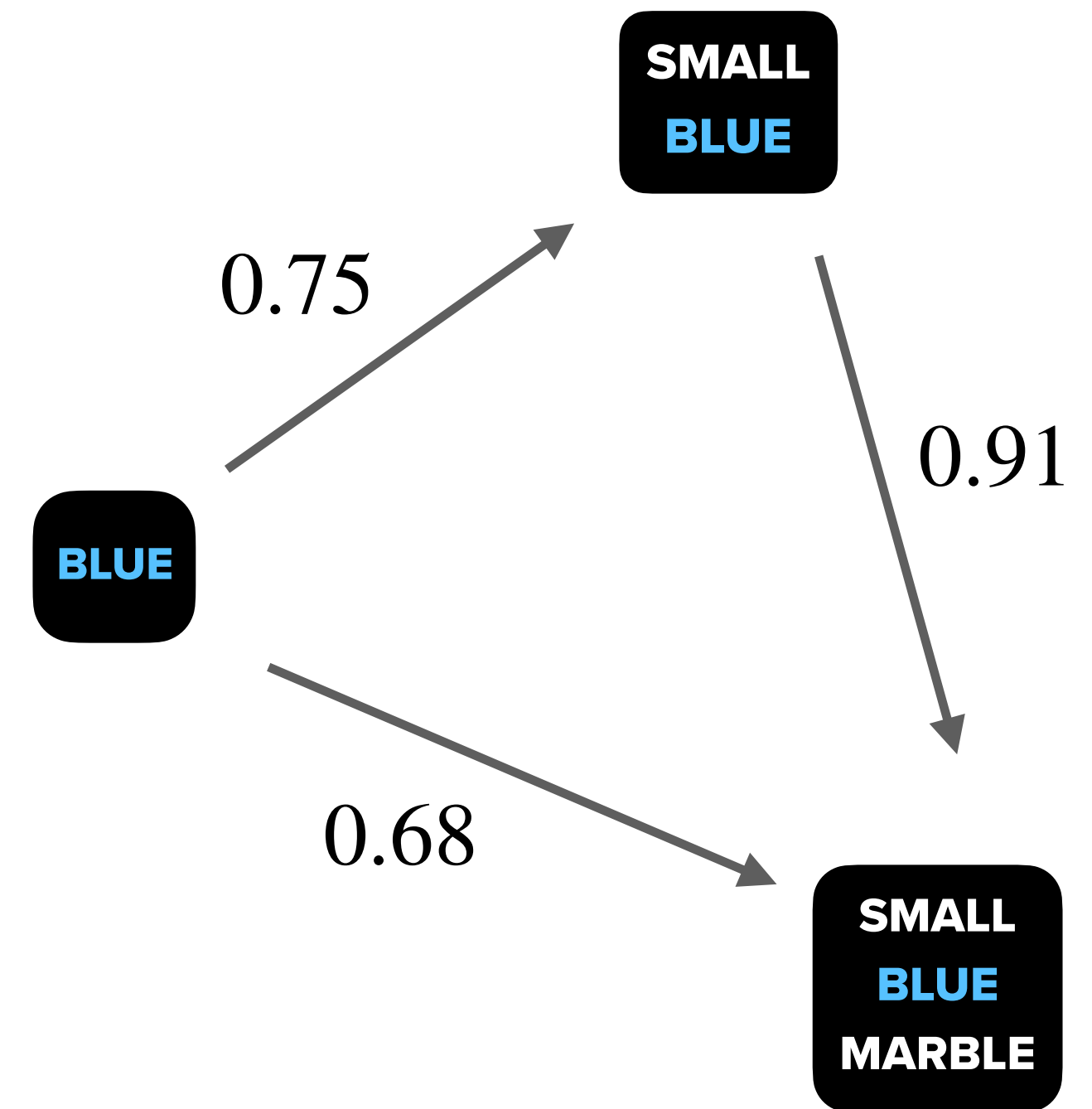
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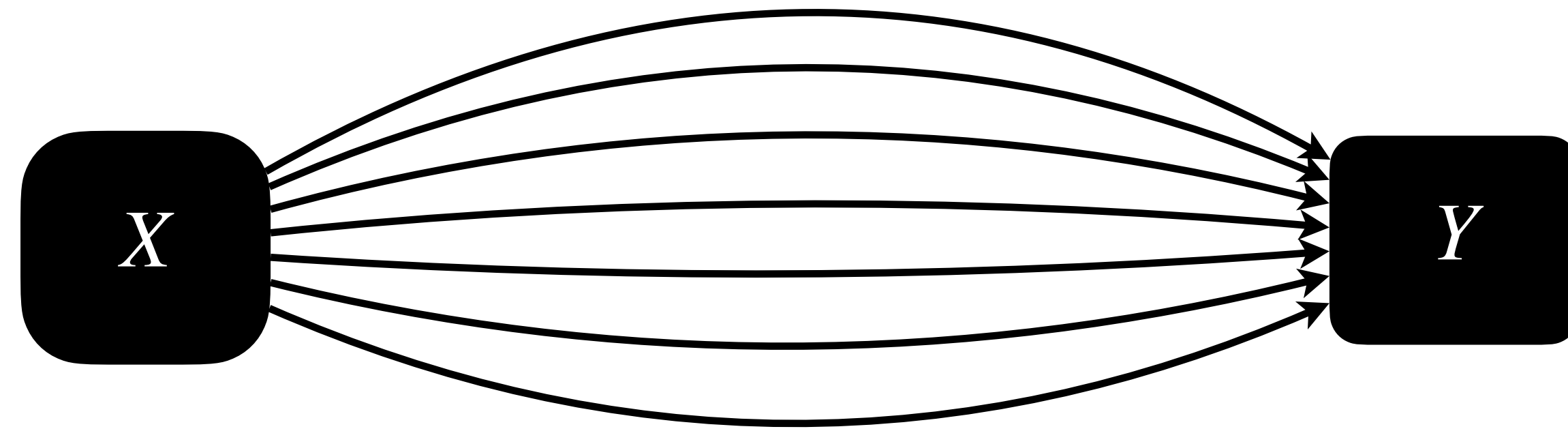
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$$L(\text{blue}, -) \stackrel{\text{SORT OF}}{=} \begin{bmatrix} 0 \\ .68 \\ .01 \\ 0 \\ .17 \\ 0 \\ \vdots \end{bmatrix}$$

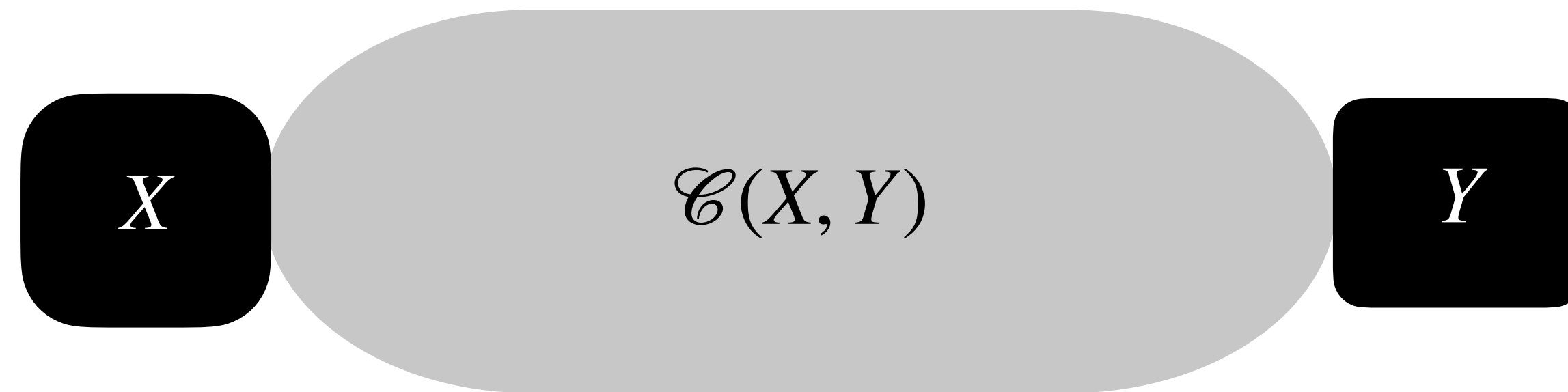
deep red Bing cherries
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⋮





In (ordinary) **category theory**, each pair of objects has an associated **set**

$$\mathbf{C}(X, Y)$$



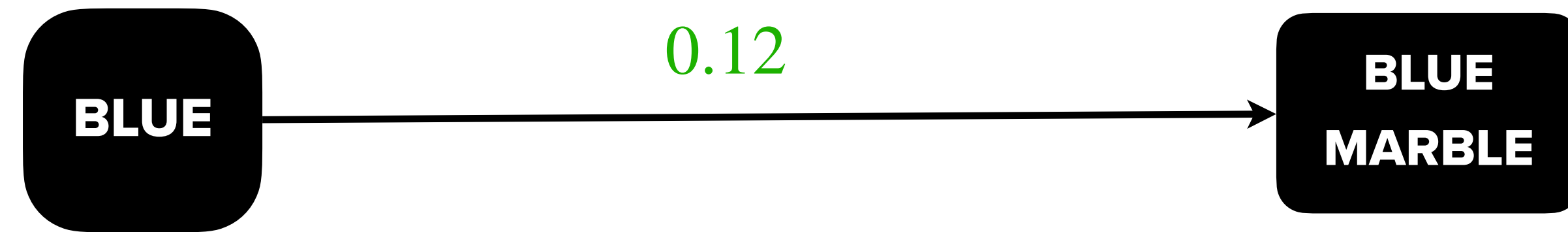
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In **enriched category theory**, each pair of objects has an associated **object**

$$\mathcal{C}(X, Y)$$

in some sufficiently nice category



In (ordinary) **category theory**, each pair of objects has an associated **set**

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In **enriched category theory**, each pair of objects has an associated **object**

$$\mathcal{C}(X, Y)$$

in some sufficiently nice category

like $[0,1]$, as hinted earlier

An Enriched Category

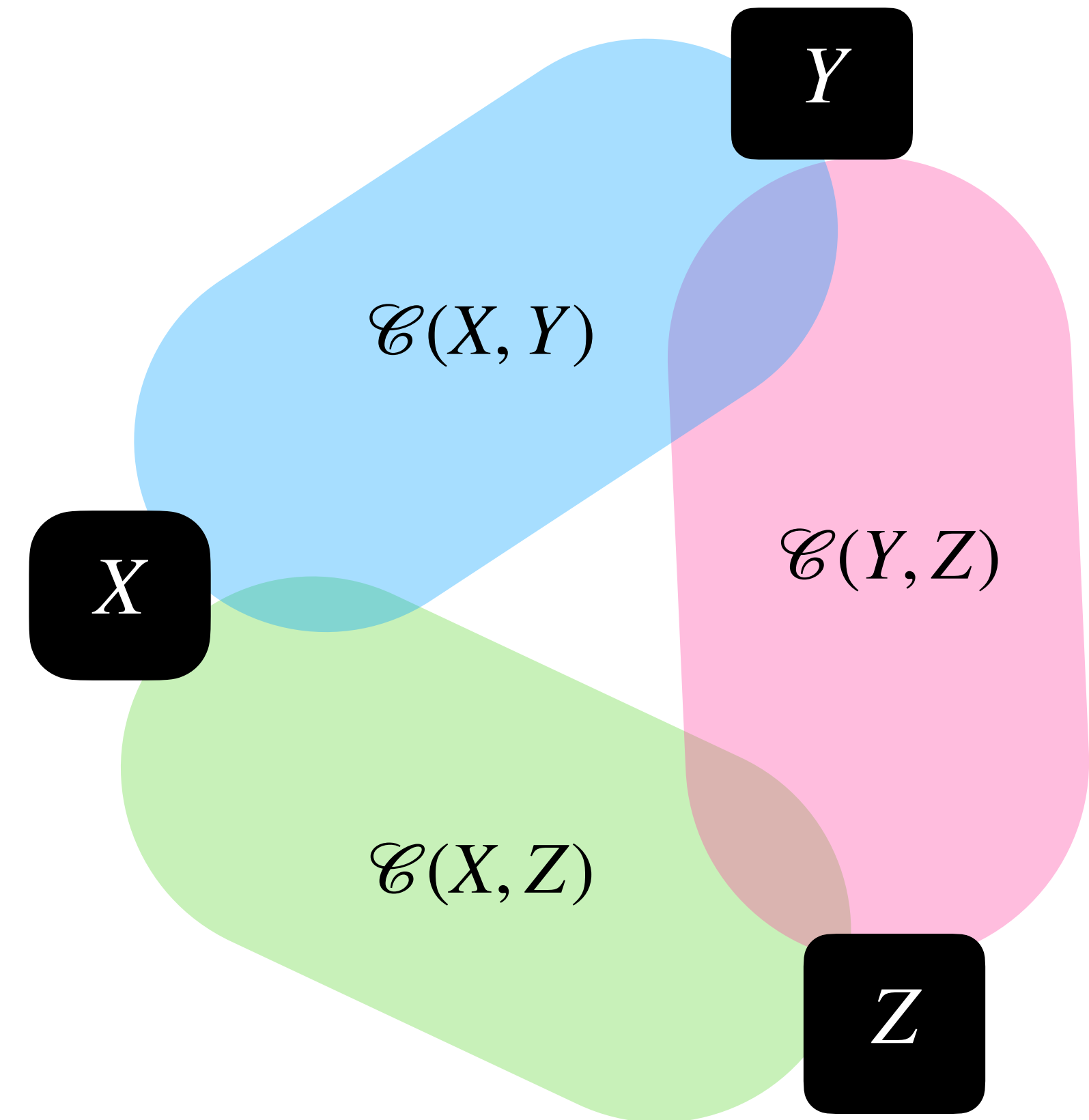
Very Loose Definition

Given a sufficiently nice category \mathcal{V} , a \mathcal{V} –enriched category \mathcal{C} has

- **objects** X, Y, \dots
- an **object** $\mathcal{C}(X, Y)$ in \mathcal{V}
- a “**composition rule**”

$$\mathcal{C}(X, Y) \otimes \mathcal{C}(Y, Z) \rightarrow \mathcal{C}(X, Z)$$

that satisfy reasonable axioms.



An Enriched Category

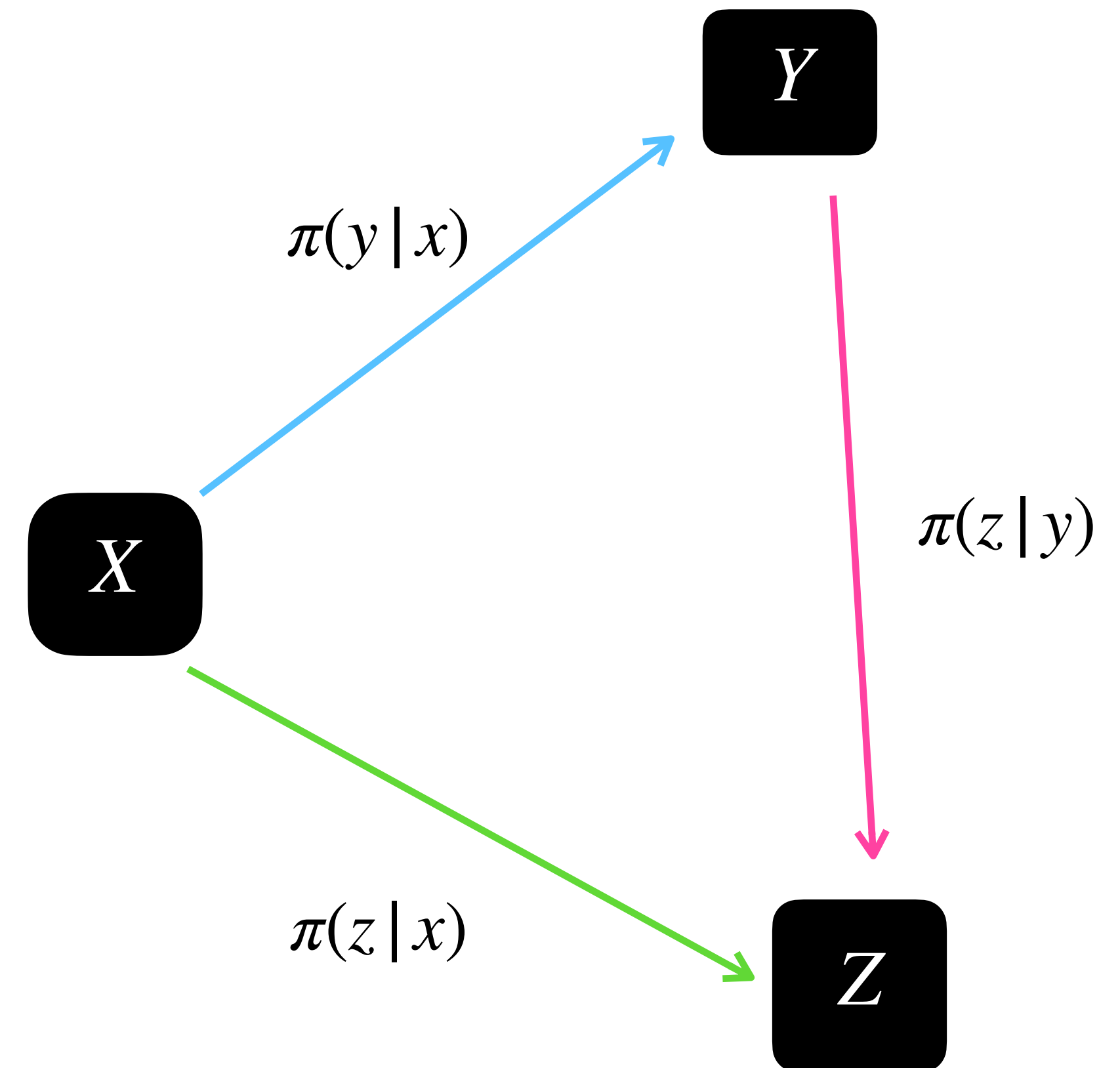
Very Loose Definition

We are interested in the case when the base category is the unit interval $[0,1]$.

We want:

- strings x, y, \dots
- a probability of continuation $\pi(y | x)$
- But do we have this inequality?

$$\pi(y | x) \cdot \pi(z | y) \leq \pi(z | x)$$



Yes, suppose we have an LLM.

For any prompt x , the LLM gives a probability distribution $p(\cdot | x)$ on the set of tokens.

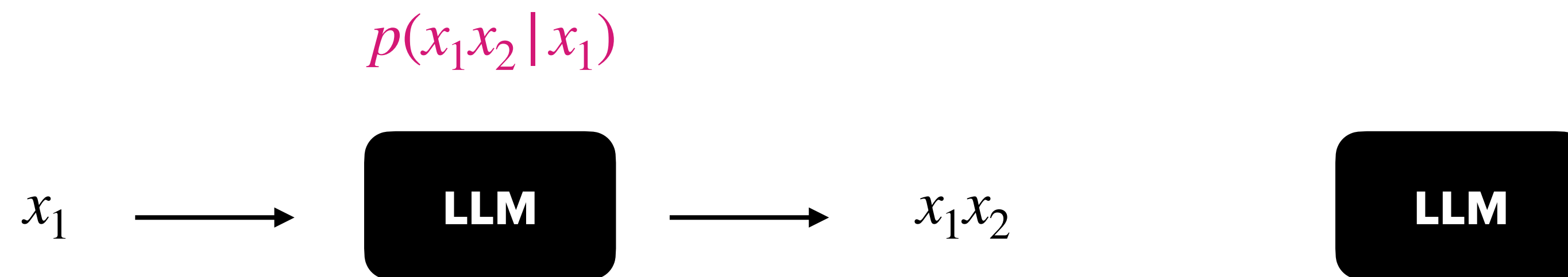
These probabilities multiply in the following sense:



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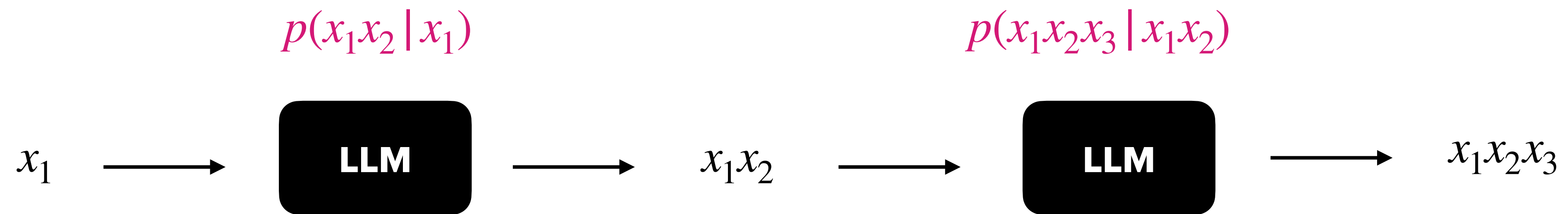
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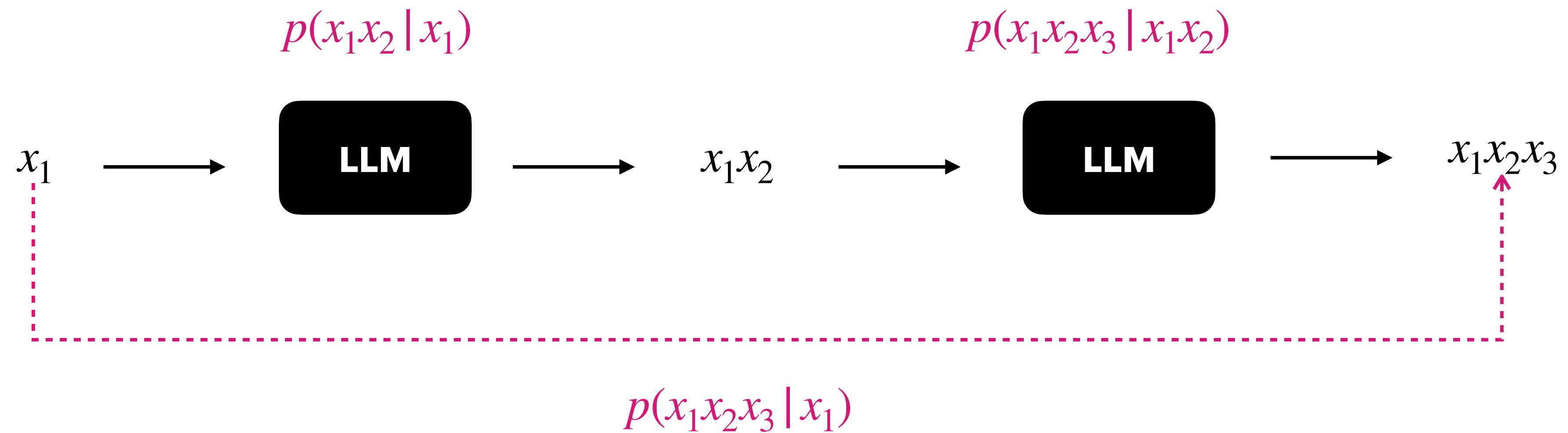
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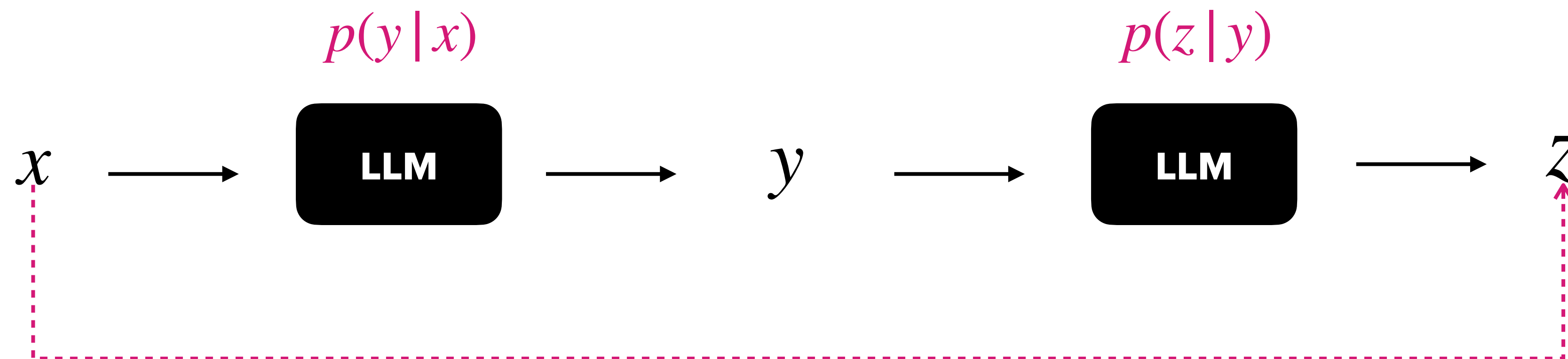
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These probabilities multiply in the following sense:



$$p(z|x) = p(y|x)p(z|y)$$

(so we get an *equality*, in fact)

Language as an Enriched Category, \mathcal{L}

Over the Unit Interval

Given an LLM and strings $x \rightarrow y$, define* the number $\pi(y | x)$ as a product of the successive probabilities used to obtain y from x one token at a time:

$$\pi(y | x) := \begin{cases} 1 & \text{if } x = y \\ 0 & \text{if } x \not\rightarrow y \\ \prod_{i=1}^{k(y)} p(x_{t+i} | x_{<t+i}) & \text{if } x \rightarrow y \end{cases}$$

* Thanks to Juan Pablo Vigneaux for this observation. In this definition, we write $x \rightarrow y$ whenever y extends x on the right.

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This number is an object in the category $[0, 1]$, and it satisfies the “**composition rule**”

$$\pi(y | x) \cdot \pi(z | y) = \pi(z | x).$$

for all strings x, y, z .

So, we view **language as a category \mathcal{L} enriched over $[0, 1]$** .

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categories

OBJECTS

X, Y, \dots

A SET

$C(X, Y)$

A COMPOSITION RULE

$$C(X, Y) \times C(Y, Z) \rightarrow C(X, Z)$$

language

STRINGS

x, y, \dots

A NUMBER

$\pi(y | x)$

A COMPOSITION RULE

$$\pi(y | x) \cdot \pi(z | y) = \pi(z | x)$$

1.
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6.
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Geometric Perspective

4. Represent Enriched Meanings

Consider enriched functors $\mathcal{L} \rightarrow [0,1]$ associated to expressions. These contain the same information as before, plus probabilities.

Ex: The functor $\mathcal{L}(\text{blue}, -)$ is supported on all texts that contain “blue.”

$$\mathcal{L}(\text{blue}, -) = \begin{array}{l} \text{SORT} \\ \text{OF} \\ \left[\begin{array}{l} 0 \\ .22 \\ .73 \\ 0 \\ .07 \\ 0 \\ \vdots \end{array} \right] \end{array} \begin{array}{l} \text{deep red Bing cherries} \\ \text{small } \mathbf{blue} \text{ marble} \\ \text{beautiful } \mathbf{blue} \text{ ocean} \\ \text{did you put the kettle on} \\ \text{red and } \mathbf{blue} \text{ fireworks} \\ \text{Sencha green tea} \\ \vdots \end{array}$$

$\pi(\text{Sencha green tea} \mid \text{blue})$

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		$\begin{bmatrix} 0 \\ .22 \\ .73 \\ 0 \\ .07 \\ 0 \\ \vdots \end{bmatrix}$	\vdots

$\pi(\text{Sencha green tea} \mid \text{blue})$

5. Operate on Enriched Representations

The enriched functor category $[0,1]^{\mathcal{L}}$ has rich structure, including the *enriched* versions of limits, colimits, and Cartesian closure.

So, we can again make sense of logical operations like conjunction, disjunction, and implication.

$$\mathcal{L}(\text{red}, -) \sqcup \mathcal{L}(\text{blue}, -) = \begin{matrix} \text{SORT} \\ \text{OF} \\ \begin{bmatrix} .10 \\ .22 \\ .73 \\ 0 \\ .59 \\ 0 \\ \vdots \end{bmatrix} \end{matrix}$$

$\max\{\mathcal{L}(\text{red}, y), \mathcal{L}(\text{blue}, y)\}$

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We can work with *distances* instead of *probabilities* by considering the function

$$-\ln : [0, 1] \rightarrow [0, \infty].$$

The **distance** between expressions x and y is defined by

$$d(x, y) = -\ln \pi(y | x).$$

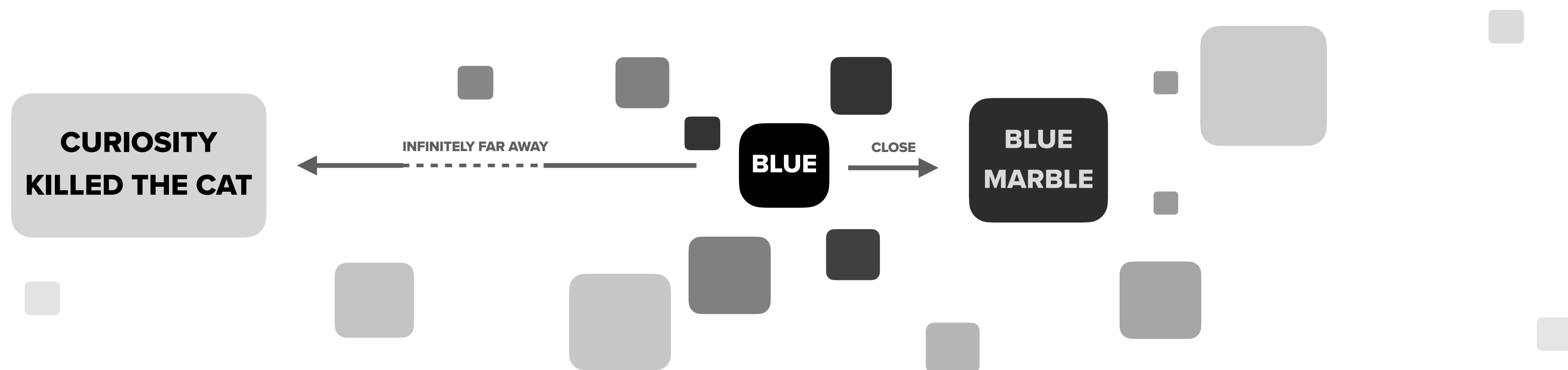
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The **distance** between expressions x and y is defined by

$$d(x, y) = -\ln \pi(y | x).$$

Likely continuations of a text x are **close** to it. Other texts that are not continuations are **infinitely far away**.



Repeat the story all over again.

Distances satisfy a “**composition rule.**”
In fact, it is enriched category theory all
over again!

$$d(x, y) + d(y, z) \geq d(x, z)$$

A $[0, \infty]$ -enriched category is also called
a **generalized metric space**, and we can
compute the versions of the previous
constructions:

- **represent** meanings as “vectors”
(i.e. enriched functors)
- **combine** those representations using
enriched categorical operations

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What *else* do we gain from a geometric perspective?

- Stéphane Gaubert (INRIA) and Yiannis Vlassopoulos (IHES,
ARC) recently interpreted this generalized metric space
through the lens of **tropical geometry**.

*The $[0, \infty]$ -category of language can be viewed as a
polyhedron, with a geometric interpretation of the
“meaning” of texts as generating this polyhedron.*

DIRECTED METRIC STRUCTURES ARISING IN LARGE
LANGUAGE MODELS

STÉPHANE GAUBERT AND YIANNIS VLASSOPOULOS

arXiv: 2405.12264

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- Juan Pablo Vigneaux (Caltech) recently computed the
magnitude of (a finite version of) this generalized metric space.

*Magnitude is a numerical invariant for finite enriched
categories.*

*You can rescale via a parameter t to obtain a **magnitude
function**, which is even more interesting.*

*For our $[0, \infty]$ -enriched category of language, the magnitude
function is a sum over prompts of Tsallis entropies:*

$$|t\mathcal{L}| = (t - 1) \sum_{x \in \text{ob}(\mathcal{L})} H_t(p(-|x))$$

$$x \left[\begin{array}{c} y \\ \vdots \\ \dots e^{-td(x,y)} \end{array} \right]$$

used for
magnitude

$$x \begin{bmatrix} \dots & y \\ \dots & \vdots \\ \dots & e^{-td(x,y)} \end{bmatrix}$$

used for
magnitude

$$x \begin{bmatrix} \dots & y \\ \dots & \vdots \\ \dots & \text{pmi}(x, y) \end{bmatrix}$$

used with
SVD

$$x \begin{bmatrix} \dots & y \\ \dots & \vdots \\ \dots & e^{-td(x,y)} \end{bmatrix}$$

used for
magnitude

$$x \begin{bmatrix} \dots & y \\ \dots & \vdots \\ \dots & \text{pmi}(x, y) \end{bmatrix}$$

used with
SVD

$$x \begin{bmatrix} \dots & y \\ \dots & \vdots \\ \dots & 0, 1 \end{bmatrix}$$

used in
formal concept analysis

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Thanks

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