

CATEGORIES OF RELATIONS WHICH COMPOSE INDEPENDENTLY

— CT 2025 —

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Joint work with
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JS Lemay D. Stein

1. MOTIVATION

2. $\perp = \perp\!\!\!\perp$

3. THE EQUIVALENCE

1. MOTIVATION



M. Di Meglio

Hilbert spaces
(& other things)



C. Heunen

Hilbert spaces
(& other things)



JS Lemay

Restriction cat.s
(& other things)



P. Pennone

Probability
(& other things)



D. Stein

Probability
(& other things)

1. MOTIVATION · Markov Chains



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Idea. Stochastic maps are "like relations which compose independently".

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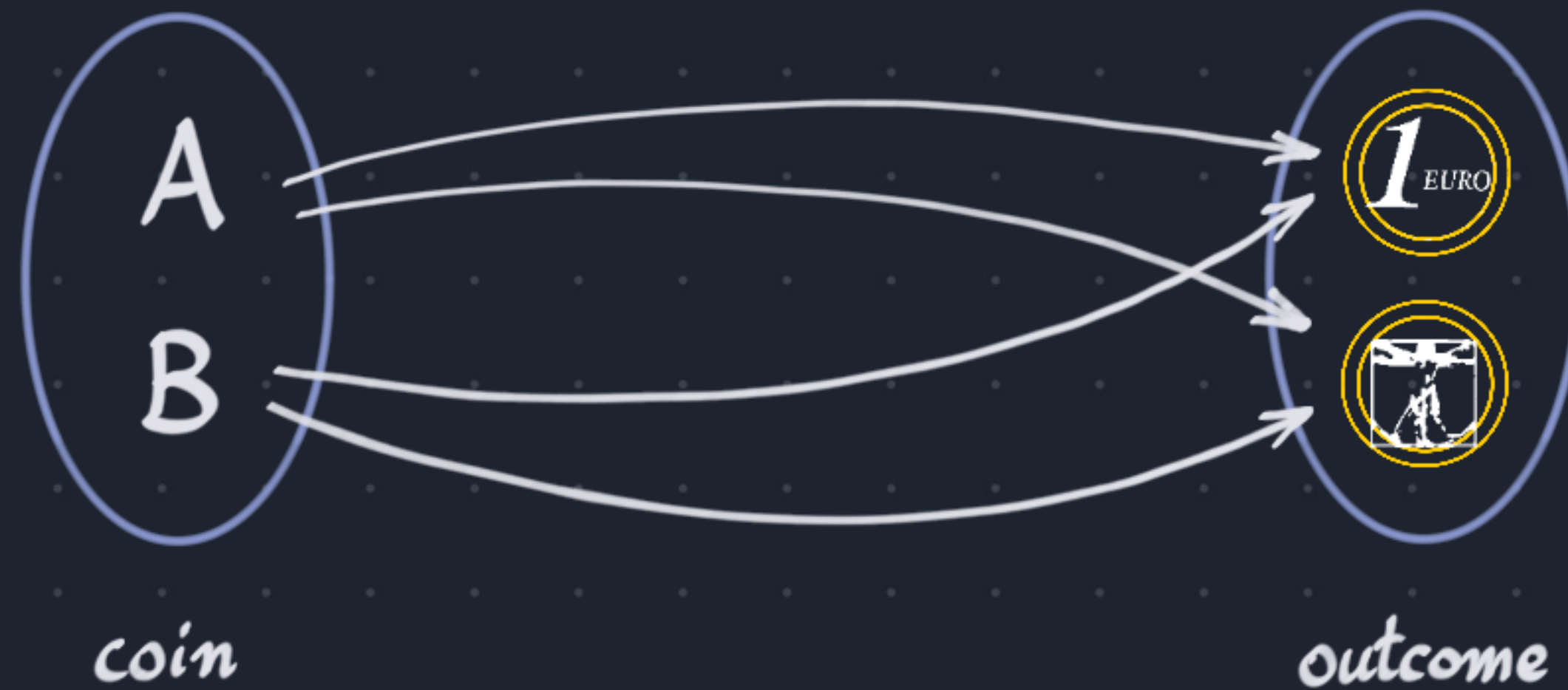
Ordinary relations:



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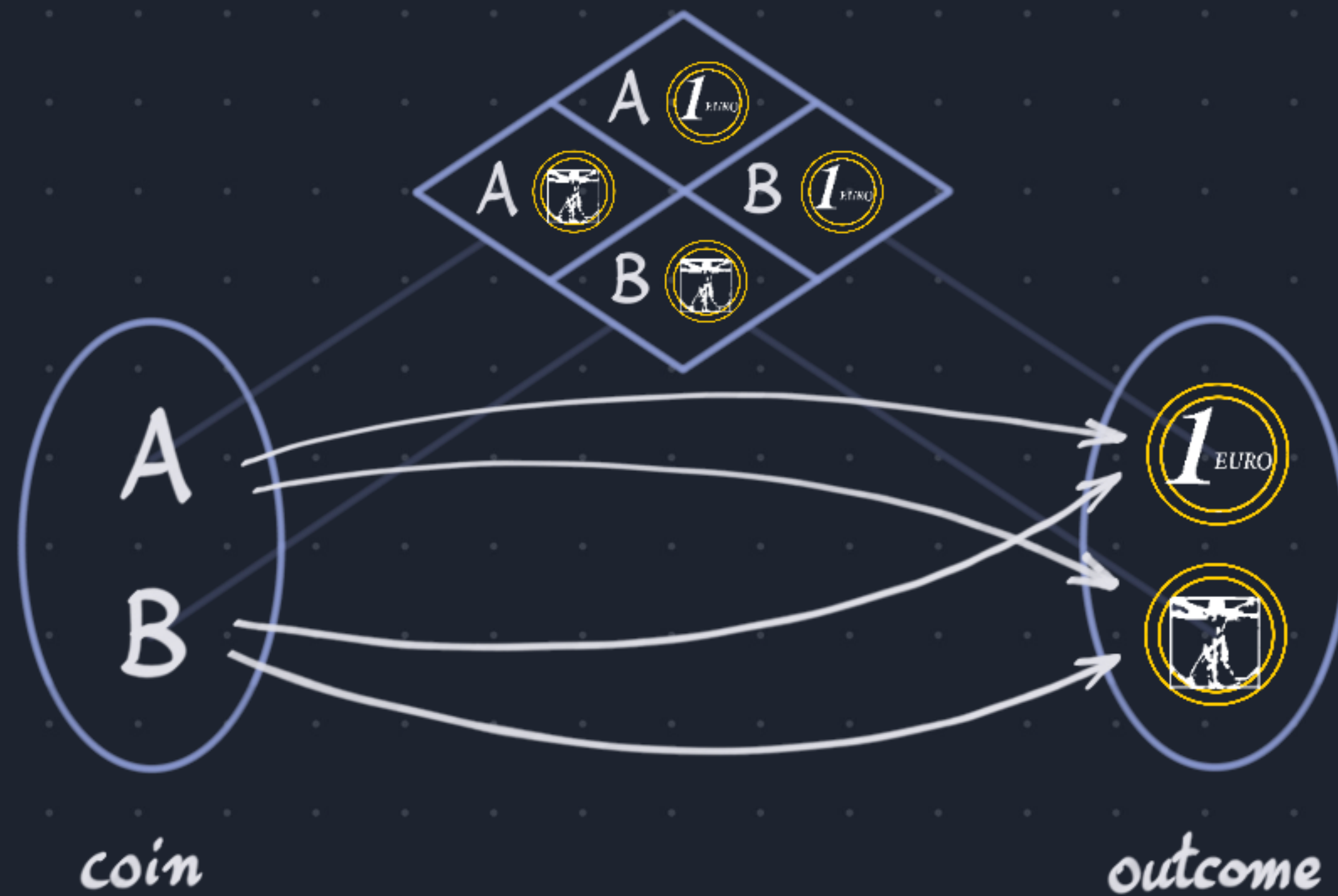
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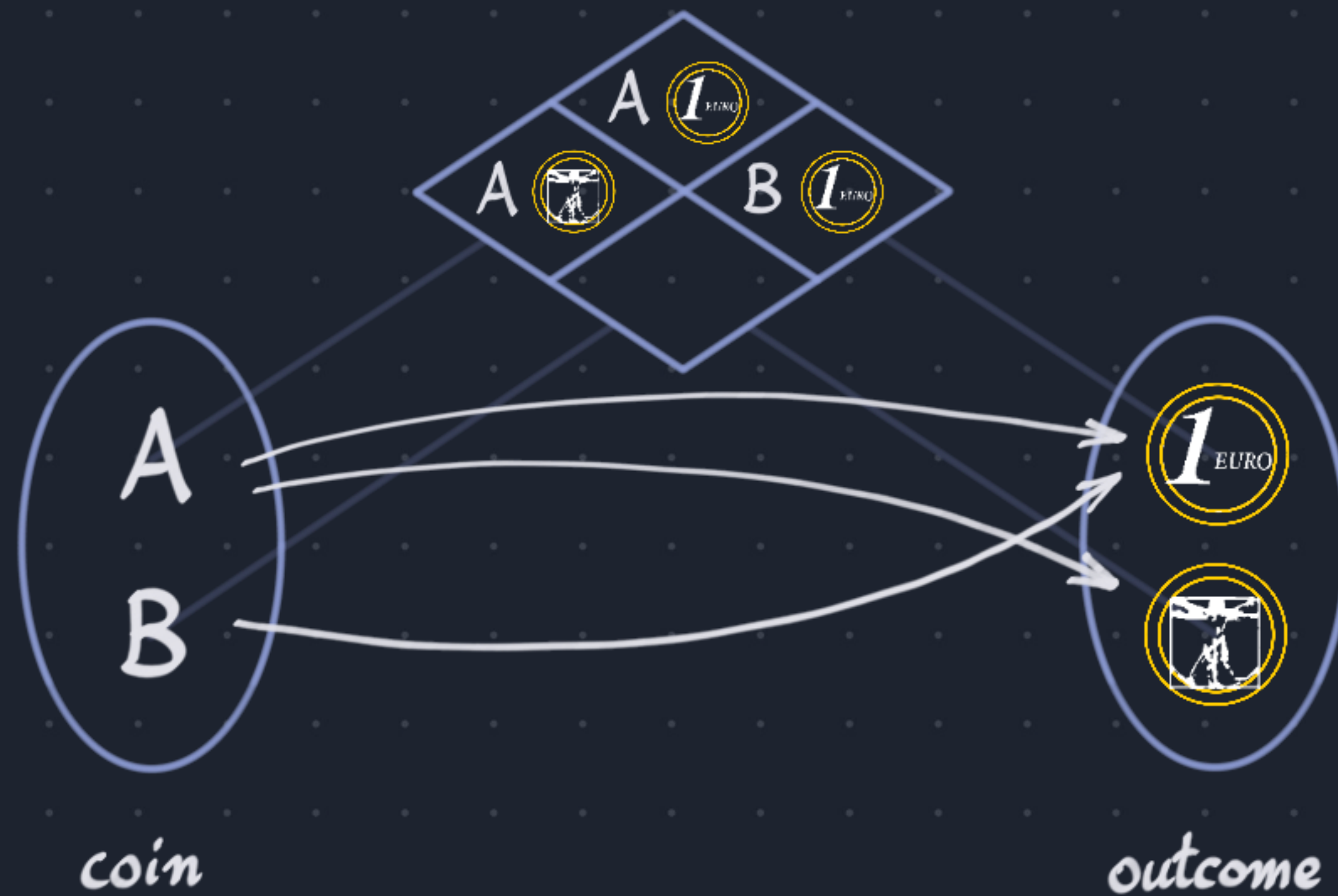
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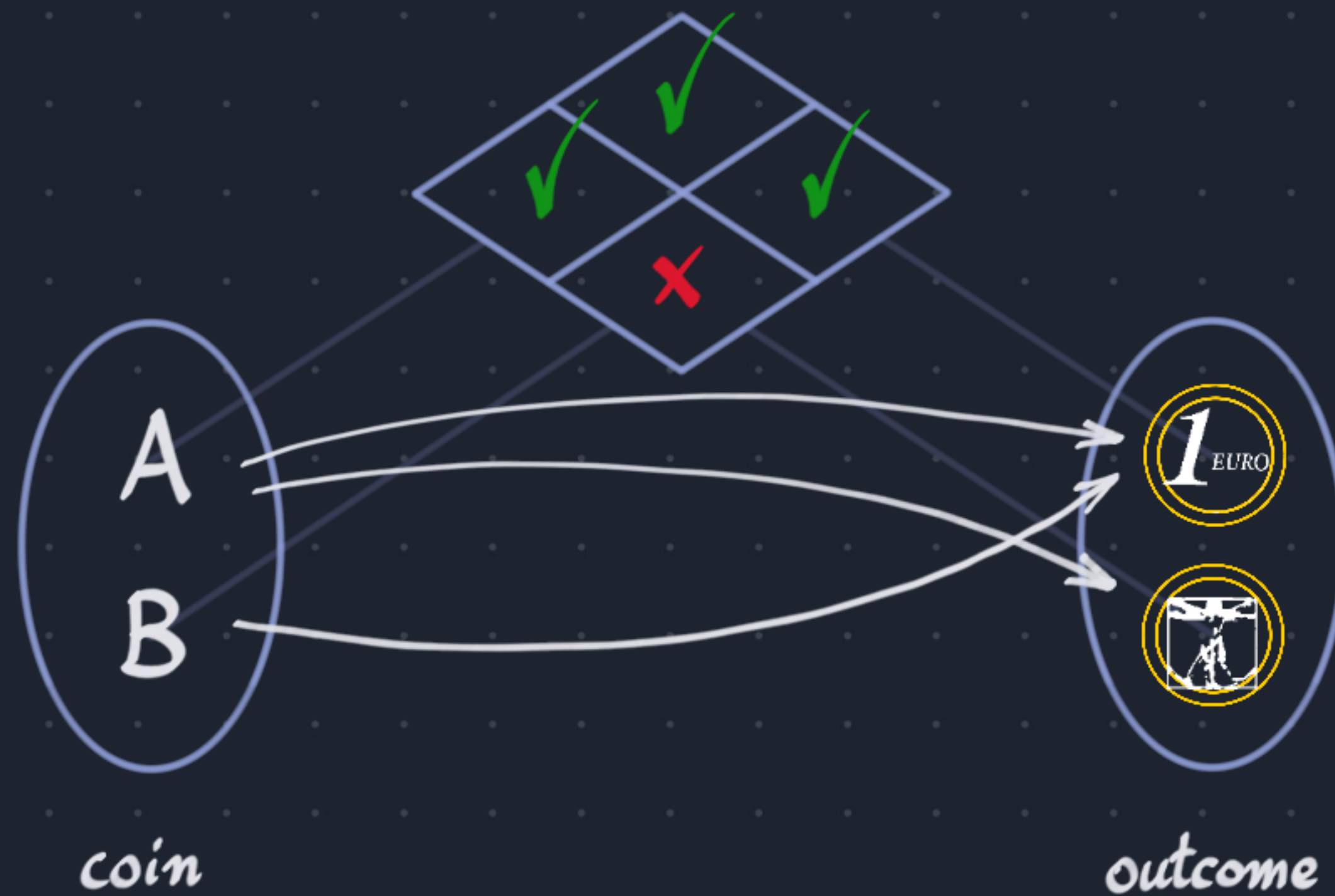
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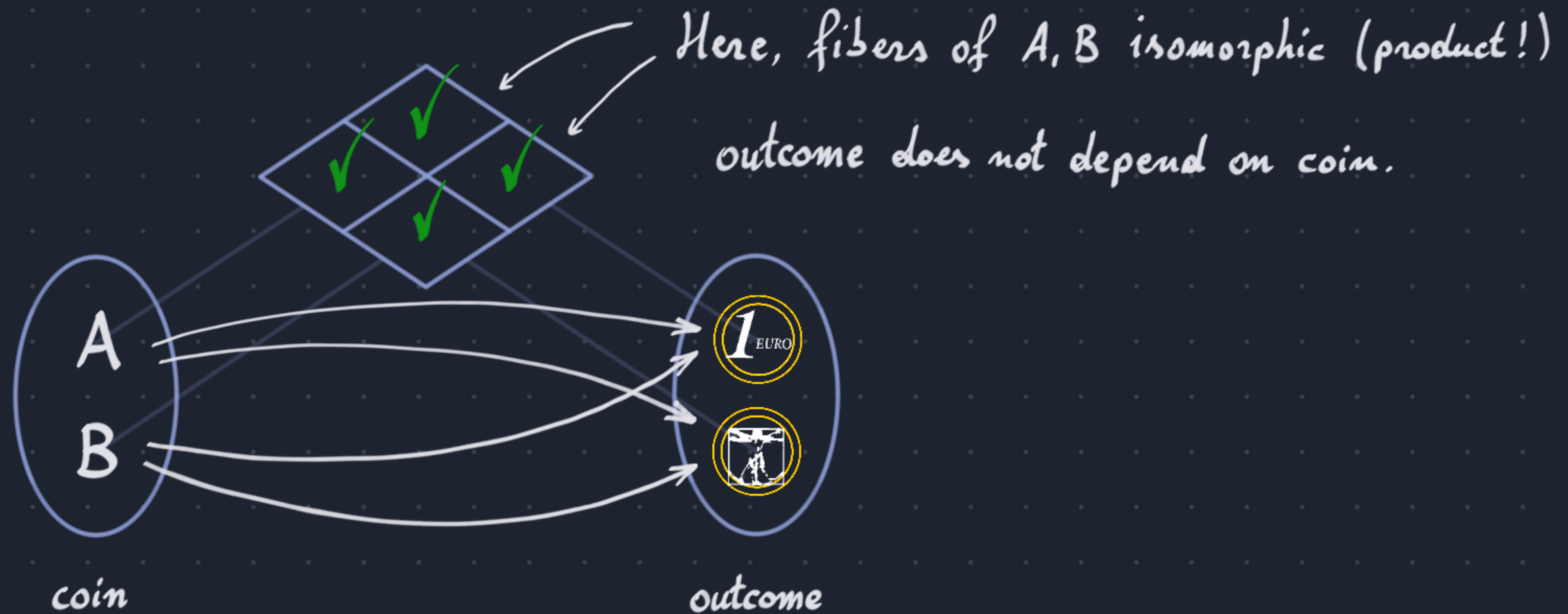
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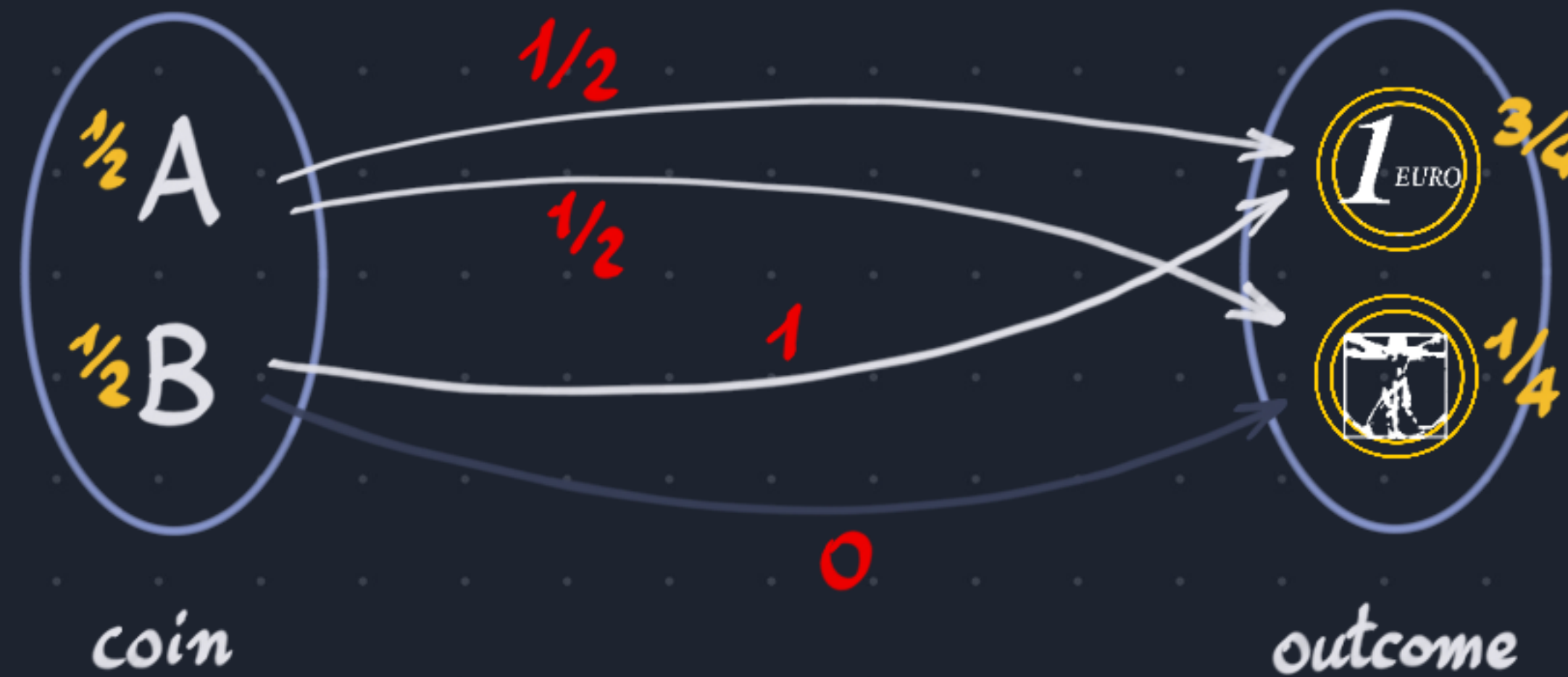
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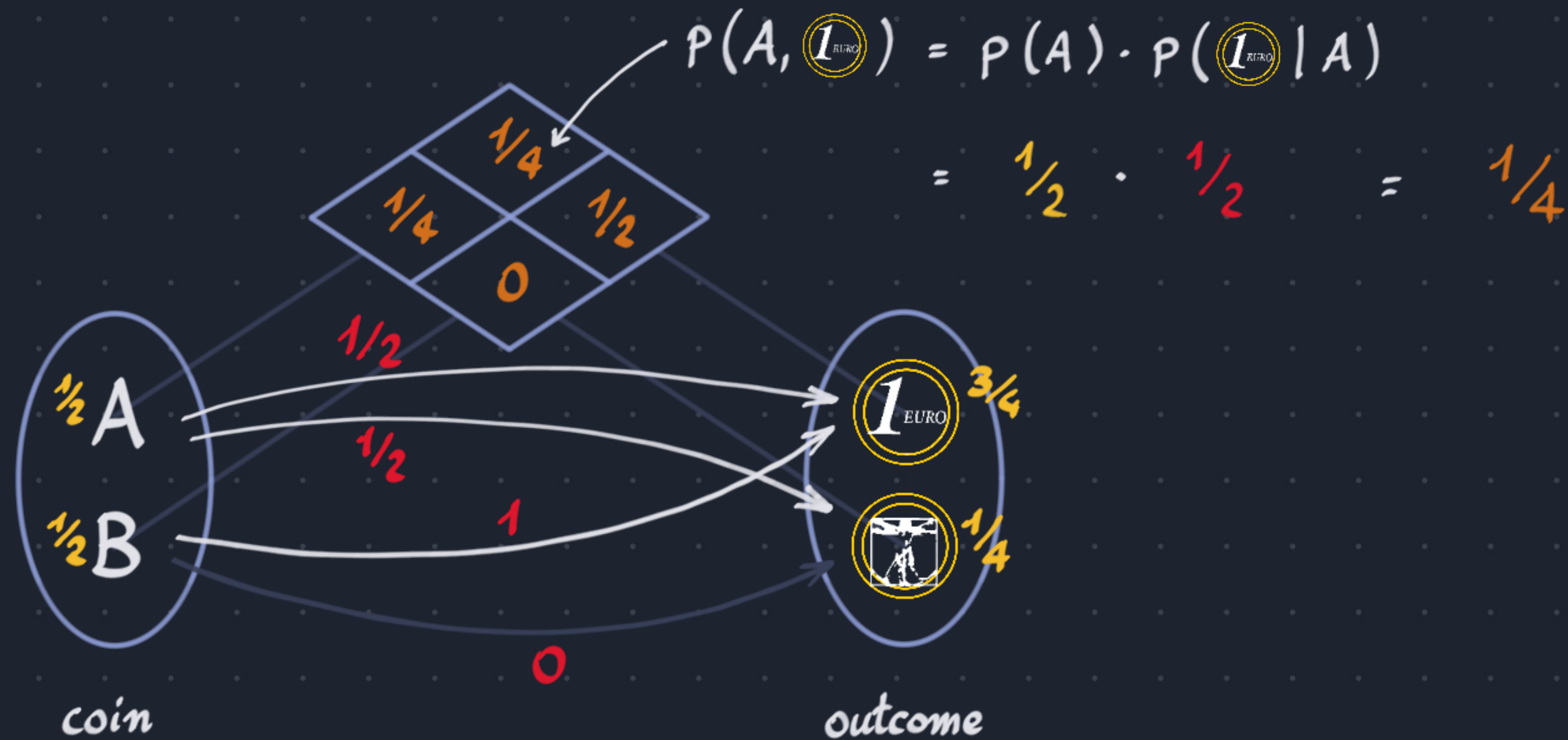
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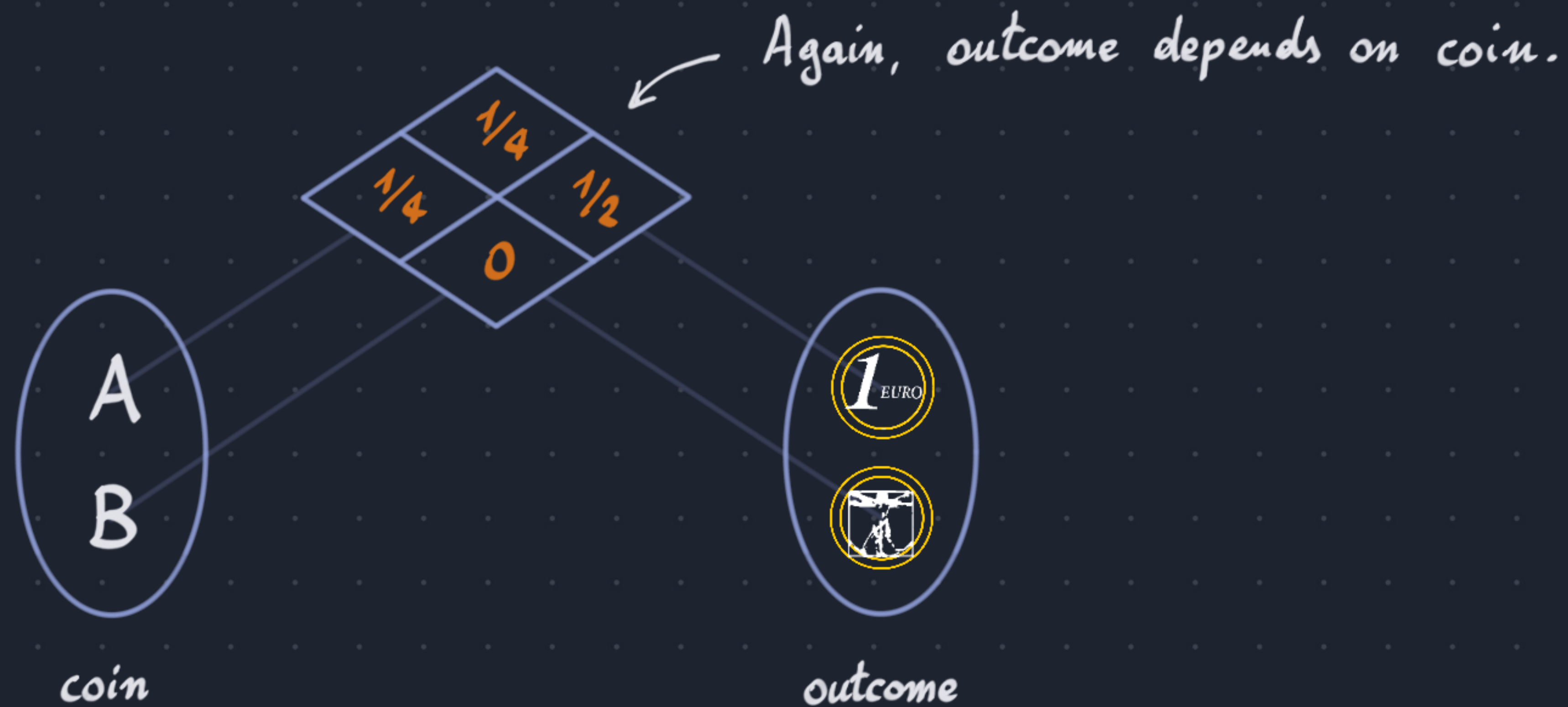
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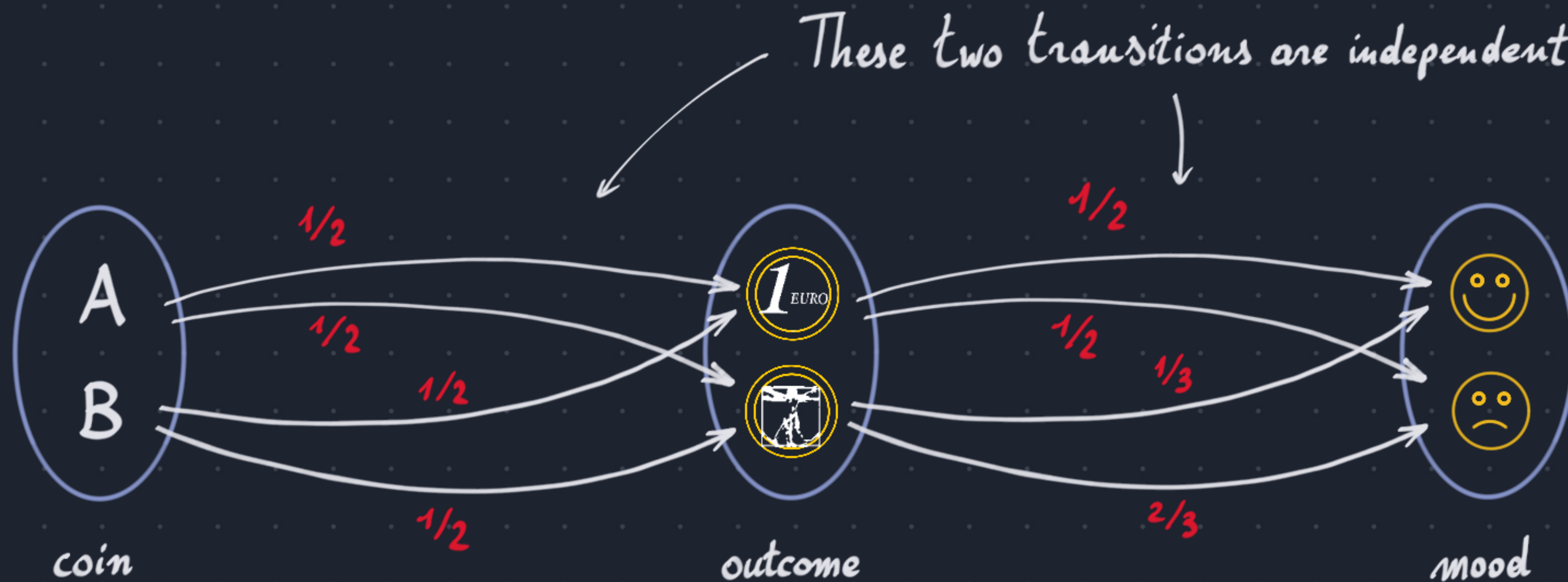
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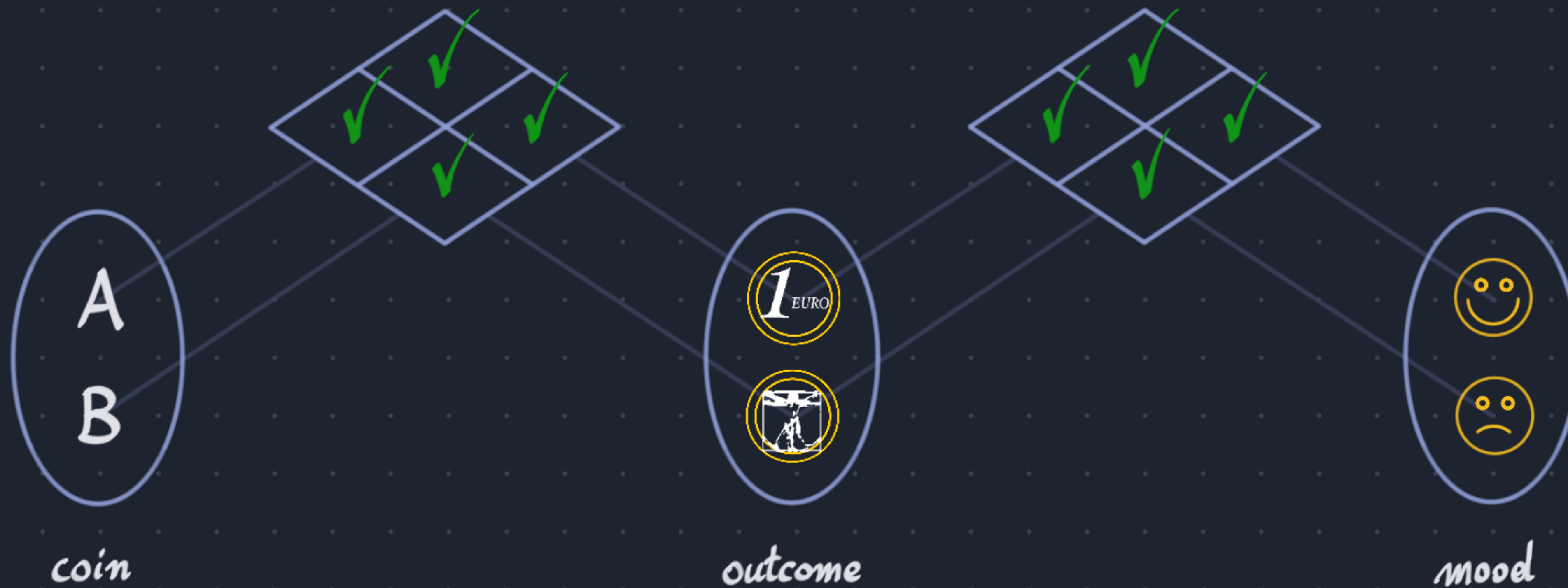
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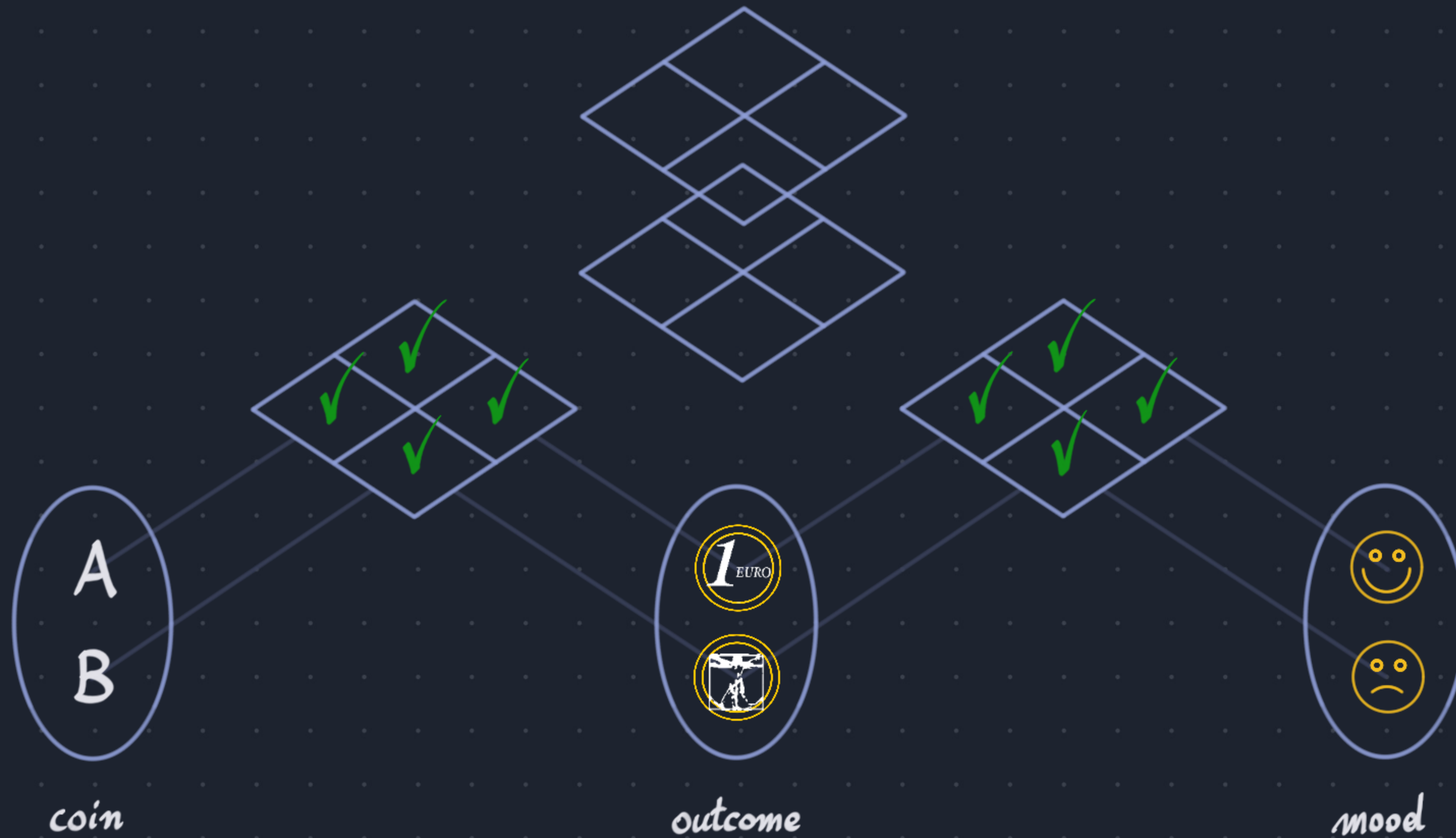
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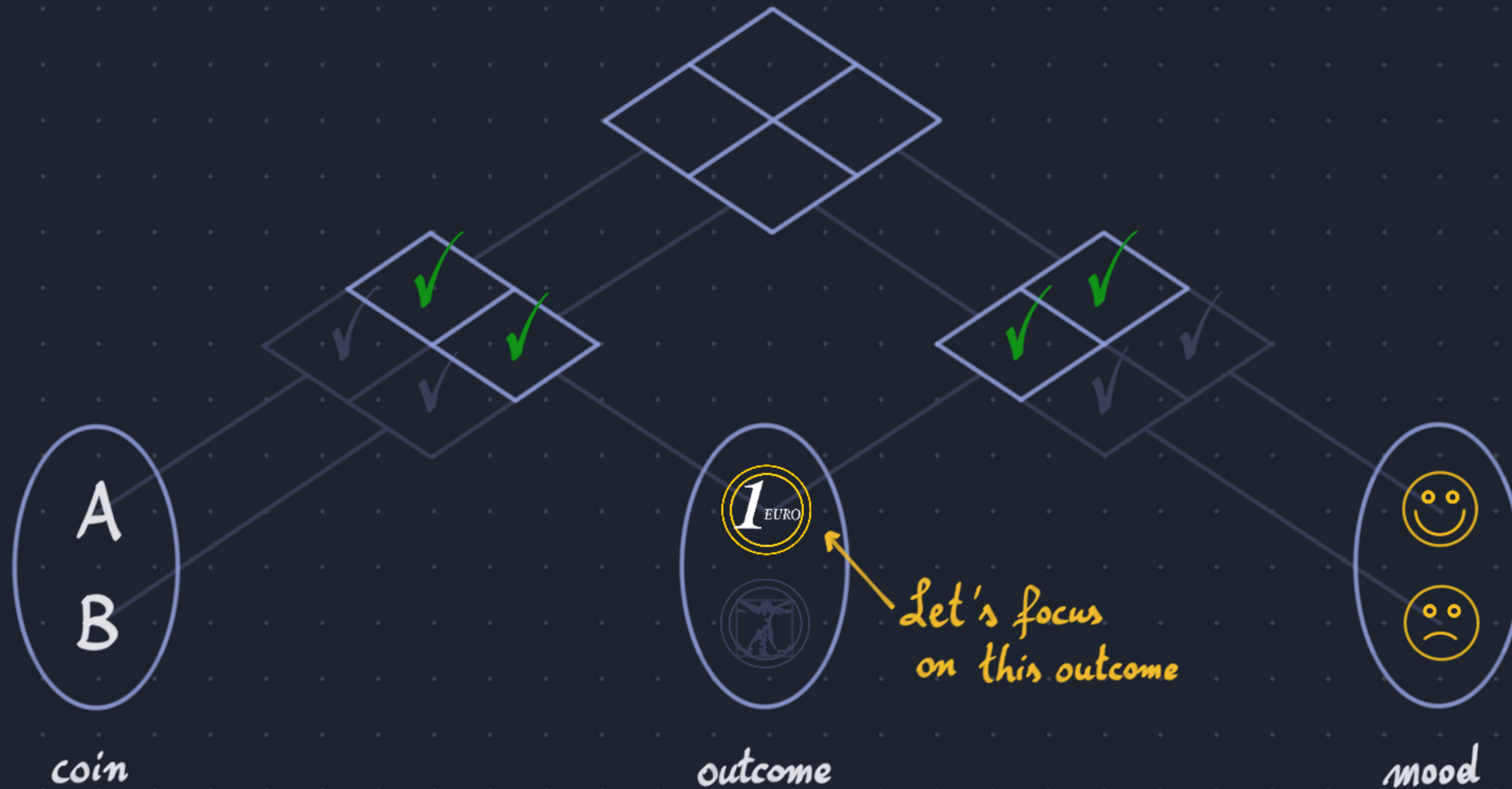
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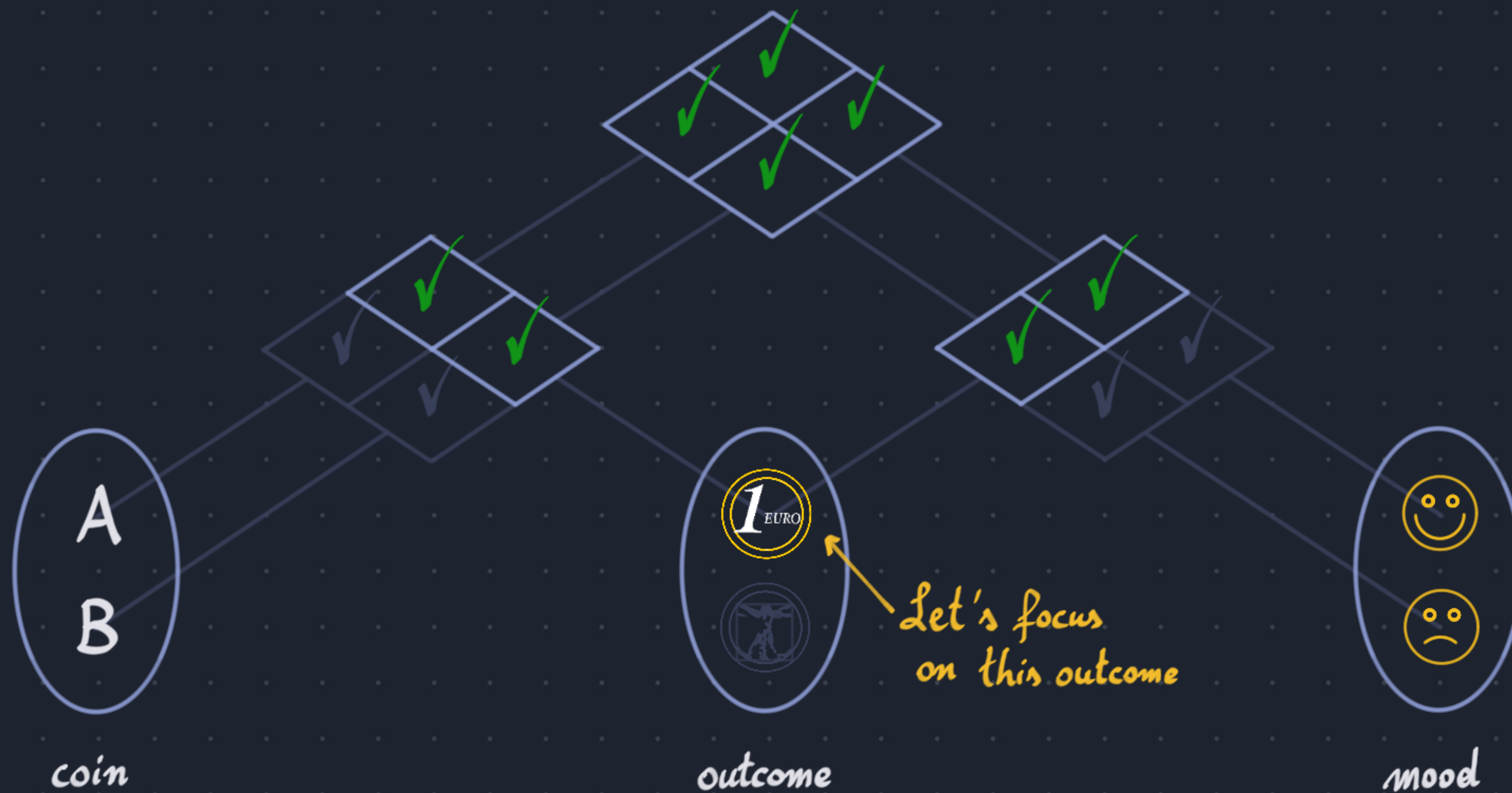
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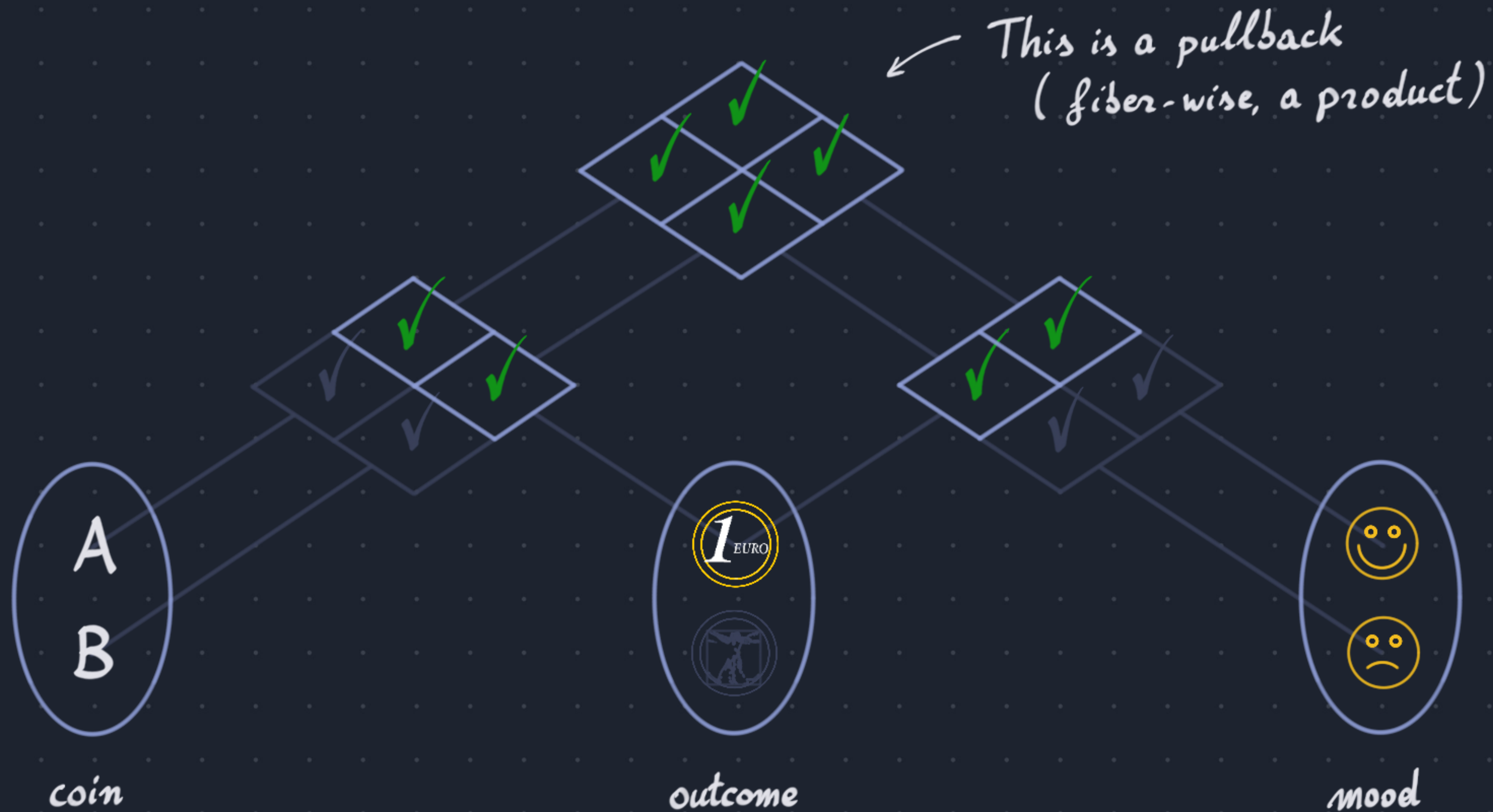
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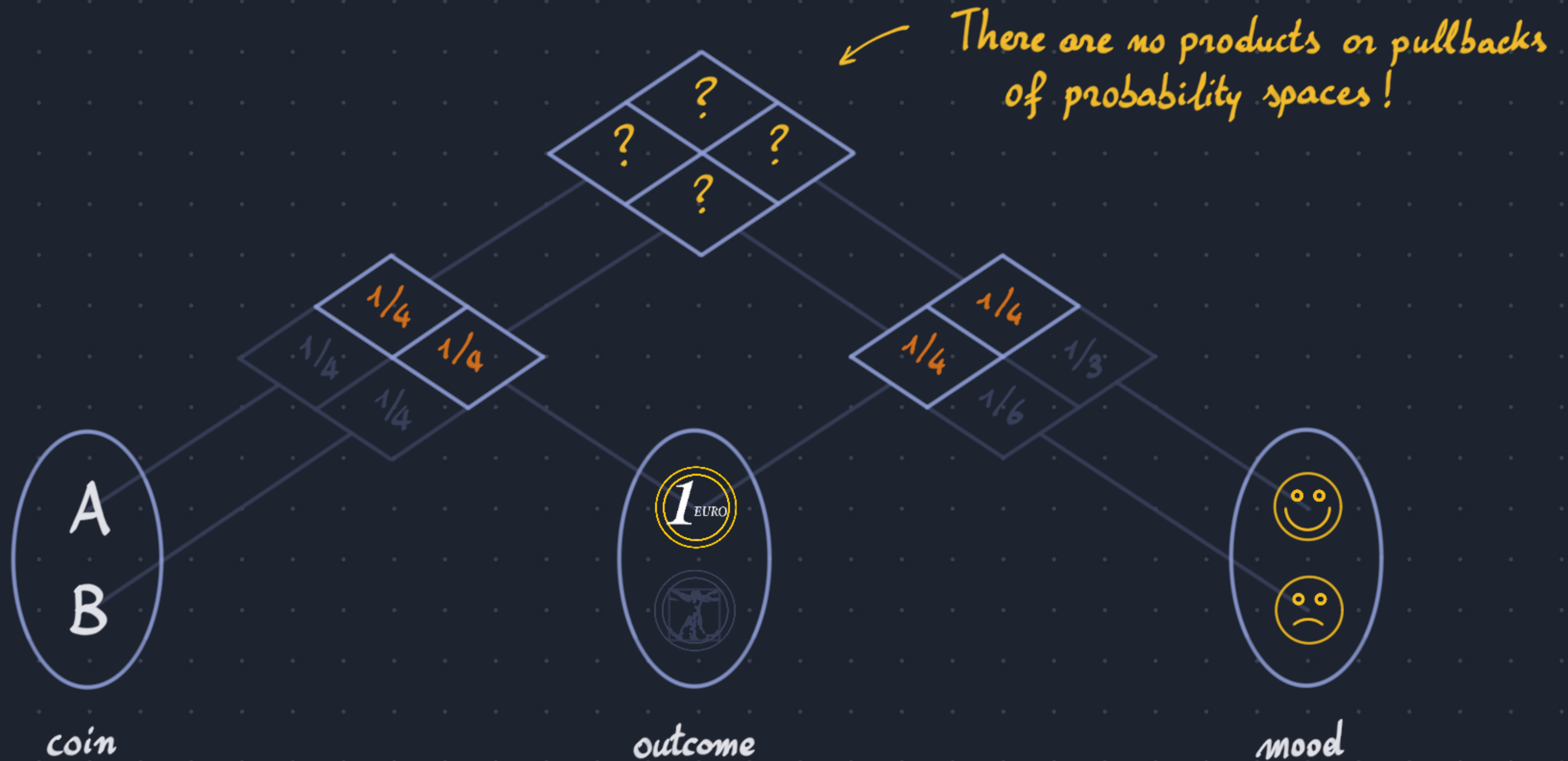
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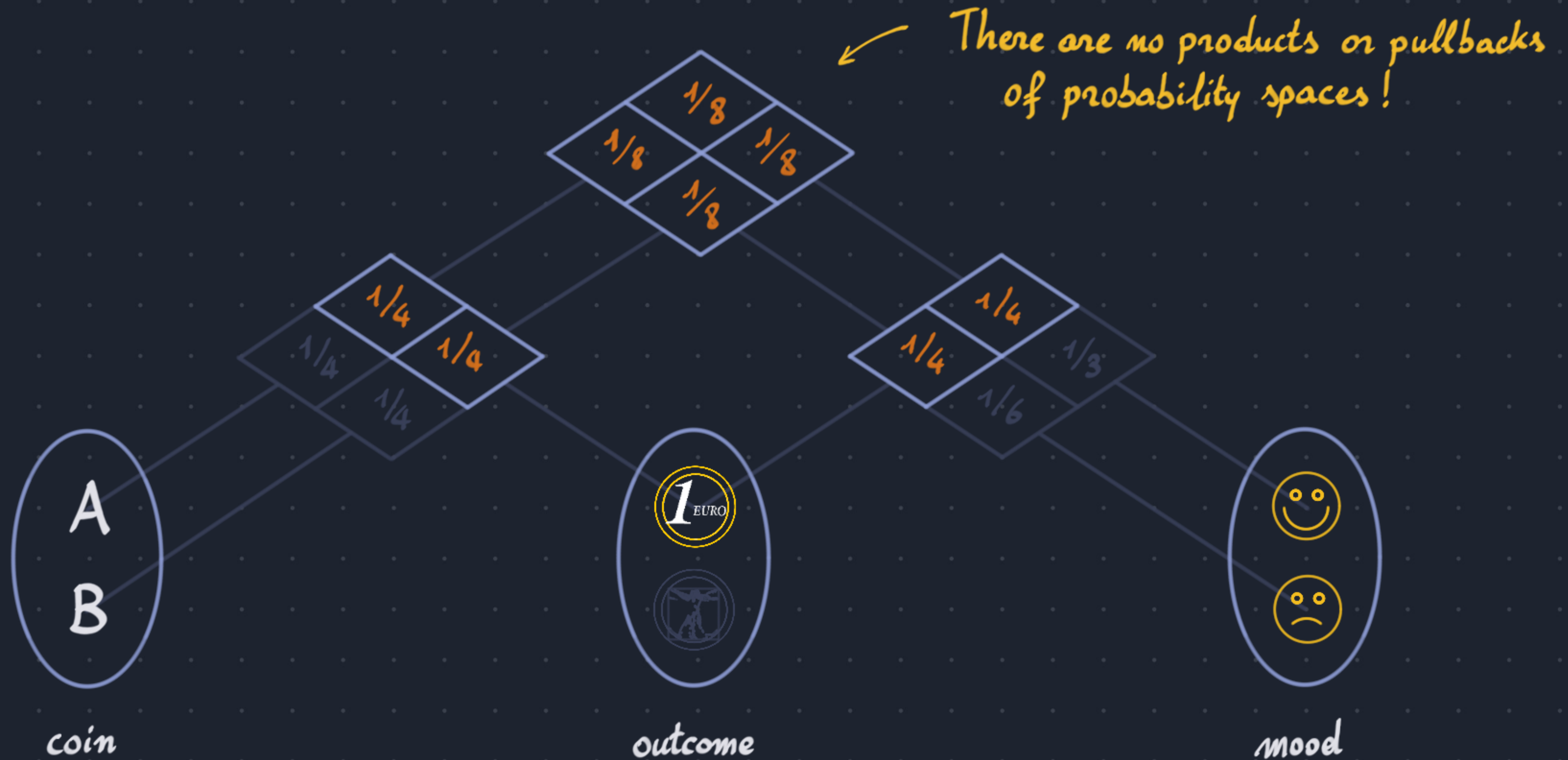
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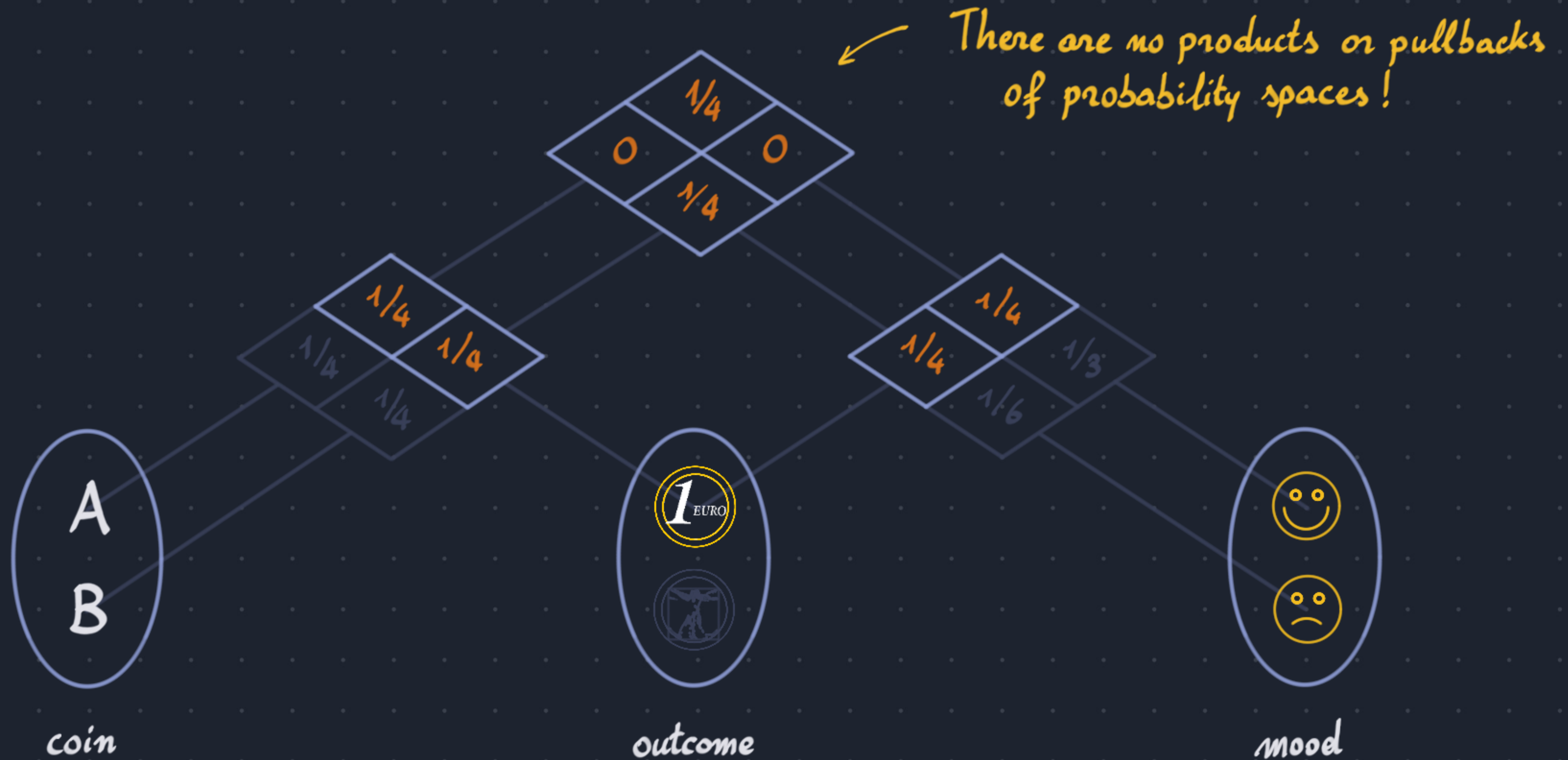
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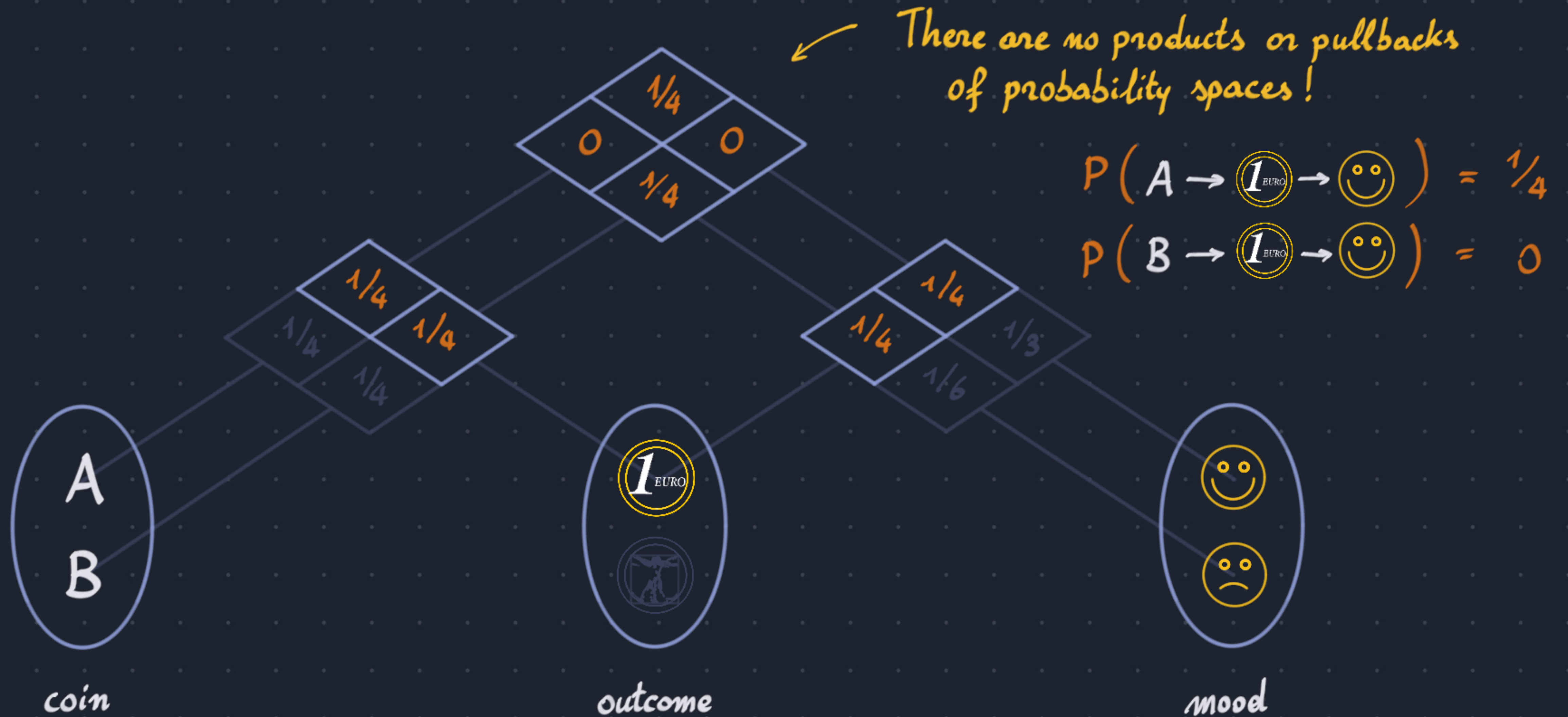
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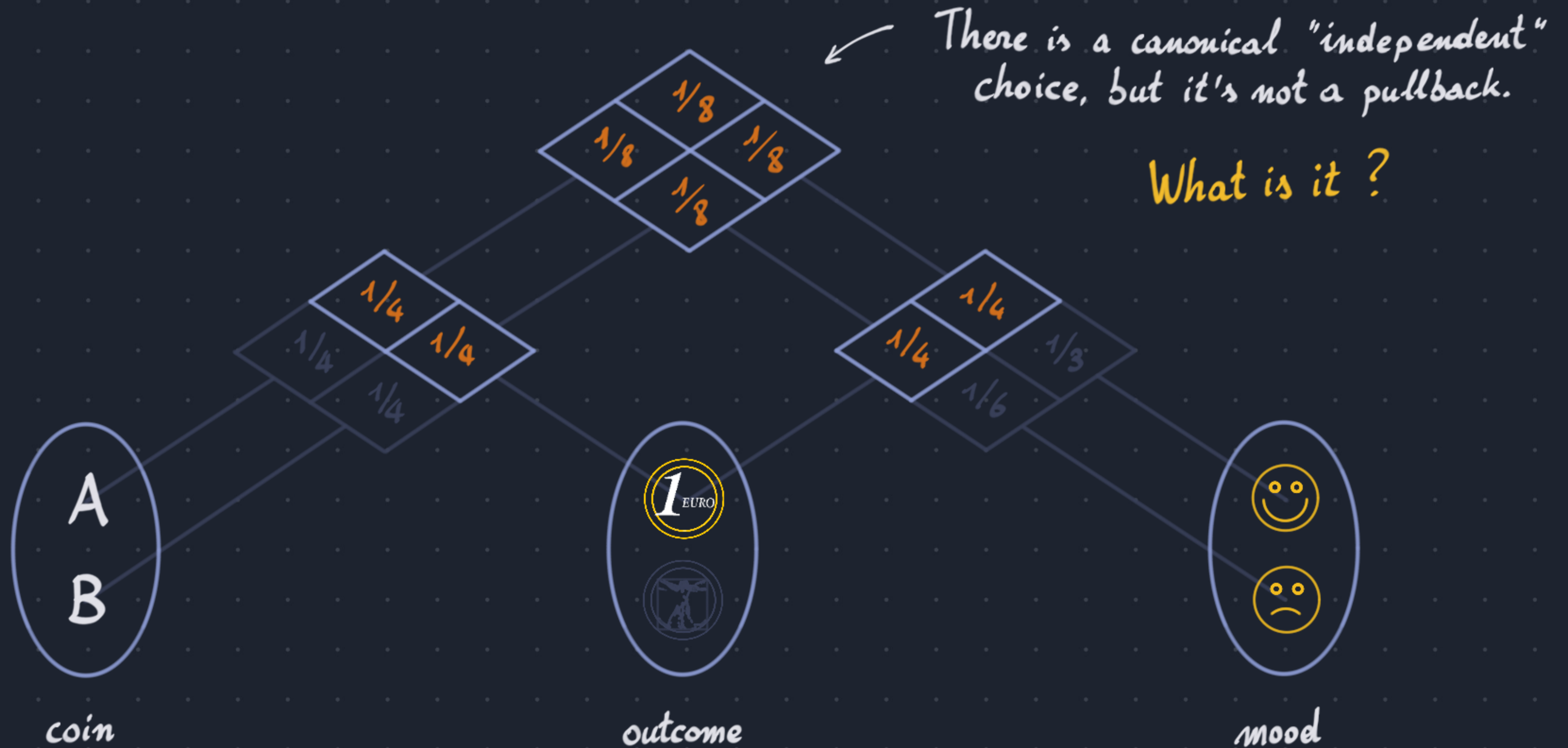
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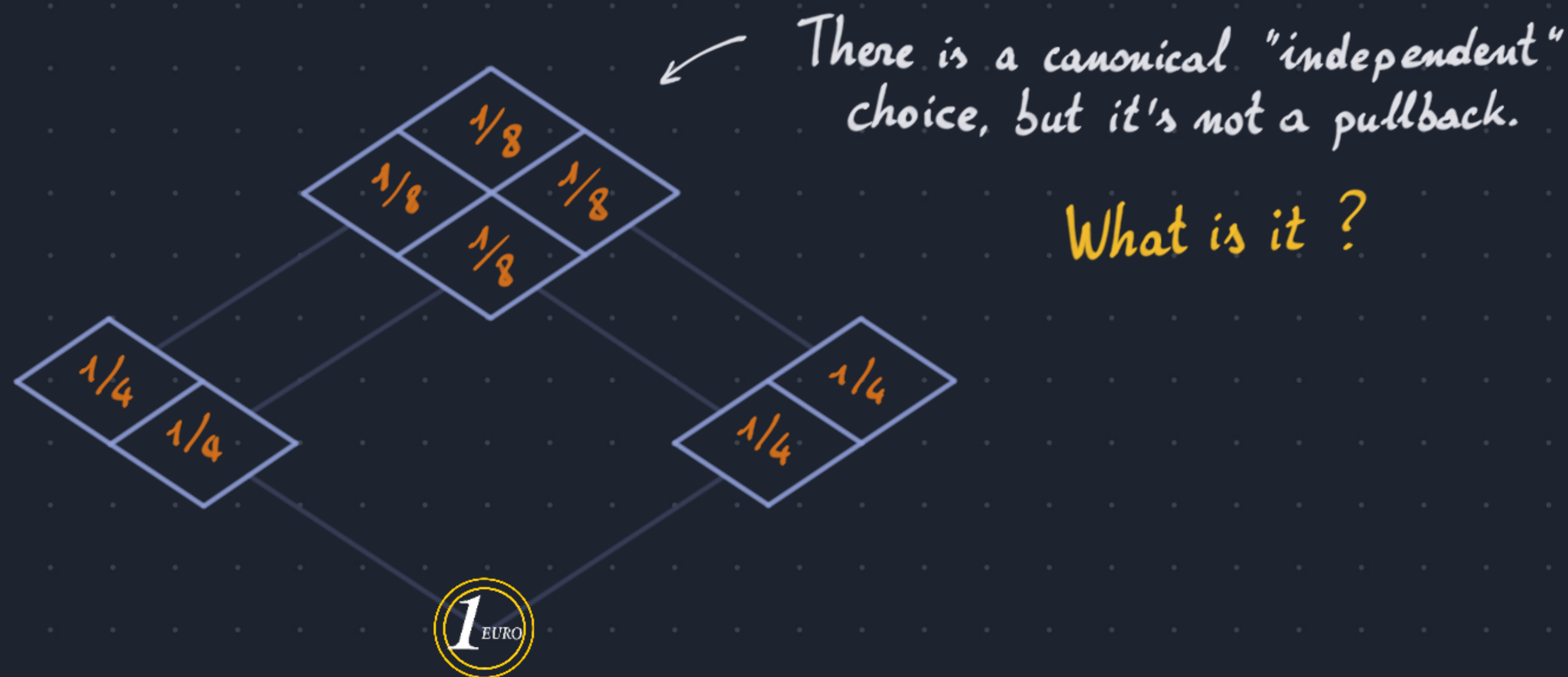
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1. MOTIVATION



- In this work, we
- Construct a **categorical theory of independence**
 - Use it to define a new **calculus of relations**.

1. MOTIVATION

2. $\perp = \perp\!\!\!\perp$

3. THE EQUIVALENCE

2. $\perp = \perp\!\!\!\perp$ · Towards a theory of independence

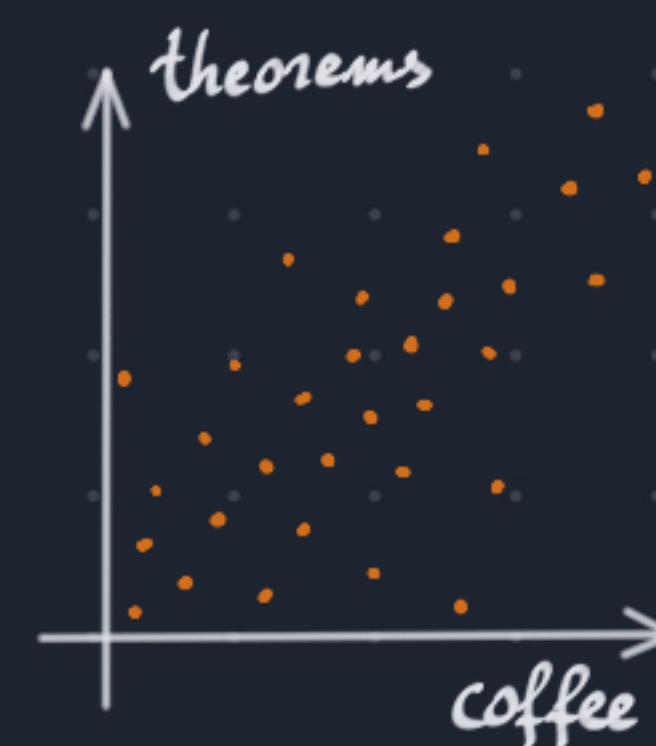
Idea. For a good categorical theory, it's ideal to have:

- 1) Conceptual understanding
of the phenomenon

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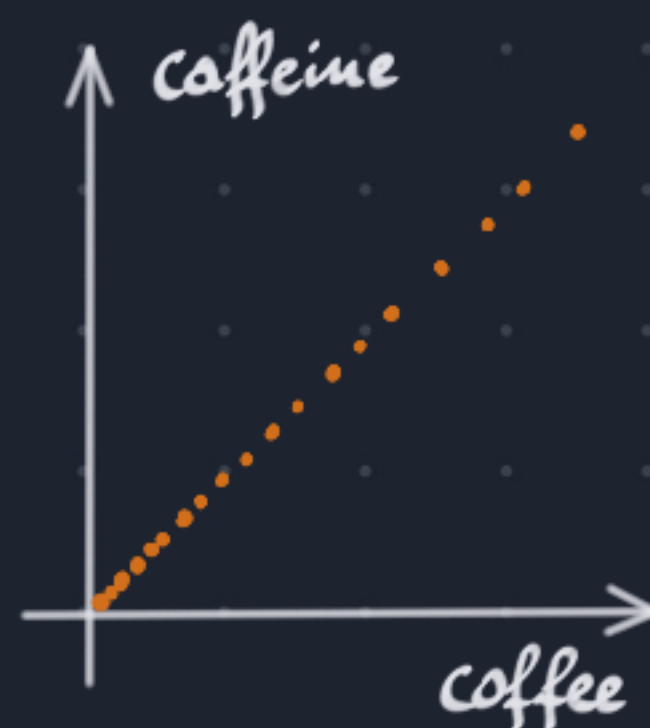


general case

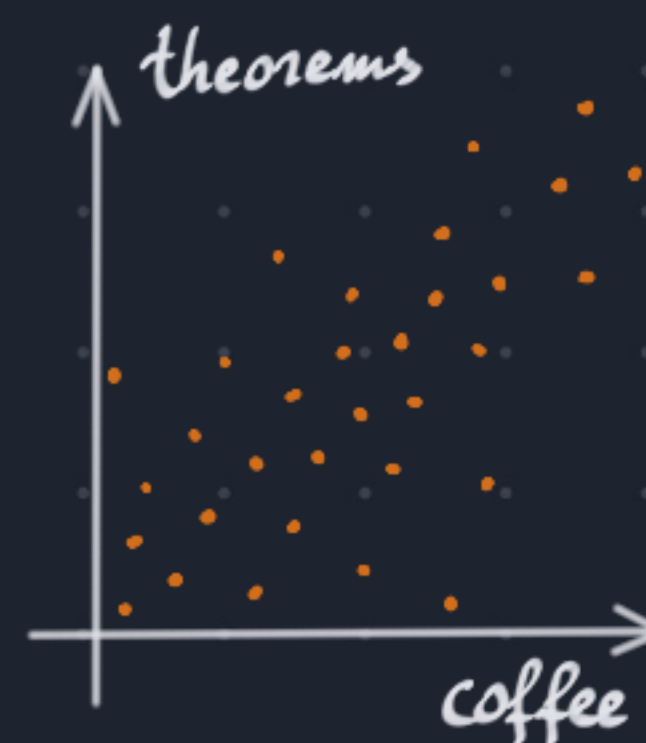
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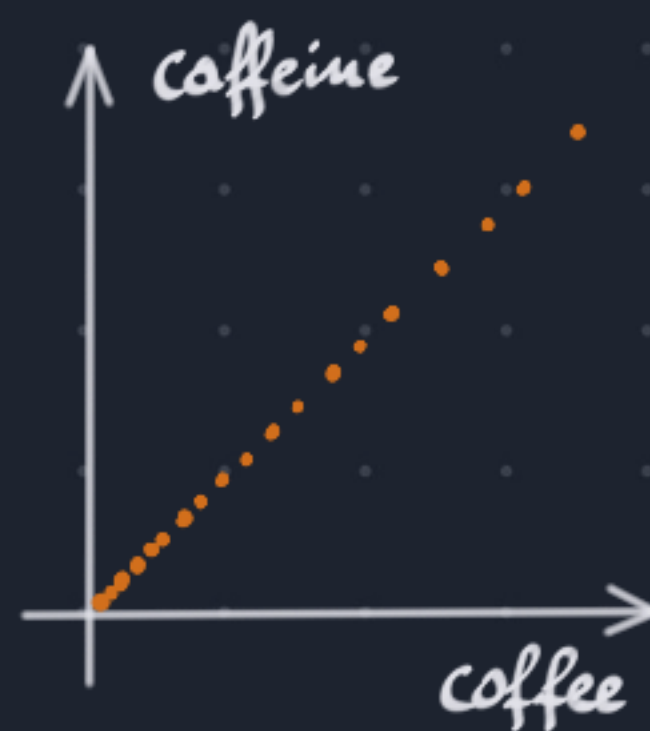


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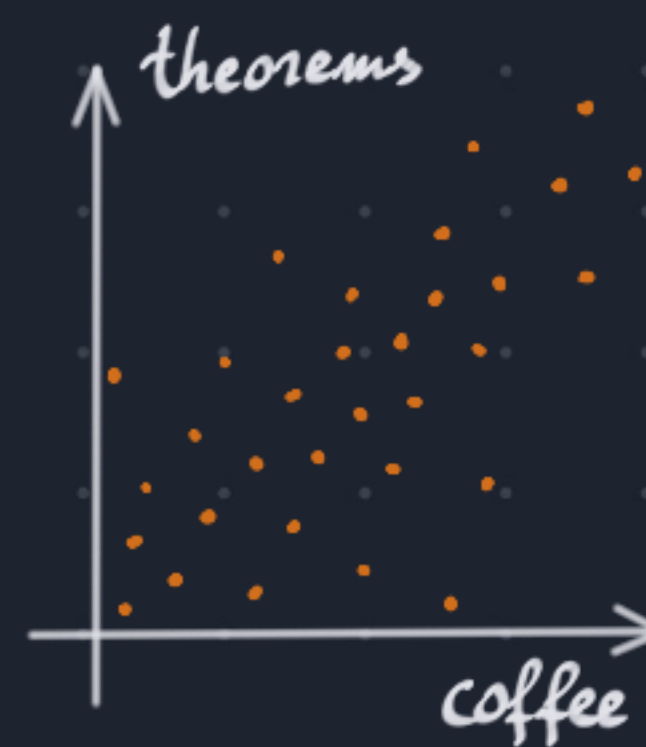
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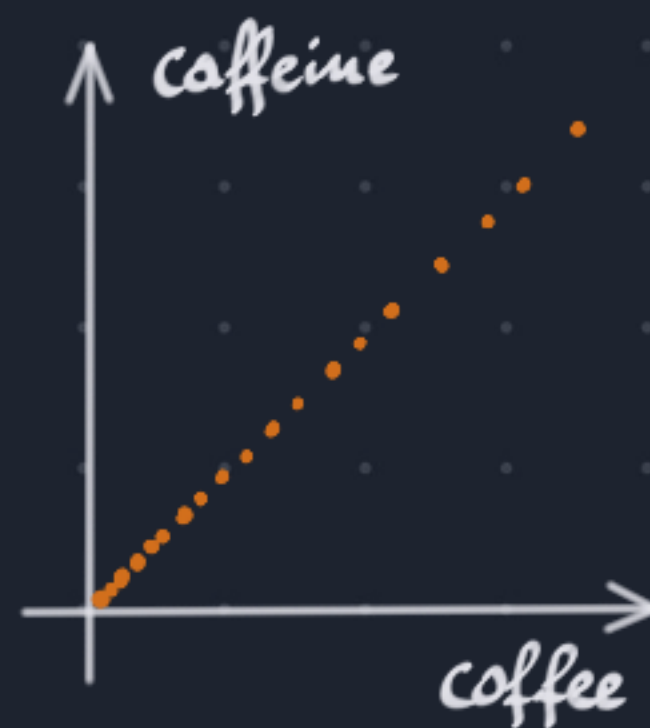


fully different

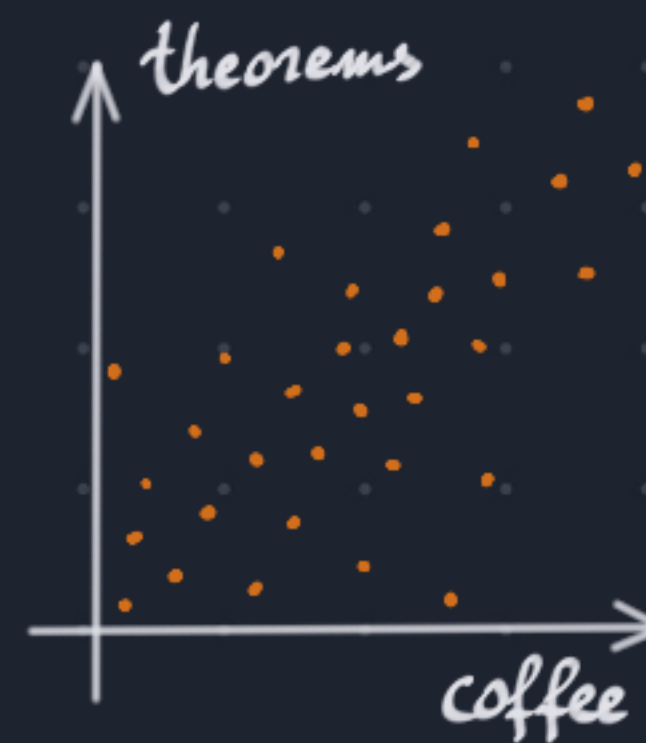
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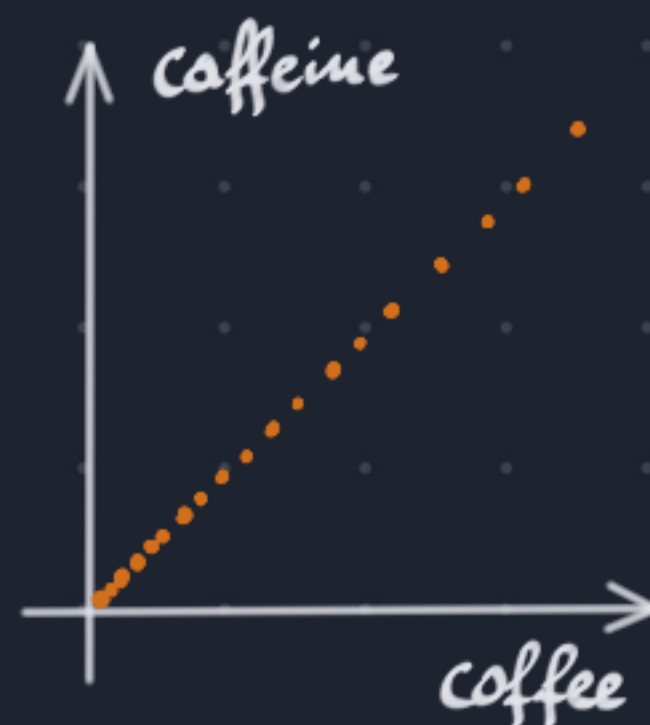
fully different

- 2) An example from a completely different field!

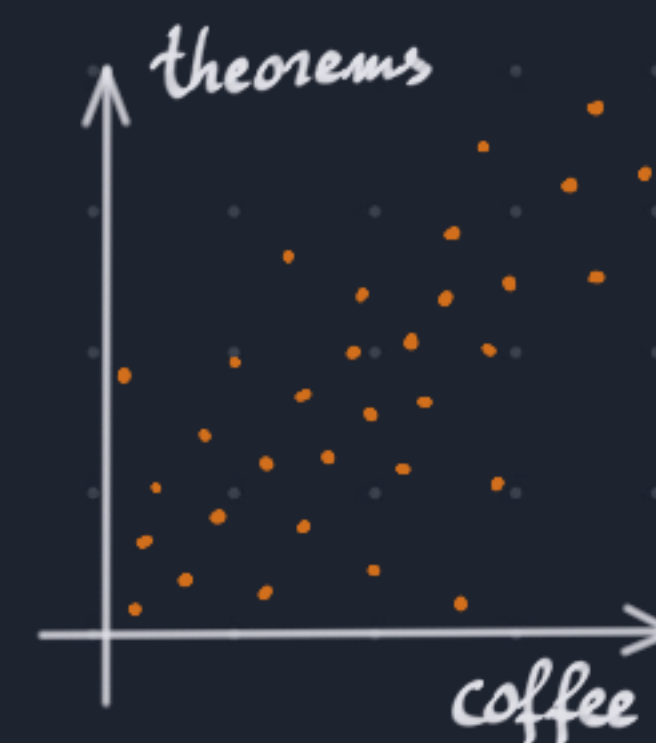
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Euclidean Geometry

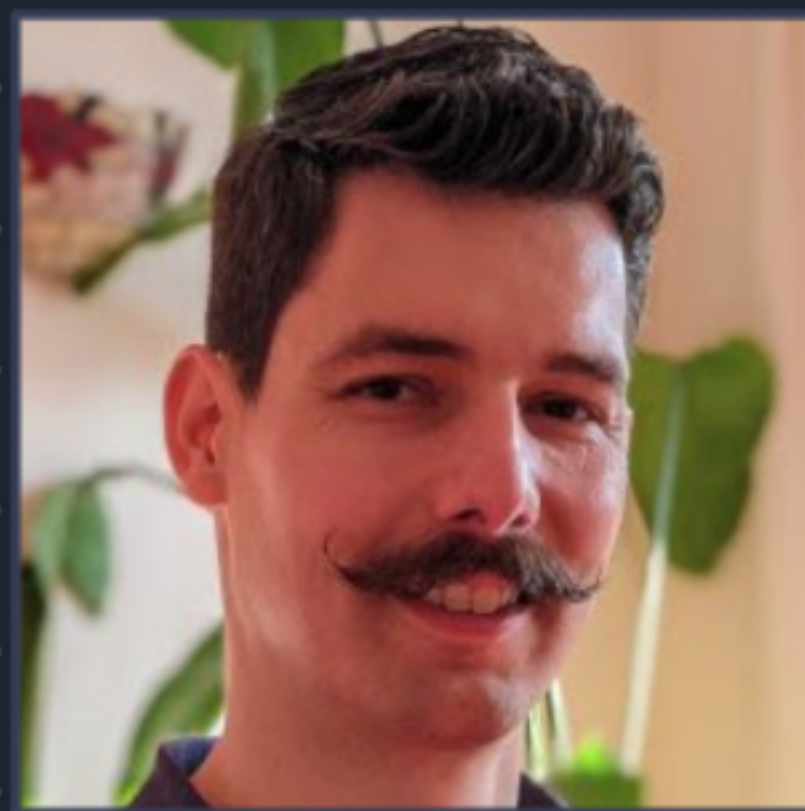


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+ - categories

2. $\perp = \perp\!\!\!\perp$ · $+$ -categories

Idea.

Dagger categories
categories

=

undirected graphs
directed graphs



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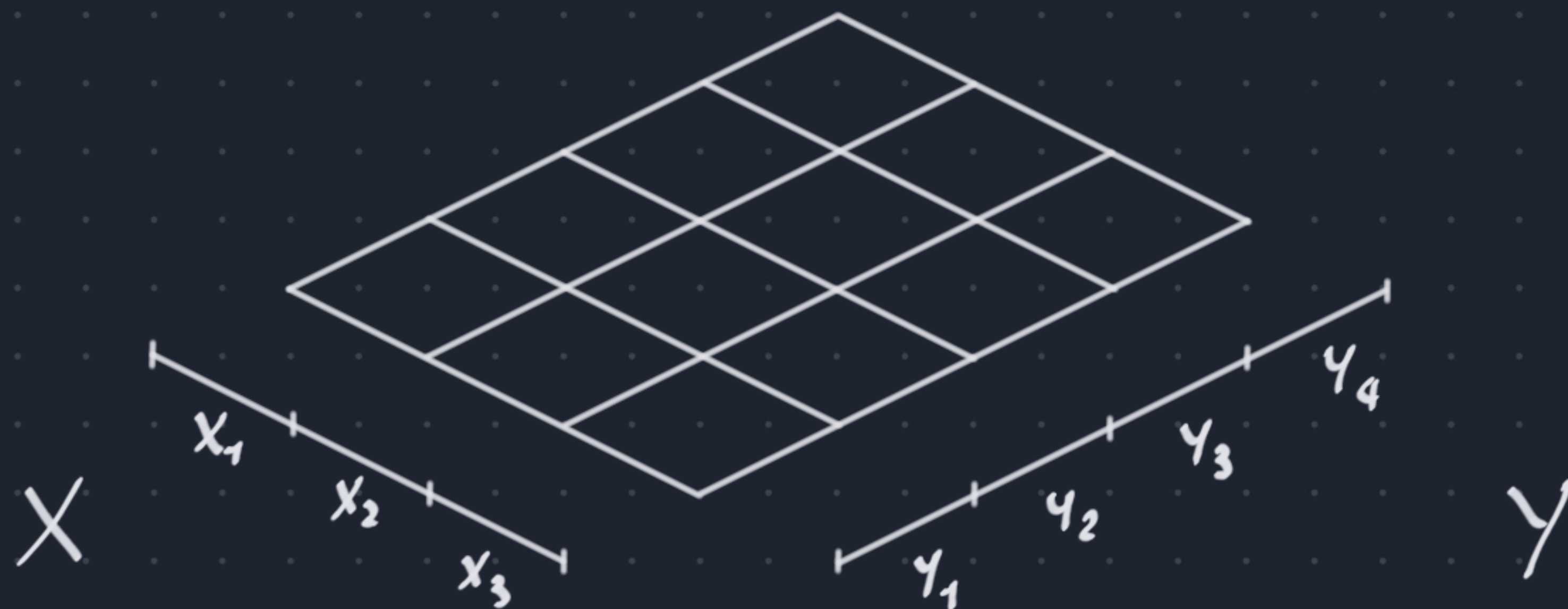
Idea.

Dagger categories
categories $=$ undirected graphs
directed graphs



Examples.

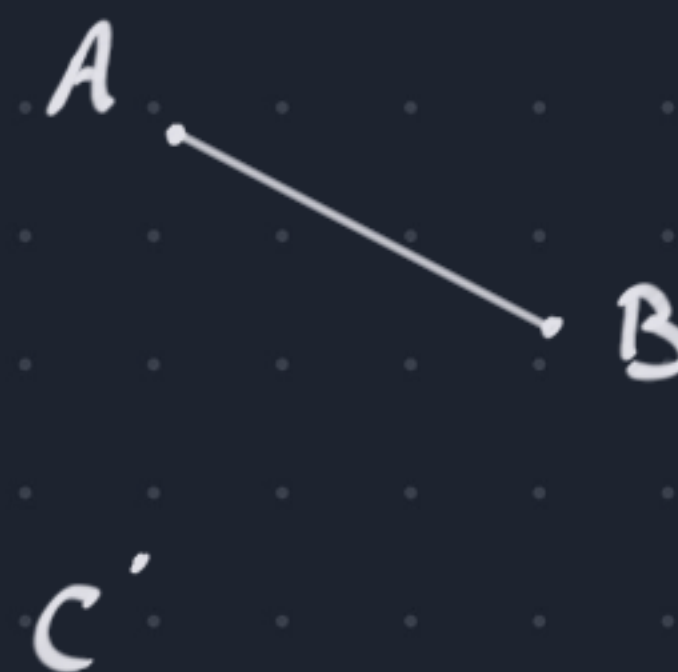
Rel



2. $\perp = \perp\!\!\!\perp$ · +-categories

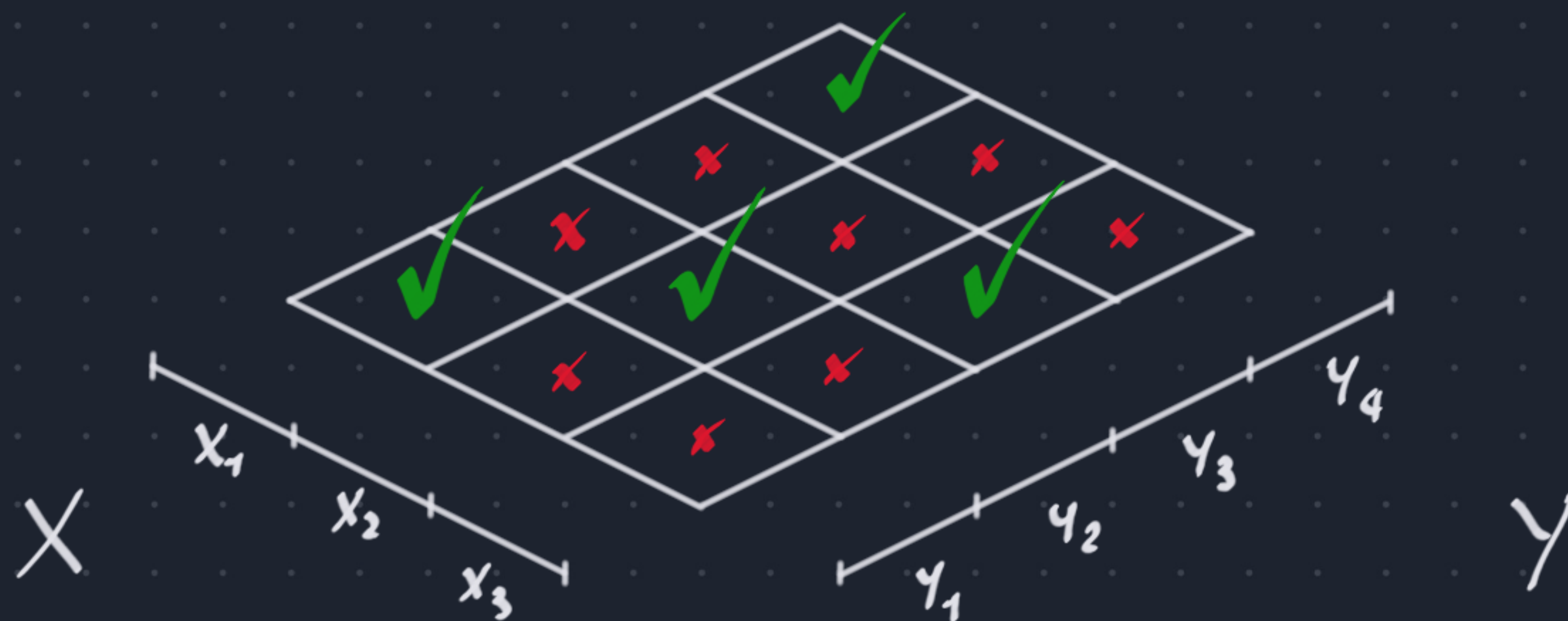
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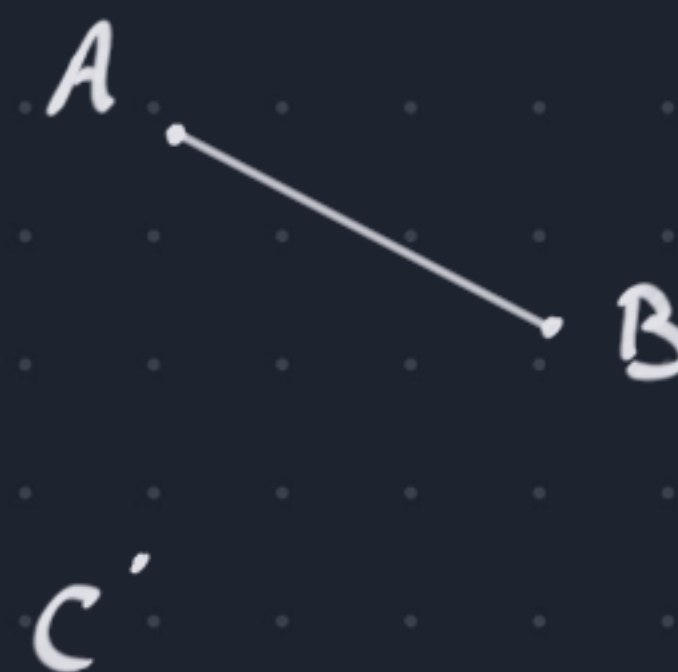
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2. $\perp = \coprod$ · $+$ -categories

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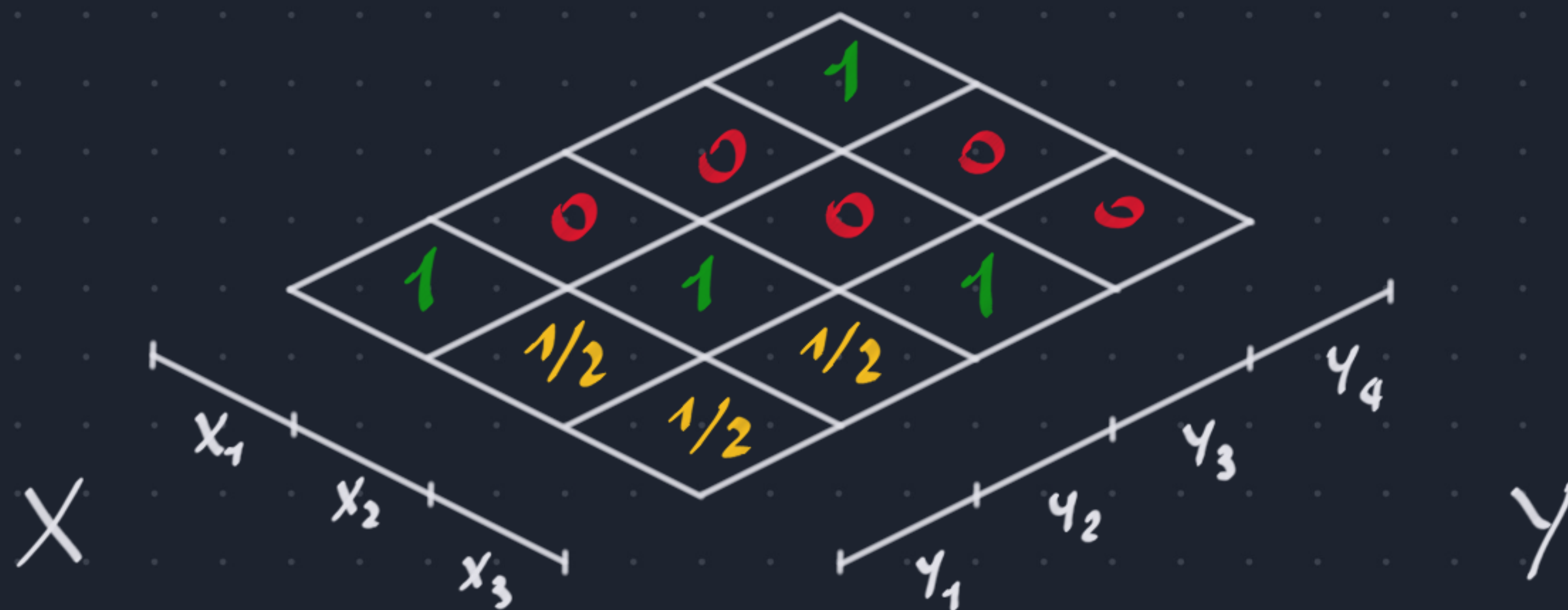
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Examples.

Rel

Mat



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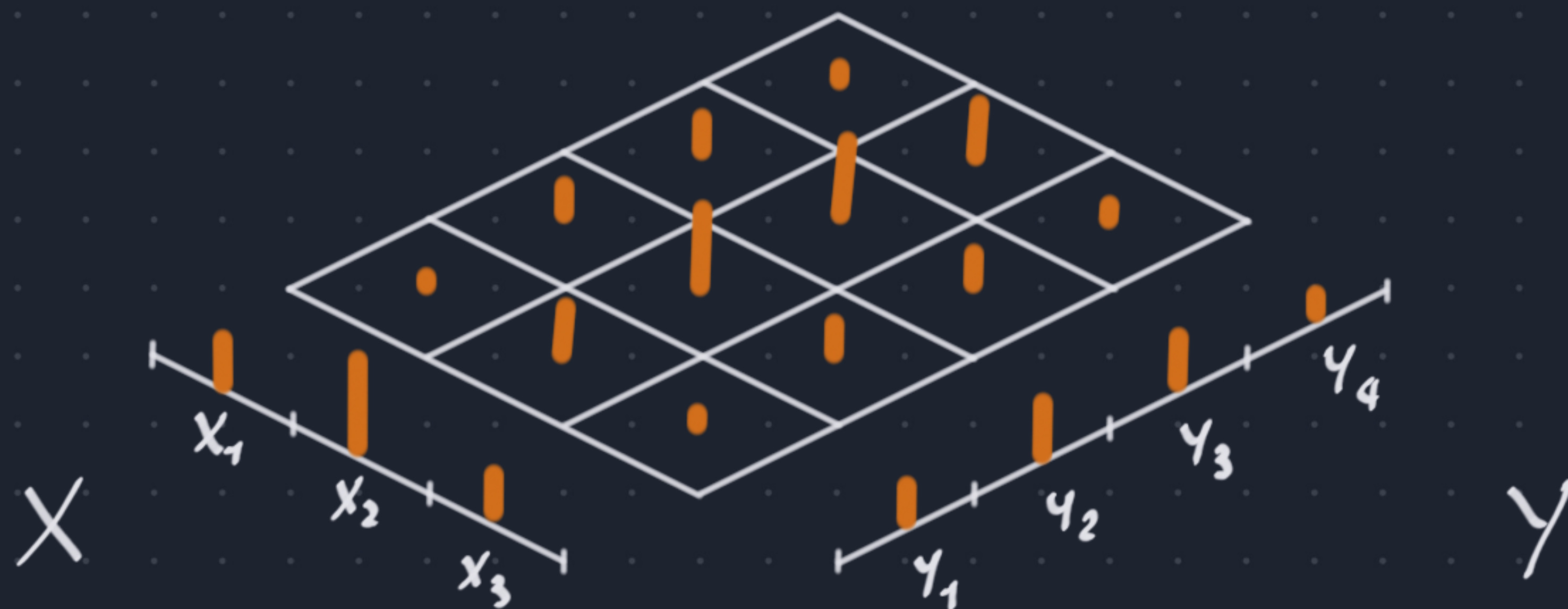


Examples.

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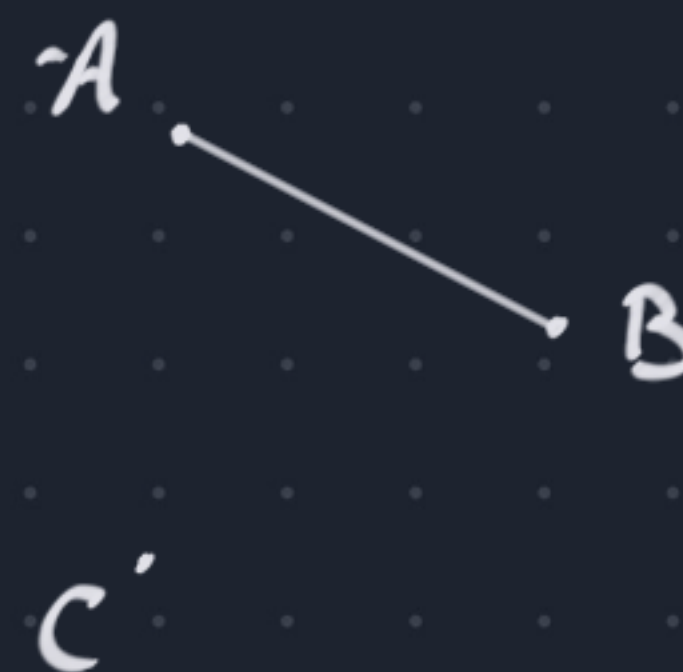
Kru



2. $\perp = \parallel$ · $+$ -categories

Idea.

$\frac{\text{Dagger categories}}{\text{categories}} = \frac{\text{undirected graphs}}{\text{directed graphs}}$



Examples.

Rel

Mat

Krn

Definition. A dagger structure on $\mathcal{C}$ is a functor $\mathcal{C}^{op} \xrightarrow[\cong]{+} \mathcal{C}$ such that:

- On objects, $X^+ = X$;
- On morphisms, $f^{++} = f$.

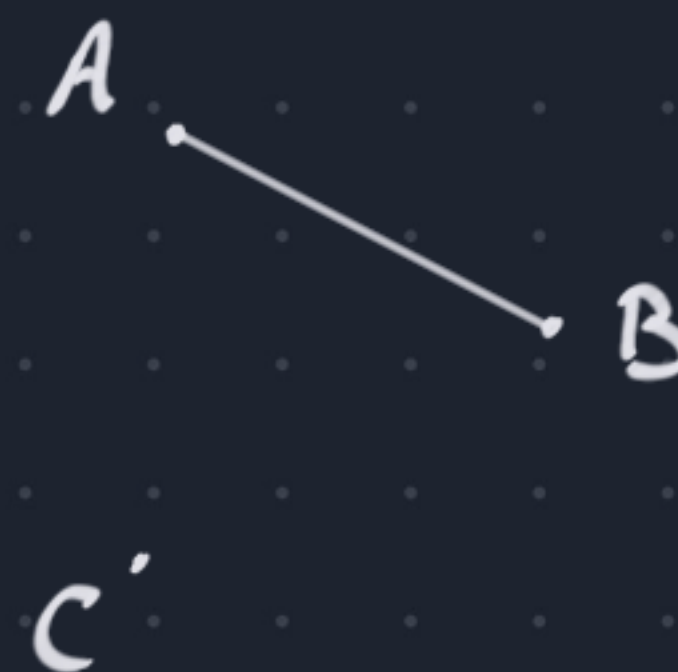
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$$(g \cdot f)^+ = f^+ \cdot g^+$$

2. $\perp = \coprod$ · "geometry" of $\perp$ -categories

Idea. $\perp$ -categories encode "geometric" notions. (This is known, but often overlooked.)

2. $\perp = \parallel$ · "Geometry" of $\dagger$ -categories

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Examples.

- The dagger operation itself $\begin{array}{c} x \\ \downarrow f \\ y \end{array} \leftrightarrow \begin{array}{c} x \\ \uparrow f^+ \\ y \end{array}$
can encode a **metric** (e.g. in Physics)

Mat

Vect_{BF}

Hilb

Krm

$$M \leftrightarrow M^T$$

$$|v\rangle \leftrightarrow \langle v|$$

$$v_i \leftrightarrow v^i$$

$$p(y|x) \leftrightarrow p(x|y)$$

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- An iso $f: X \xrightarrow{\cong} Y$ is called
a $\dagger$ -iso or unitary if $f^{-1} = f^+$.

Mat	$\text{Vect}_{\mathbb{R}}^f$	Hilb	Krn
$M \leftrightarrow M^T$		$ v\rangle \leftrightarrow \langle v $	
	$v_i \leftrightarrow v^i$		$p(y x) \leftrightarrow p(x y)$
Orthogonal matrix	(Hyp.) rotation, symplectic map...	Unitary operator	(Deterministic) a.s. isomorphism

2. $\perp = \parallel$ · "Geometry" of $\perp$ -categories

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Examples.

- A mono $f: X \hookrightarrow Y$ is called a $\perp$ -mono or isometry if $f^+ f = \text{id}$.

Mat Vect_{BF} Hilb Krn

Embedding / isometry

Proper kernel

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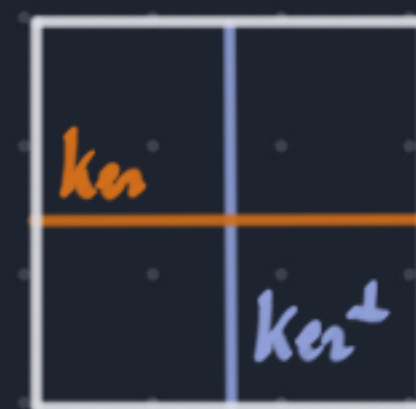
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Mat Vect_{BF} Hilb K_{rn}

Embedding / isometry

Proper kernel



$\cong$
 $\rightarrow$

Det. map
(coarse-graining)

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Examples.

- A mono $f: X \hookrightarrow Y$ is called a $\dagger$ -mono or isometry if $f^\dagger f = \text{id}$.

$f^\dagger$ is called a $\dagger$ -epi or coisometry.

- An idempotent $X \xrightarrow{e} X$ is called a $\dagger$ -idempotent or projector if $e^\dagger = e$.

Mat Vect_{BF} Hilb Krn

Embedding / isometry

Proper kernel



Det. map
(coarse-graining)

Orthogonal projection

Conditional
expectation

2. $\perp = \perp\!\!\!\perp$ · Independent squares

Definition (new). A commutative square in a $\dagger$ -category

$$\begin{array}{ccc} A & \xrightarrow{g} & C \\ f \downarrow & & \downarrow v \\ B & \xrightarrow{u} & D \end{array}$$

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Definition (new). A commutative square in a $\dagger$ -category

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is called *orthogonal*

$$\begin{array}{ccc} & \longrightarrow & \\ \downarrow & \perp & \downarrow \\ & \longrightarrow & \end{array}$$

or *independent*

$$\begin{array}{ccc} & \longrightarrow & \\ \downarrow & \perp\!\!\!\perp & \downarrow \\ & \longrightarrow & \end{array}$$

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$$\text{if } gf^+ = v^+u.$$

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$$\text{if } gf^{\dagger} = v^{\dagger}u.$$

- We are mostly interested in squares of "maps" (e.g. $\dagger$ -epis).

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Examples. · In Rel, with functions:



$$\forall b \in B, c \in C:$$

$$\exists a \in A : f(a) = b, g(a) = c$$

$$\Updownarrow$$

$$\exists d \in D : u(b) = d = v(c)$$

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Examples. · In Rel, with functions:



$$\forall b \in B, c \in C : u(b) = v(c),$$

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i.e. weak pullback (BC)

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Examples. · In Rel, with functions: weak pullback

· In Mat, for vectors, $D = 0$:



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Examples. · In Rel, with functions: weak pullback

· In Mat, for vectors, $D = 0$: usual orthogonality

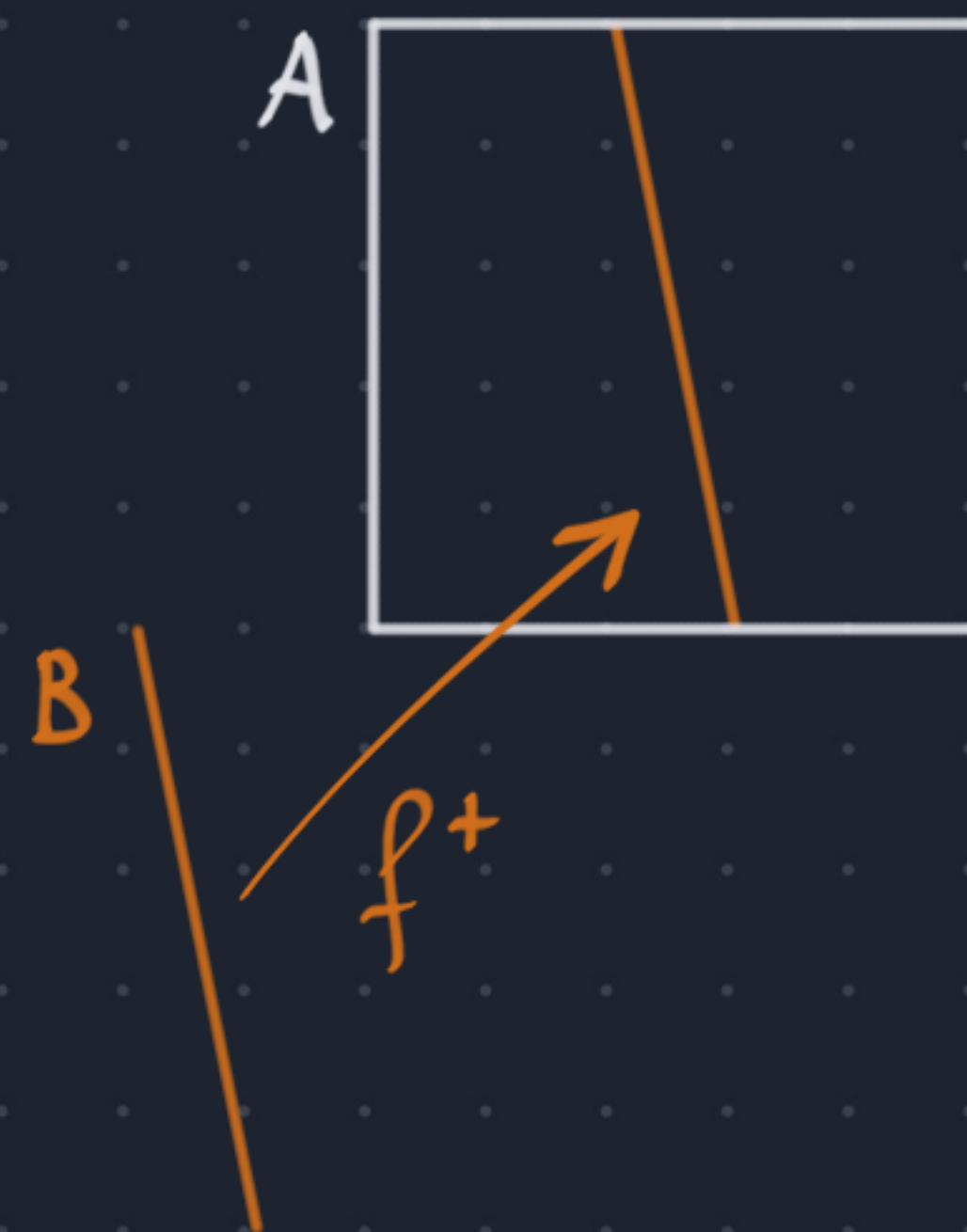


$$v^+ w = 0$$

2. $\perp = \parallel$ · Independent squares

Examples. · In Rel, with functions: weak pullback

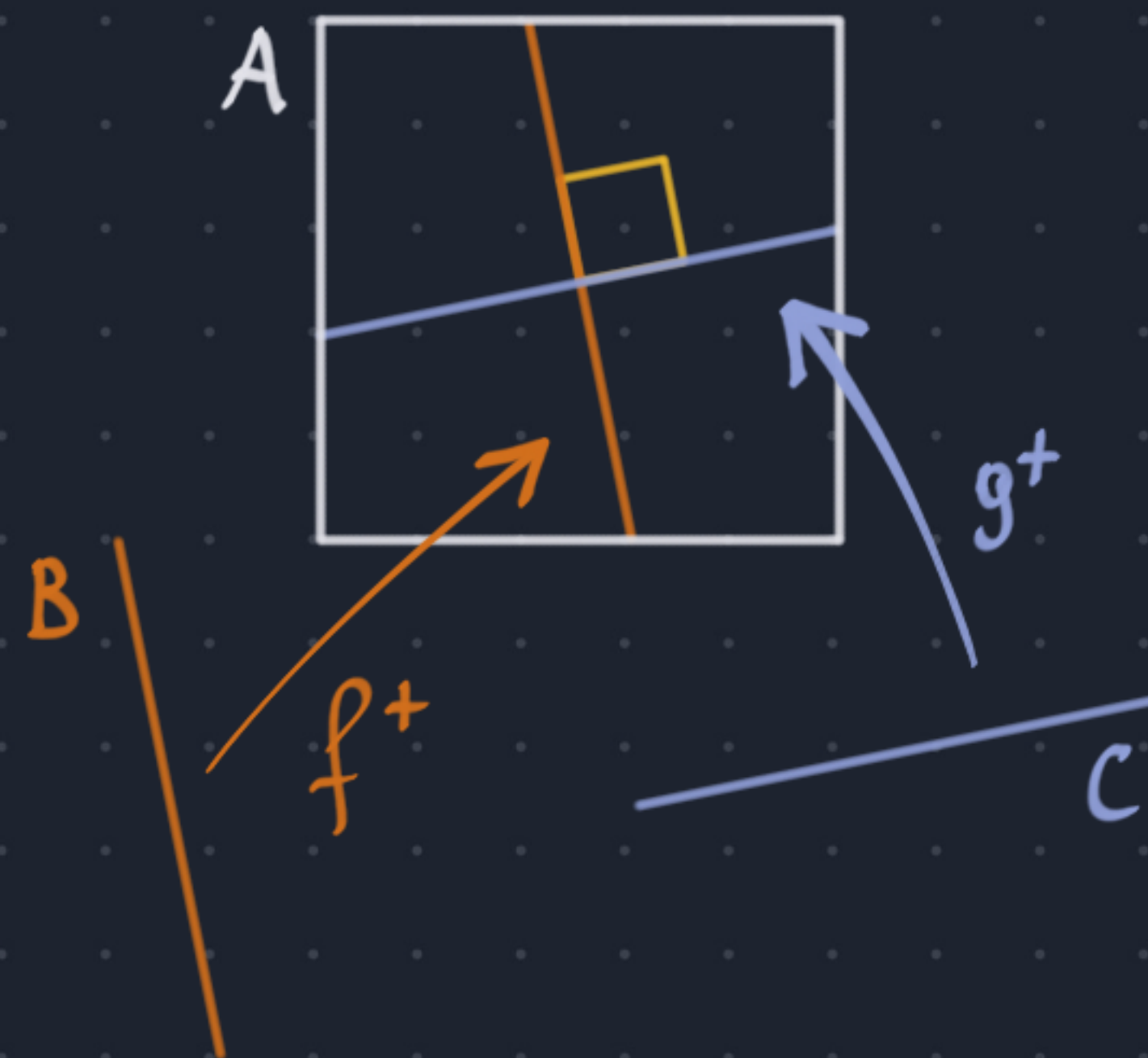
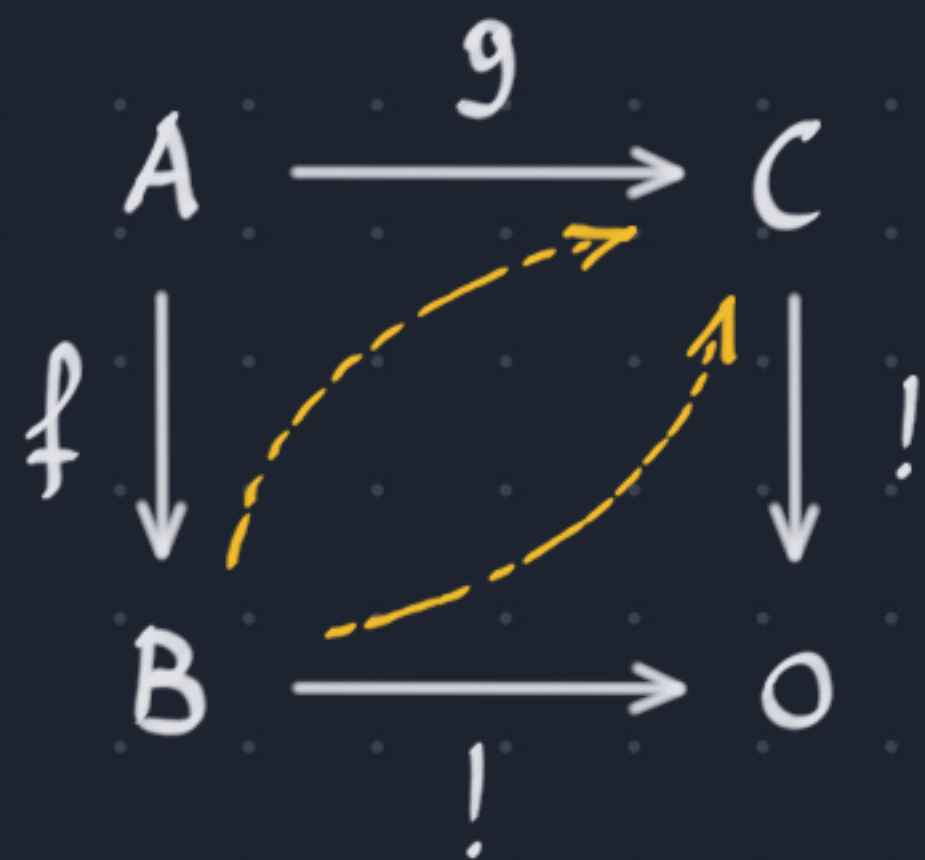
- In Mat, for vectors, $D = 0$: usual orthogonality
- for $\dagger$ -epis, $D = 0$: orthogonal subspaces



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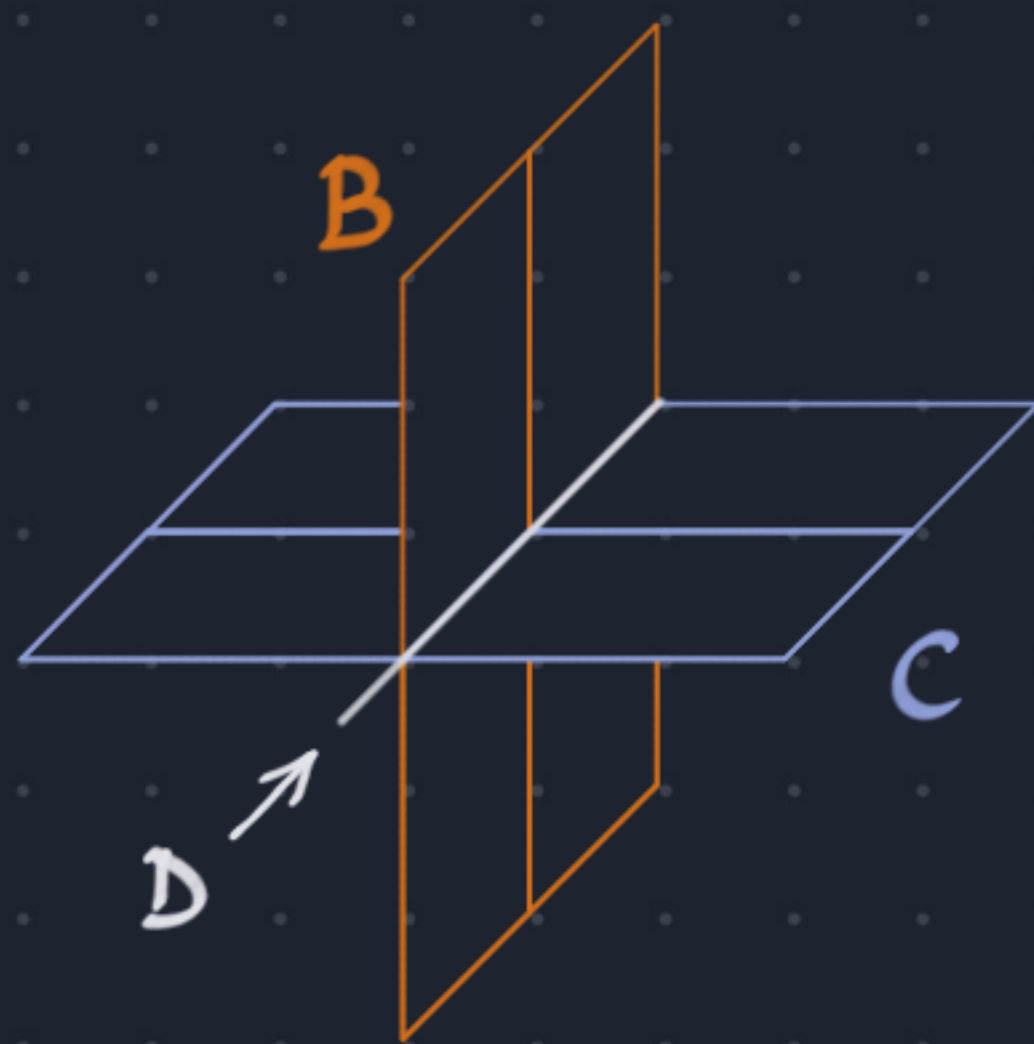
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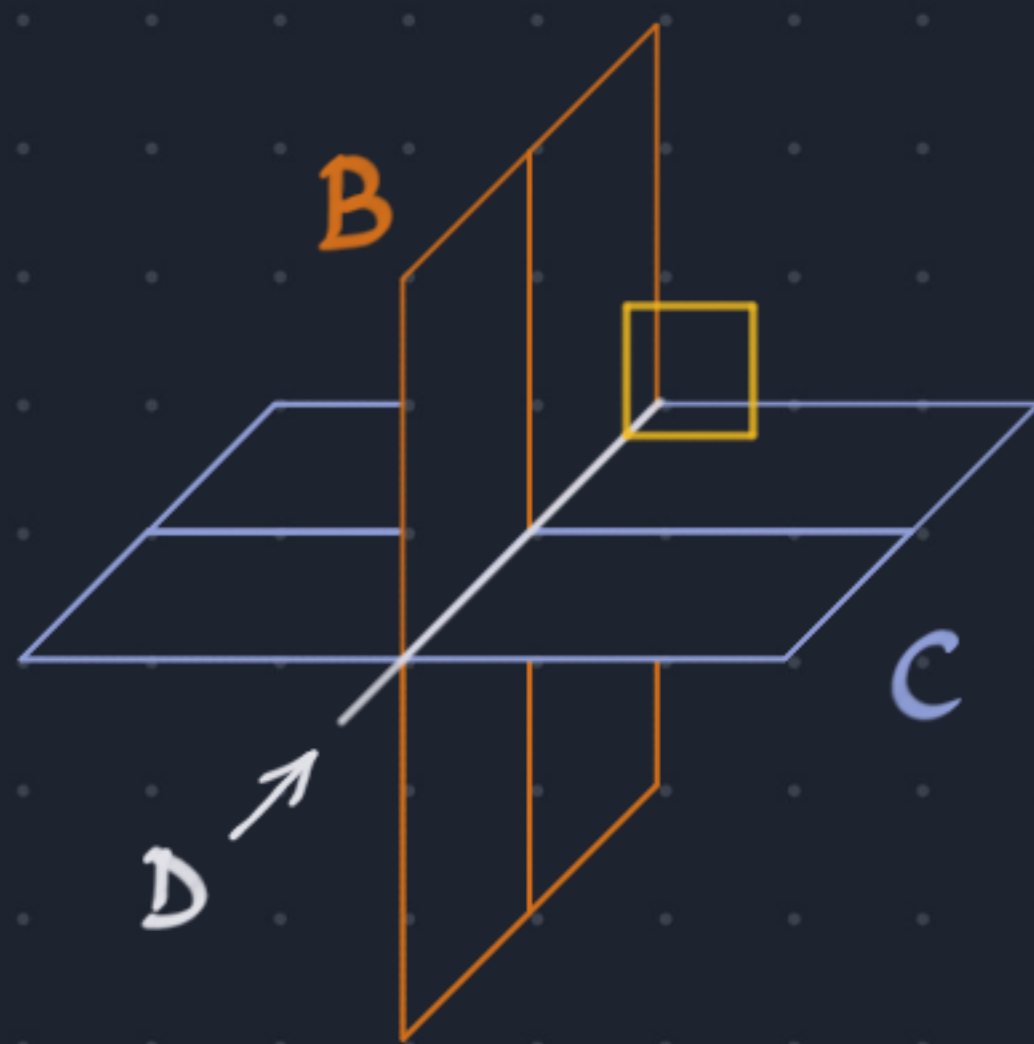
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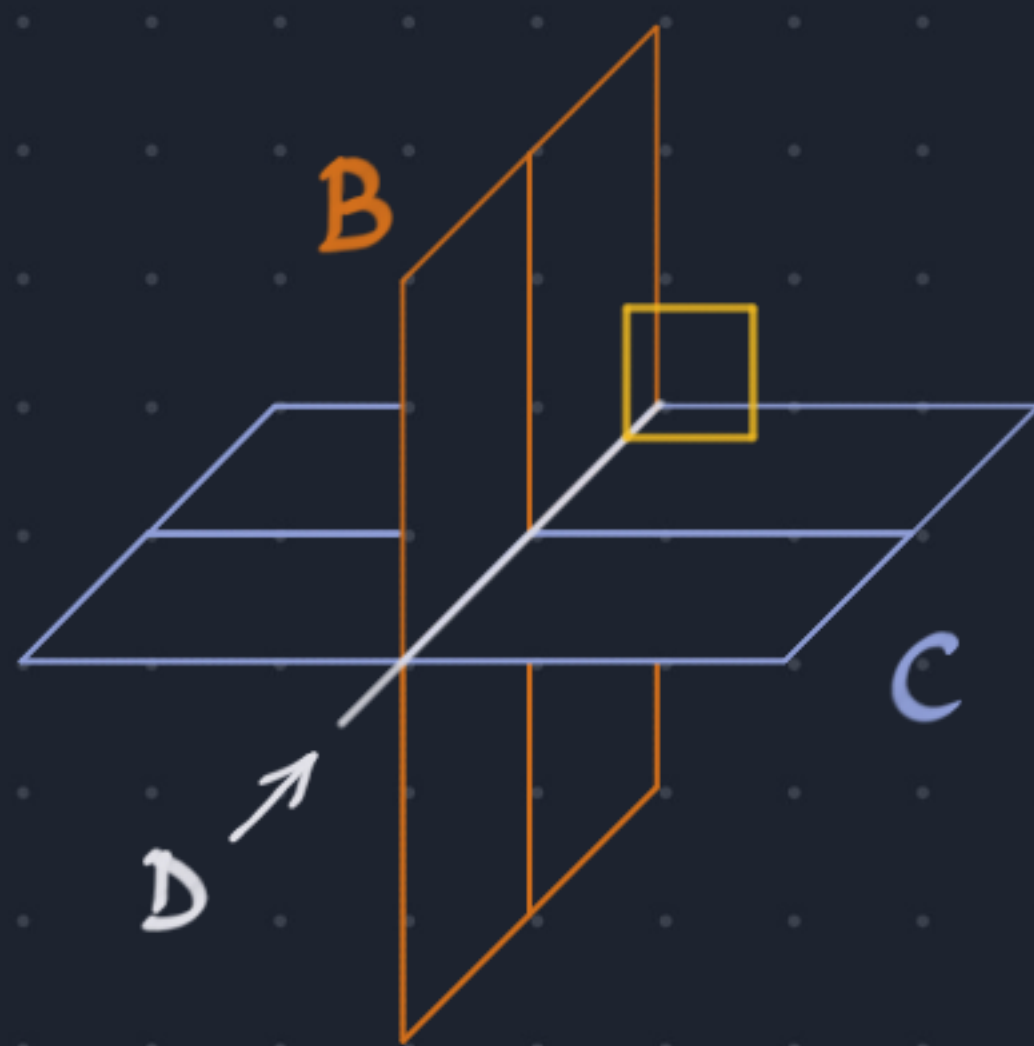
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for $\dagger$ -epis, any D : relatively orthogonal subspaces



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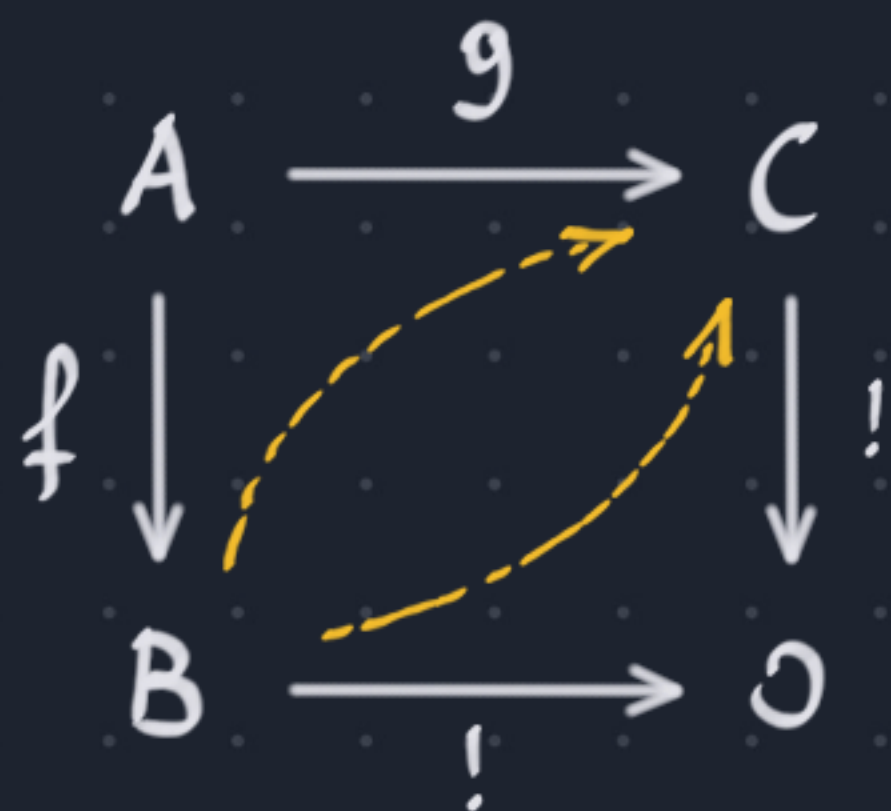
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for $\dagger$ -epis, $D = 0$: orthogonal subspaces

for $\dagger$ -epis, any D : relatively orthogonal subspaces

· In $\mathbf{Krn}$, for $\dagger$ -epis (= det. maps / RVs), $D = 0$: **independent RVs**



$$\mathbb{P}[f=b, g=c] = \mathbb{P}[f=b] \cdot \mathbb{P}[g=c]$$

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Examples. · In Rel, with functions: weak pullback

· In Mat, for vectors, $D = 0$: usual orthogonality

for $\dagger$ -epis, $D = 0$: orthogonal subspaces

for $\dagger$ -epis, any D : relatively orthogonal subspaces

· In $\mathbf{Krn}$, for $\dagger$ -epis (= det. maps / RVs), $D = 0$: independent RVs

any D : conditionally indep. RVs

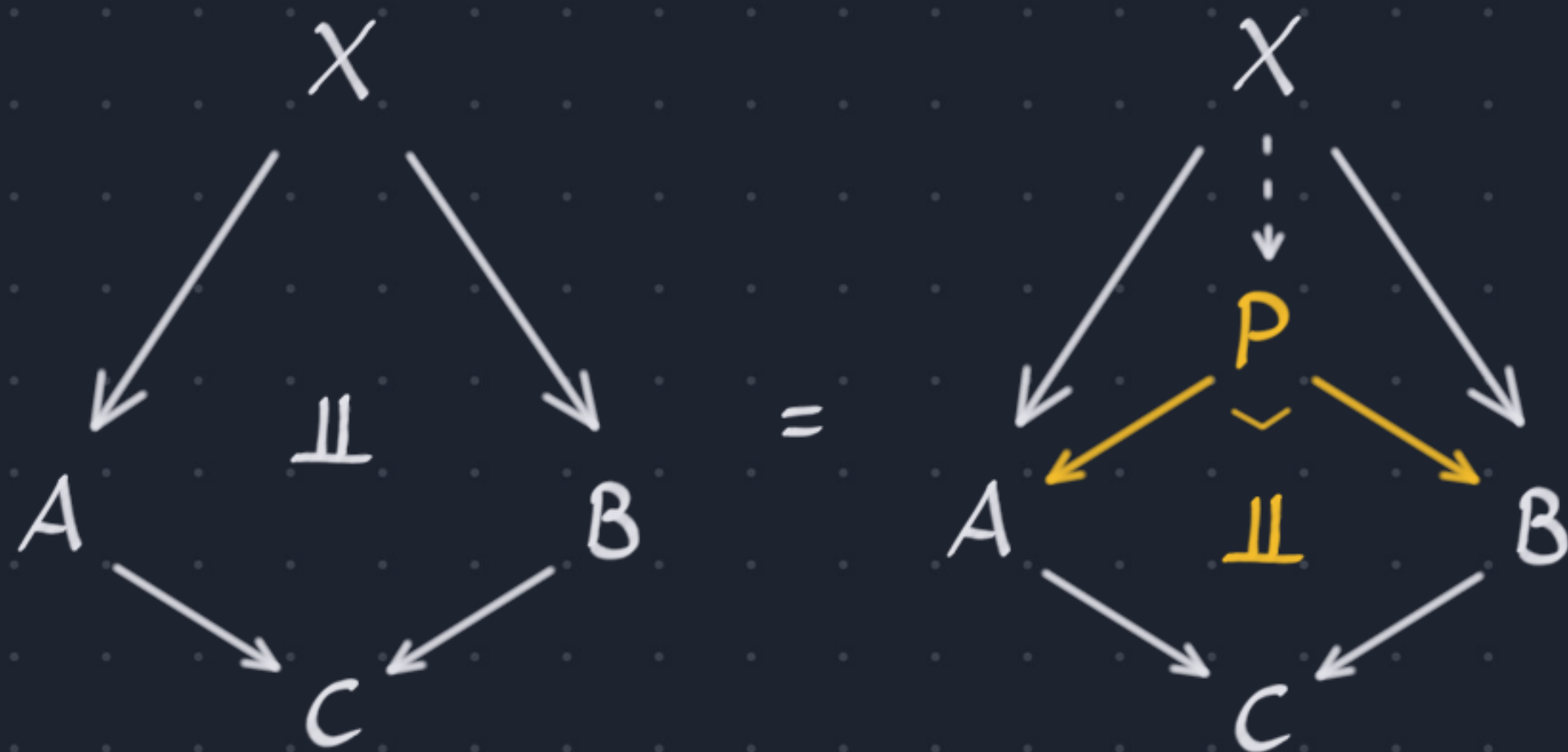


$$\mathbb{P}[f=b, g=c \mid u \cdot f = v \cdot g = d]$$

$$= \mathbb{P}[f=b \mid u \cdot f = d] \cdot \mathbb{P}[g=c \mid v \cdot g = d]$$

2. $\perp = \perp\!\!\!\perp$ · Independent squares

Definition. An independent pullback^{*} is a universal independent square of $\dagger$ -epis.

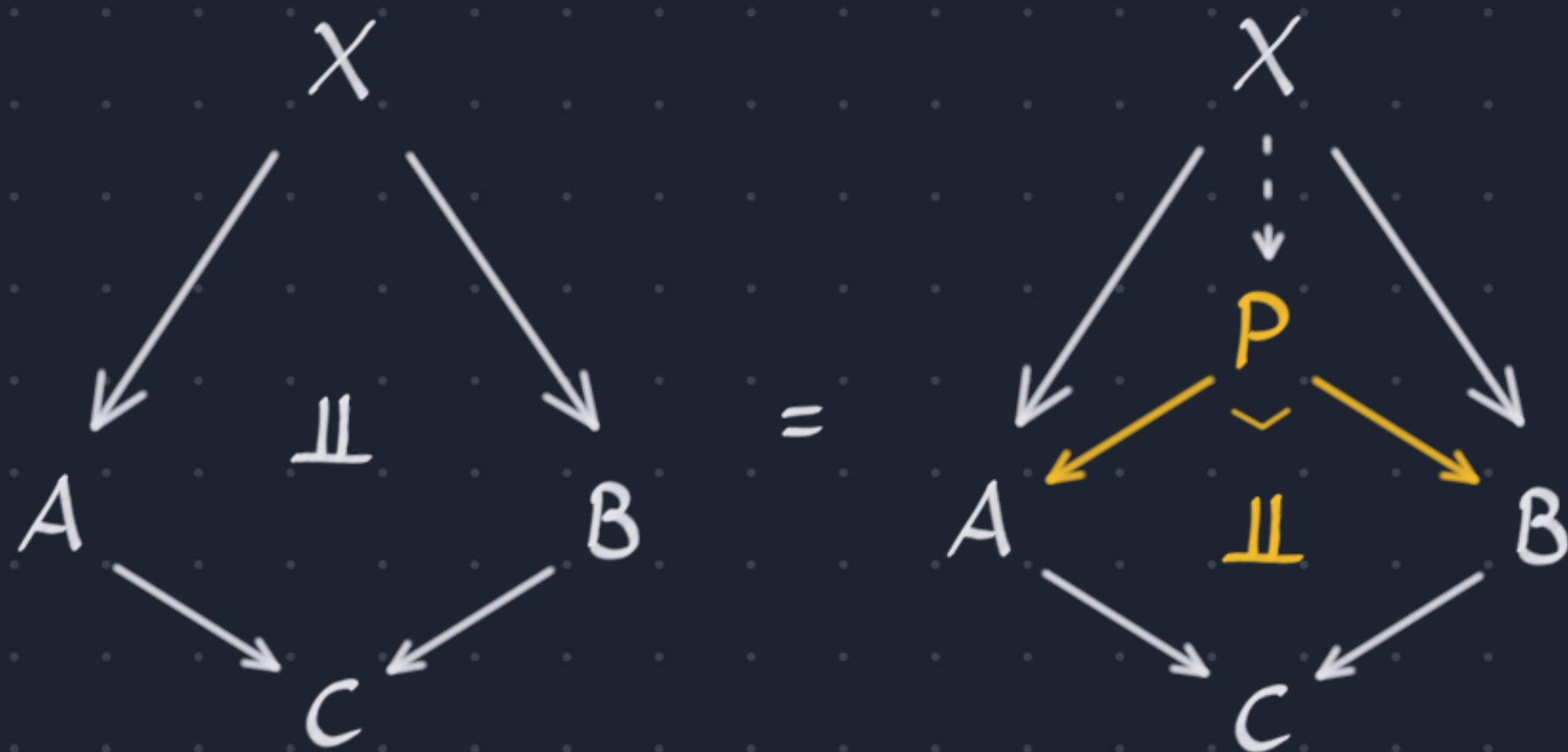


* not (always)
a pullback!

If $C = 0$: independent product.

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Examples.

- In $\mathbf{Mat}$, **orthogonal** (relative) direct sums
- In $\mathbf{Kin}$, (conditional) products of probability spaces.

2. $\perp = \perp\!\!\!\perp$ · Independent squares

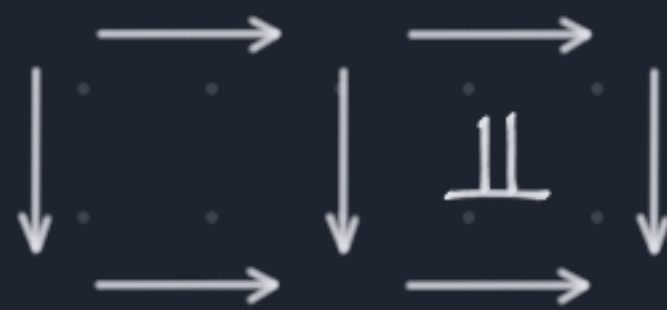
Remark. There are mathematical structures for "independence":

- Matroids [Whitney '35, Mac Lane '36] ← combinatorics of linear independence
- Graphoids [Pearl-Paz '85] ← combinatorics of stochastic independence
- Separoids [Dawid 2001] ← common generalization

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- Separoids [Dawid 2001] ← common generalization
- Independence Structures [Simpson 2018]: choice of particular squares



Question. Do our independent squares form an independence structure?

1. MOTIVATION

2. $\perp = \perp\!\!\!\perp$

3. THE EQUIVALENCE

3. THE EQUIVALENCE · Epi-Regular ind. categories

Definition. [Carboni - Walters] A category is *regular* if:

1) Every cospan has a pullback



2) Every span has a
(strong epi - jointly monic) factorization



3) Strong epis are pullback-stable

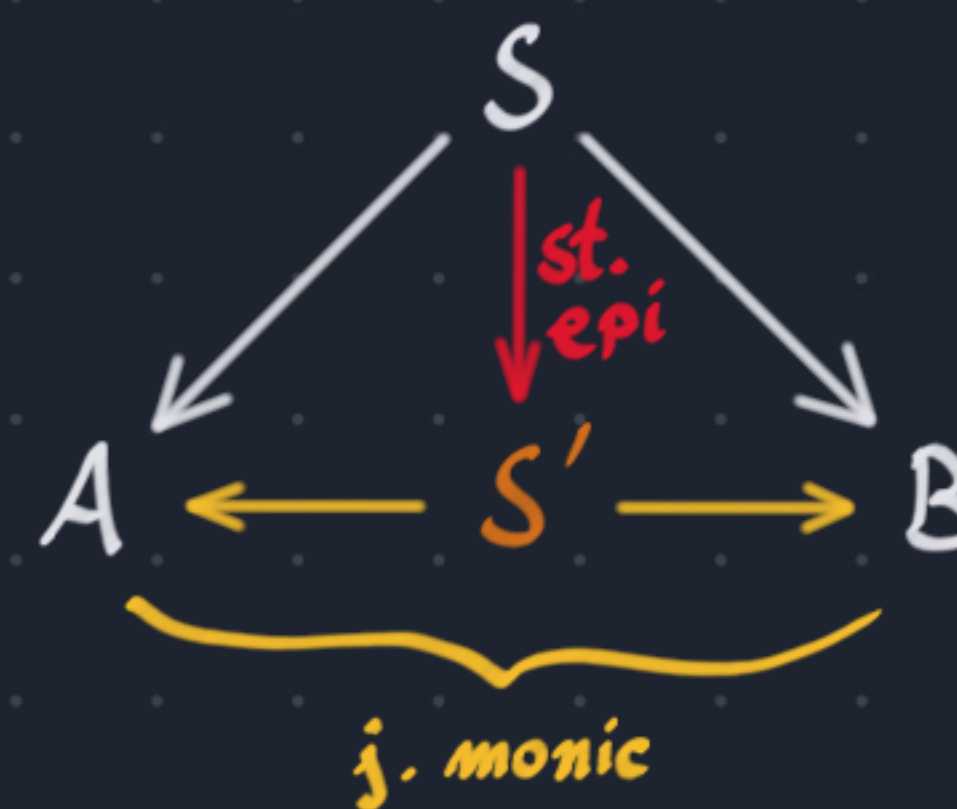
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3) ~~Strong epis are pullback stable~~

Every morphism is strong epi!

3. THE EQUIVALENCE · Epi-Regular ind. categories

- Examples.
- Prob: (std. Borel) prob. spaces, meas.-pres. (det.) maps [Simpson]
 - Hilbert spaces, coisometries (same for Mat)

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Properties.

- Every morphism is the coequalizer of its **independent** kernel pair

· f monic

$\iff$



3. THE EQUIVALENCE · Dilations & Dilators

Idea. Expressing a relation as a "table" (= span)

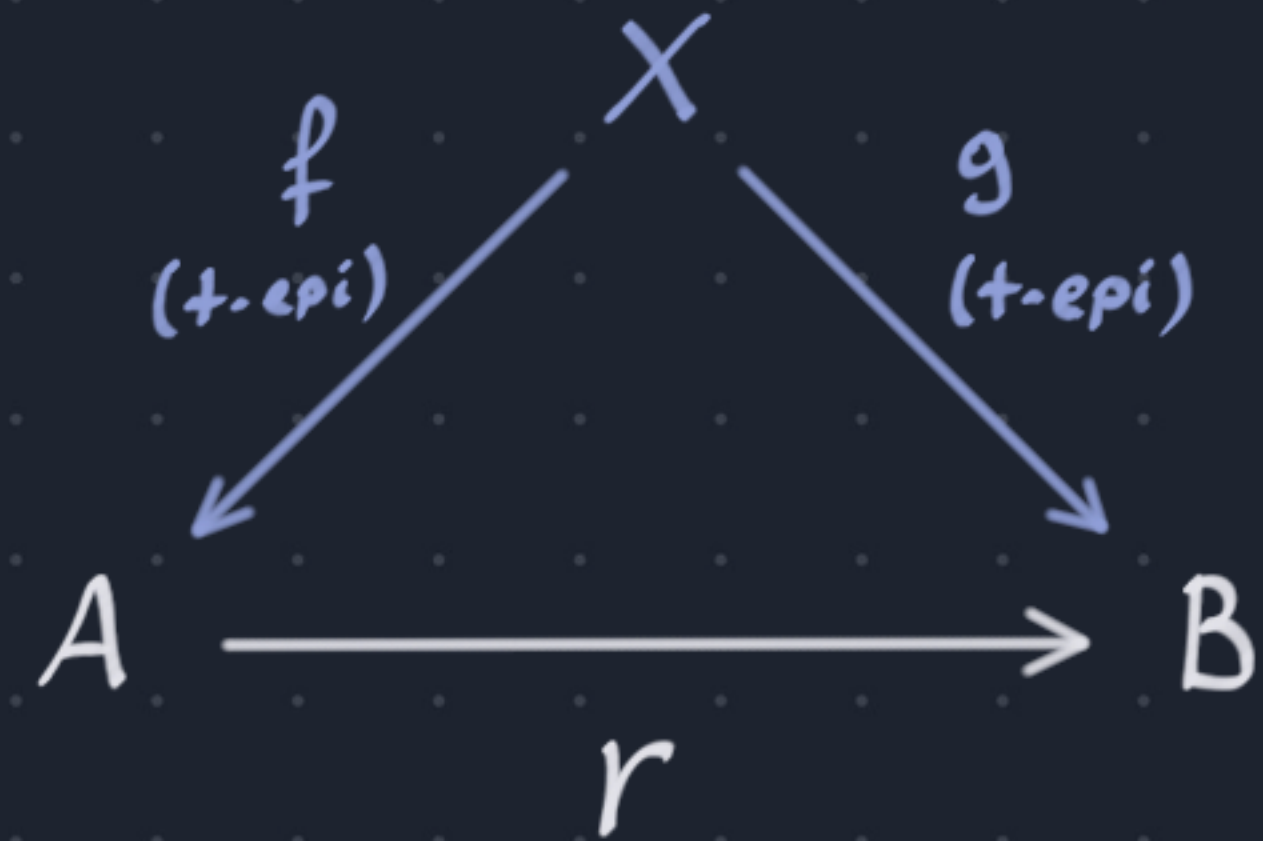


Classically: **tabulator** (in allegories, double cats...)

3. THE EQUIVALENCE · Dilations & Dilators

Definition. Given $A \xrightarrow{r} B$ in a dagger category,

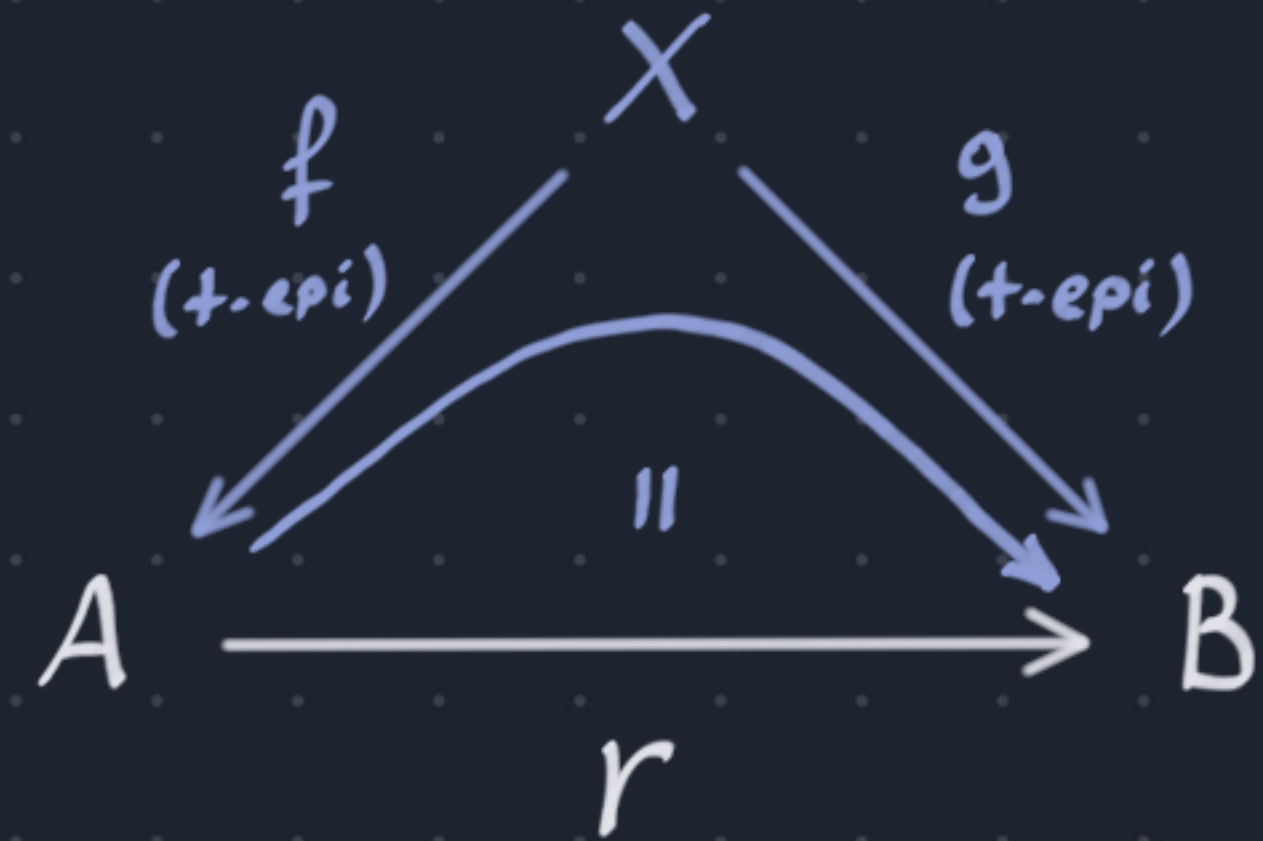
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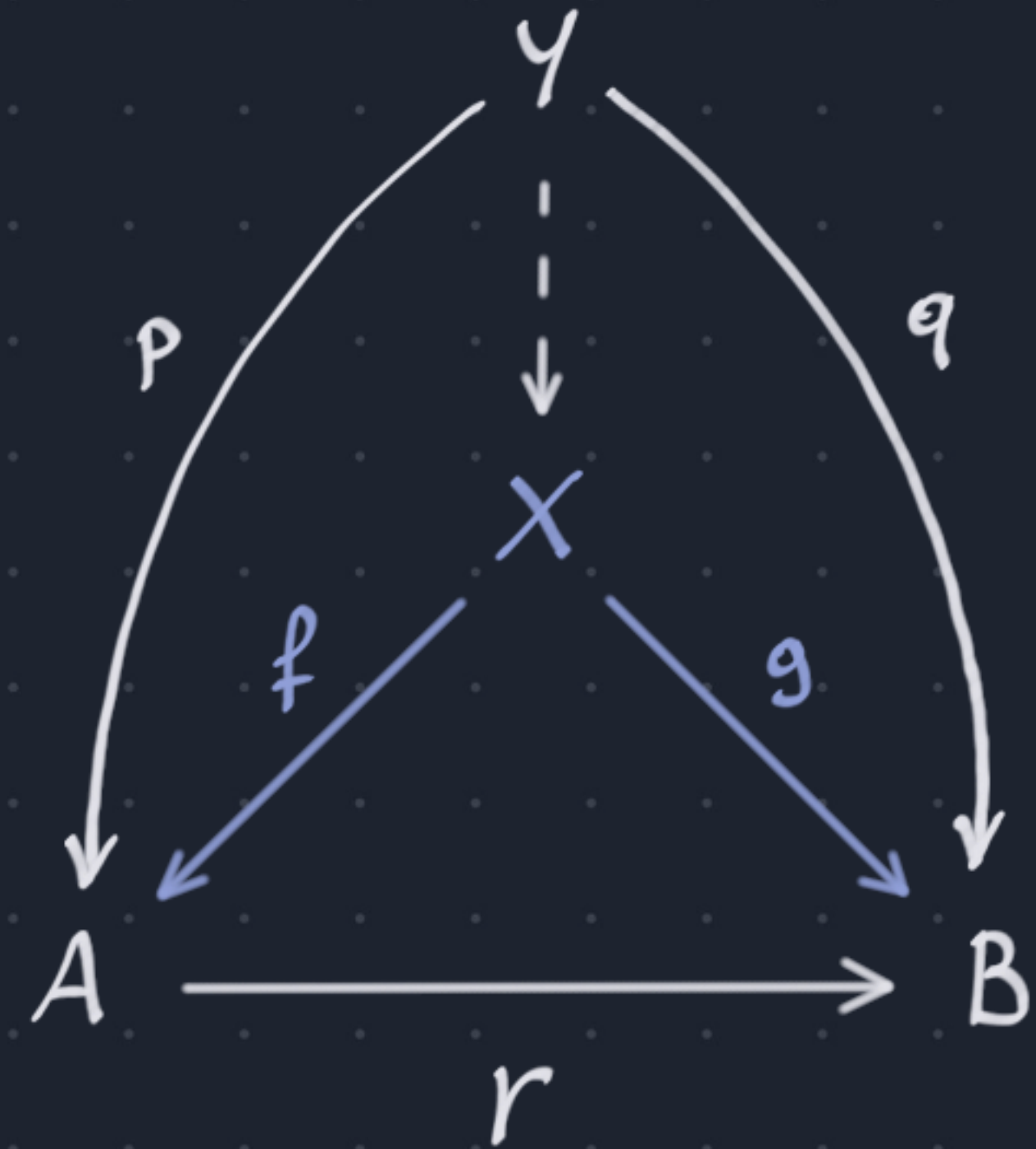
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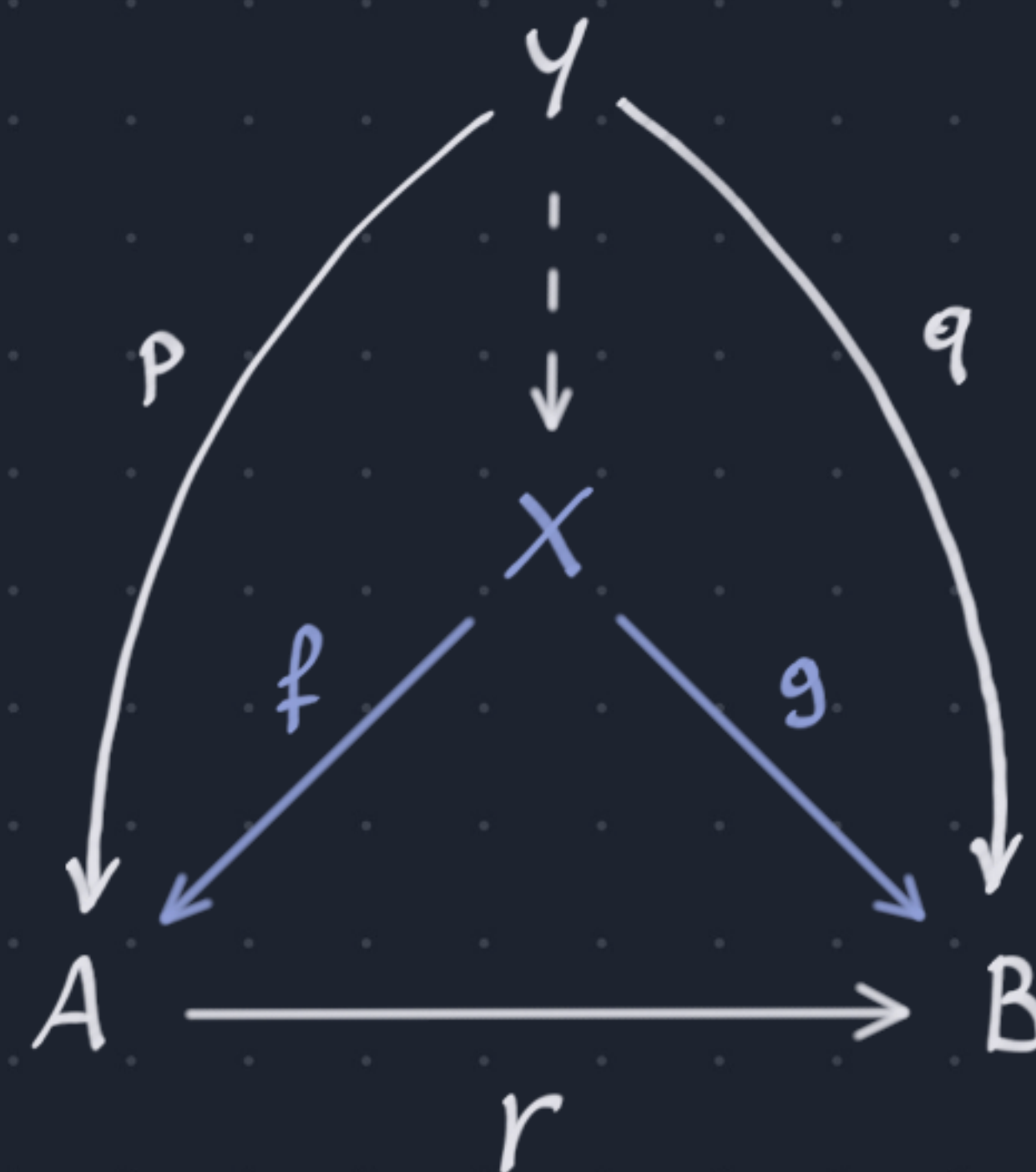
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- A **dilation** of r is a span (f, g) of dagger epis s.t. $gf^+ = r$.
- A **dilator** of r is a universal dilation.



- A dagger category is **dilatory** if every morphism has a dilator.

In subcat. of $\dagger$ -epis, dilator = jointly monic dilation.

3. THE EQUIVALENCE · Dilations & Dilators

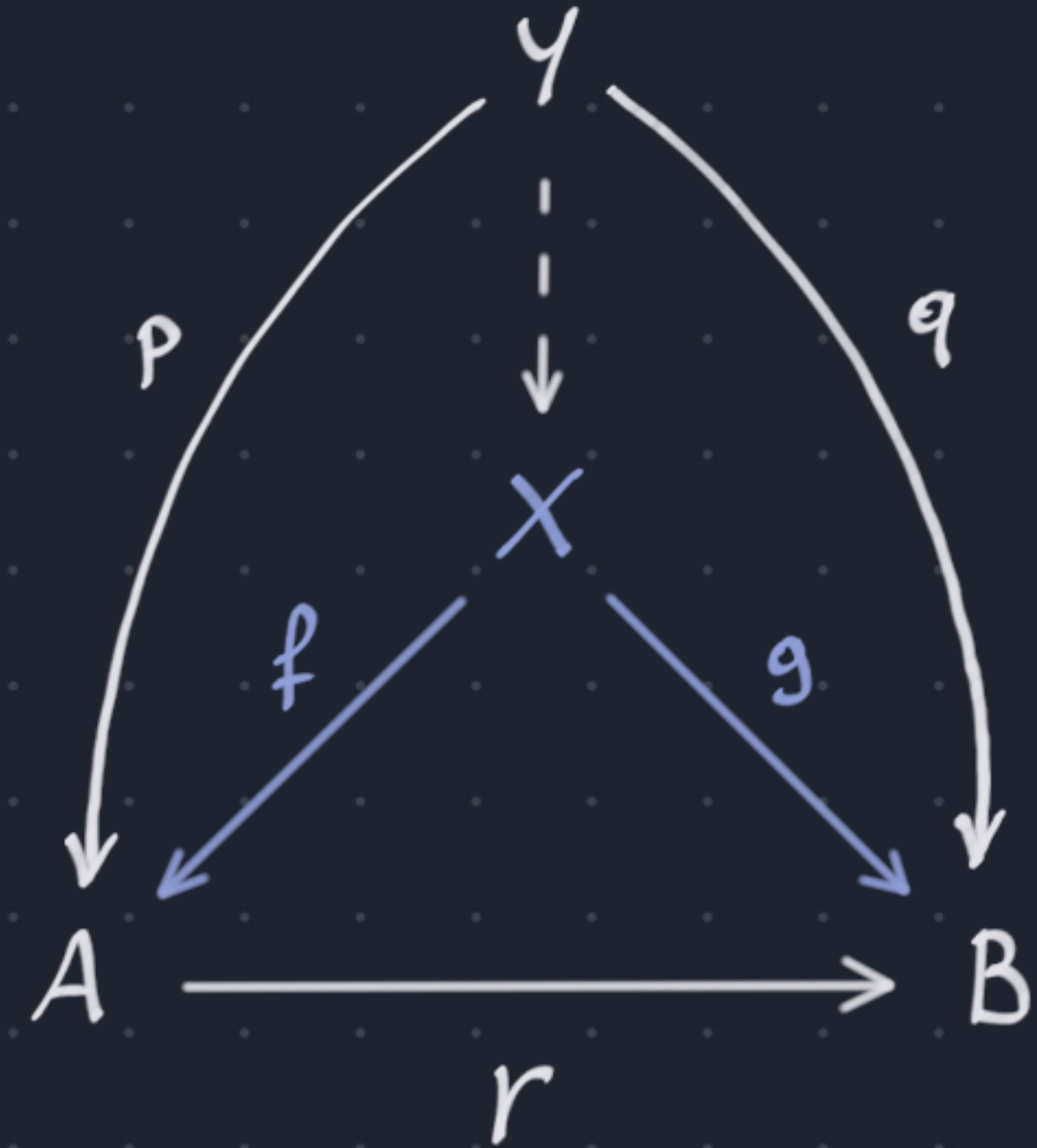
Examples.

· $\text{Mat}_{\leq 1}$ has dilators:

rel. to Choleski decomposition of matrices

· Krn has dilators:

bloom-shriek factorization of kernels



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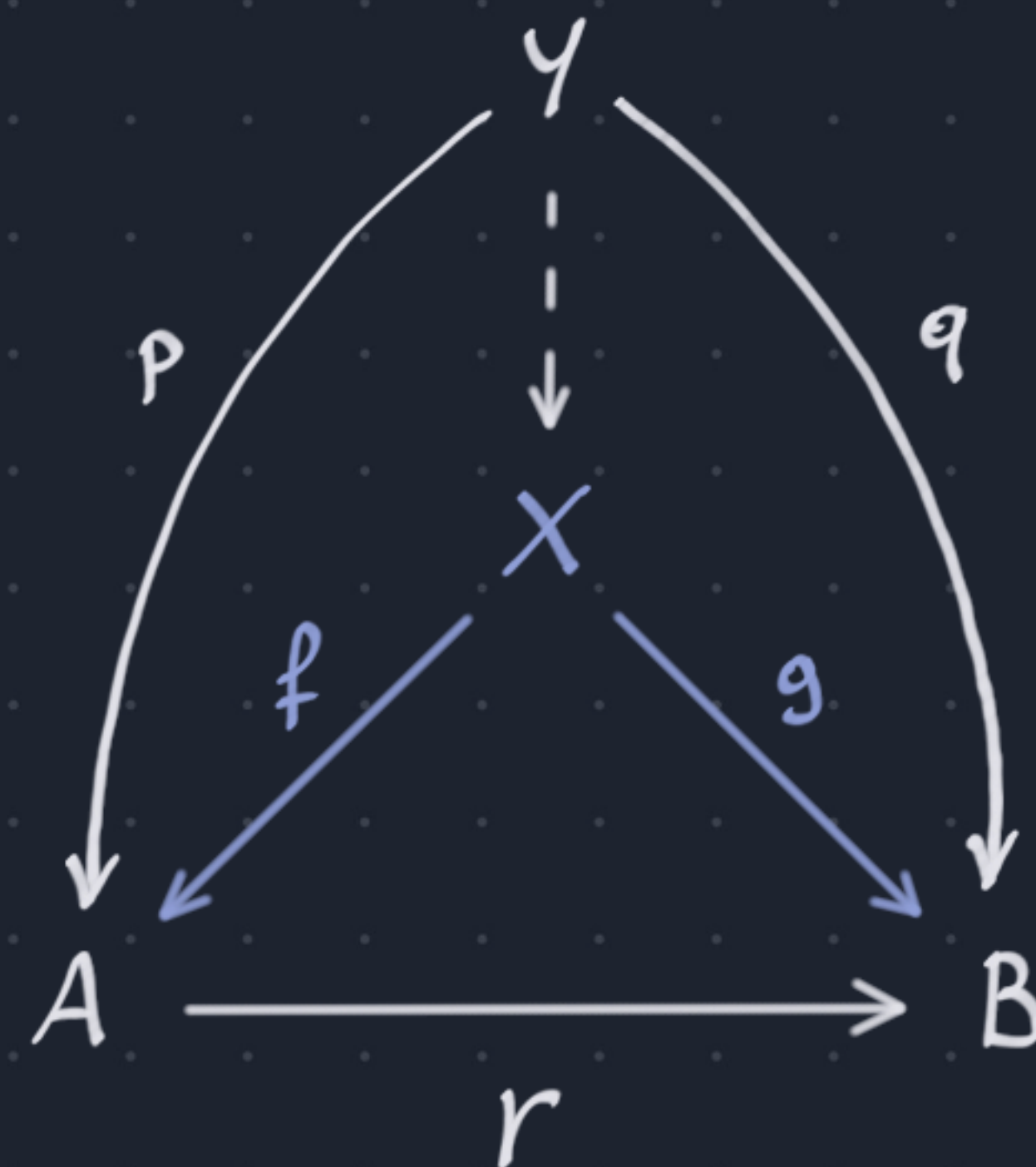
- Krn has dilators:

bloom-shriek factorization of kernels

- The following subcategories of Rel have dilators:

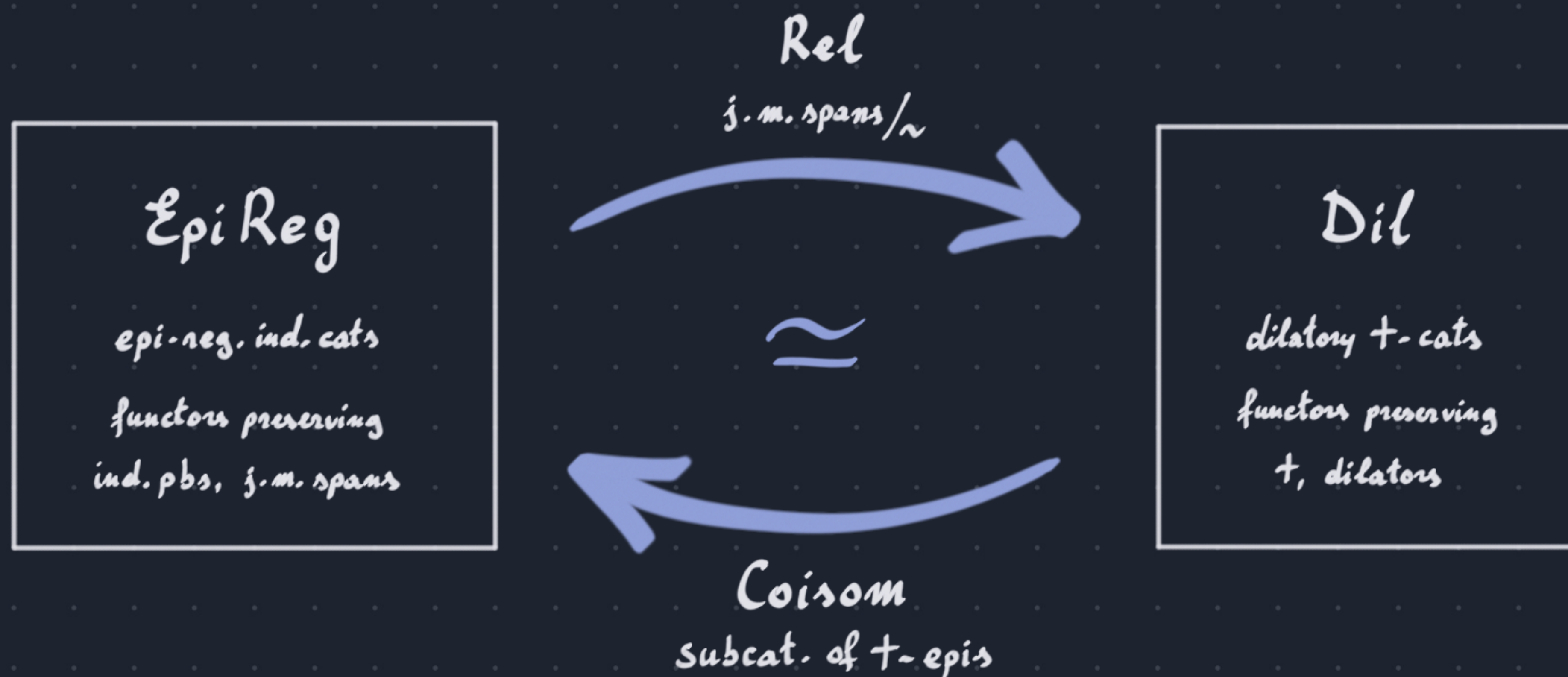
- MSurj - surjective multi-valued functions (†-epi: single-valued)

- PInj - injective partial functions (†-epi: co-total)



3. THE EQUIVALENCE • Main theorems

Theorem 1. We have an adjoint equivalence of categories

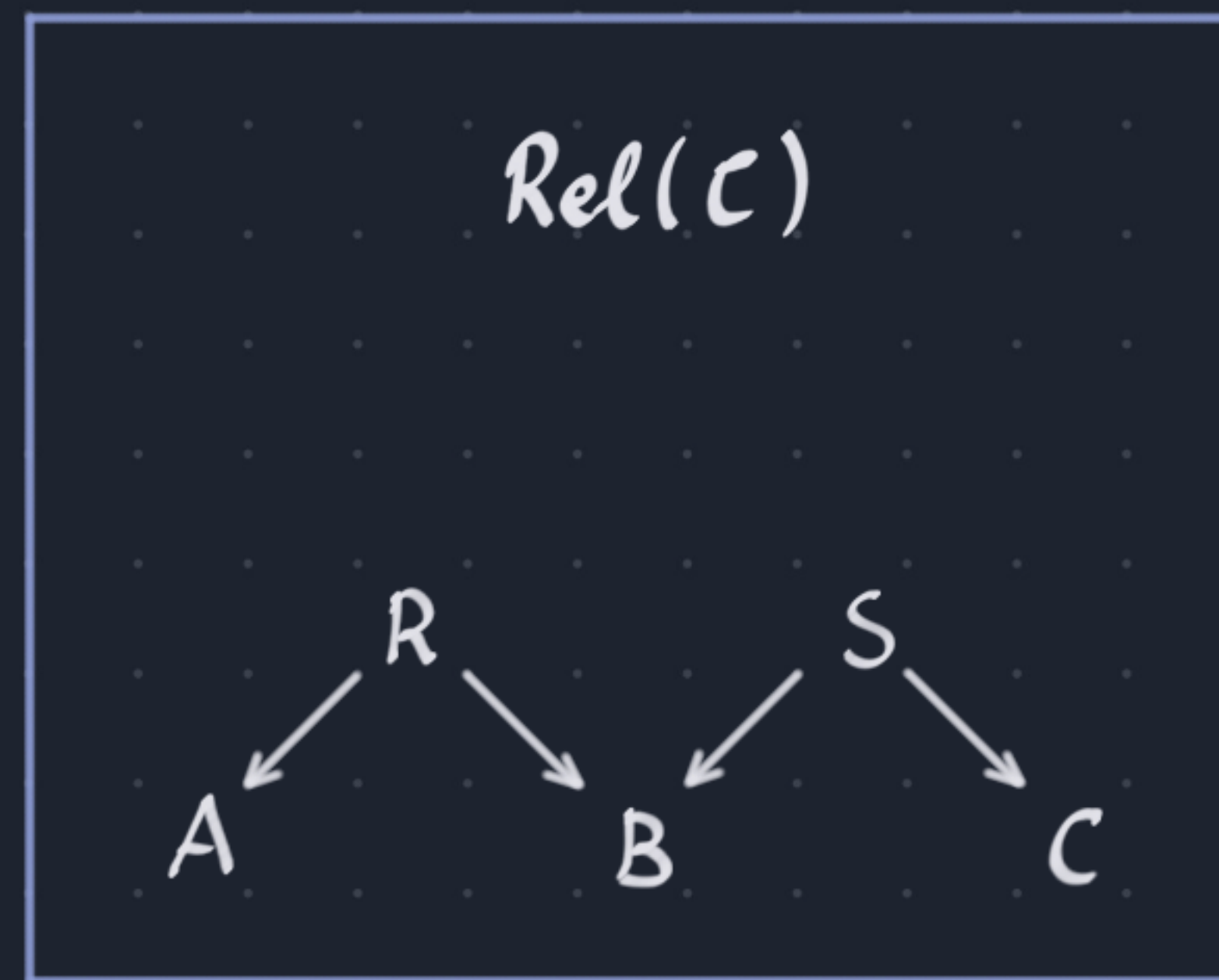


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$\mathcal{C}$

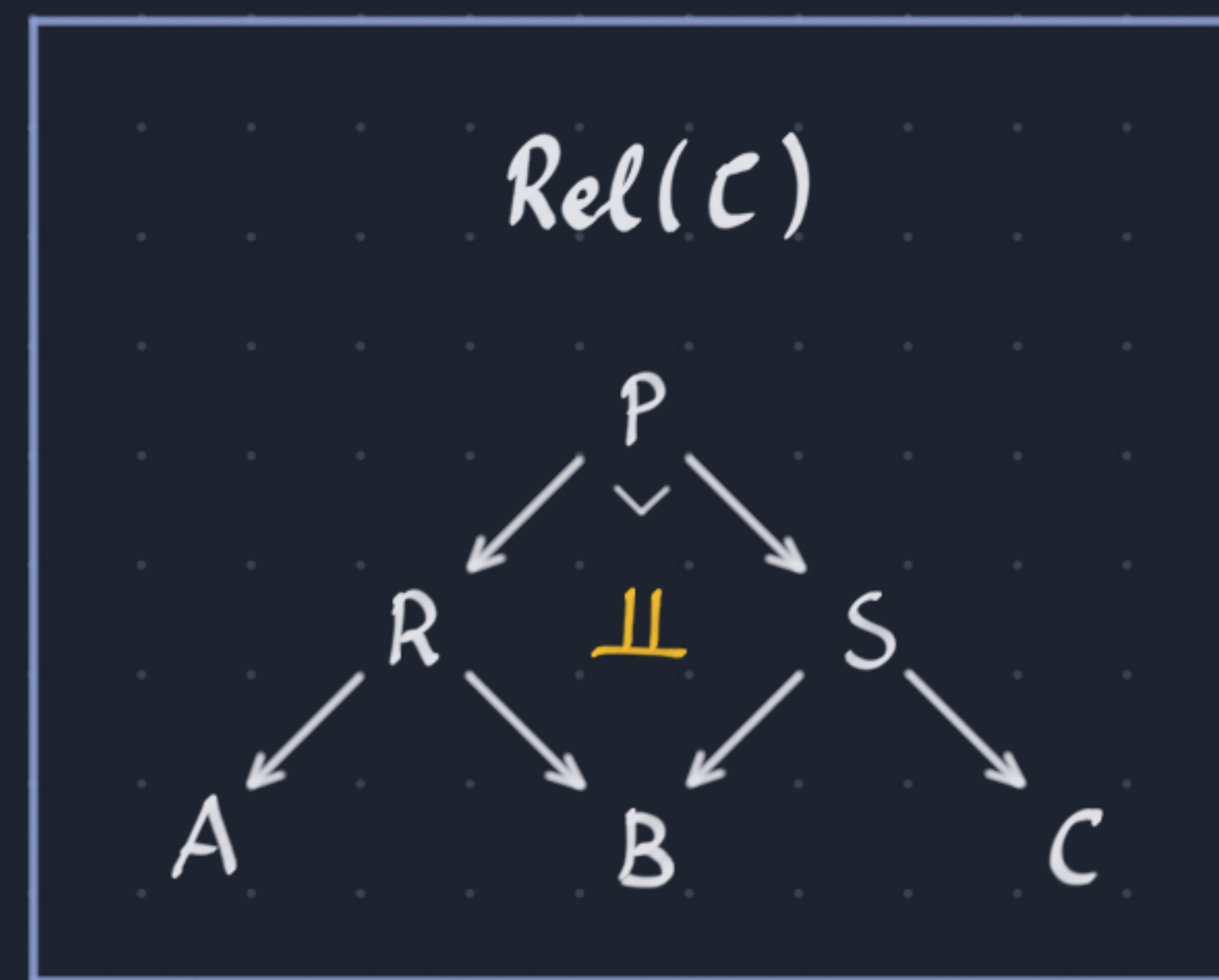


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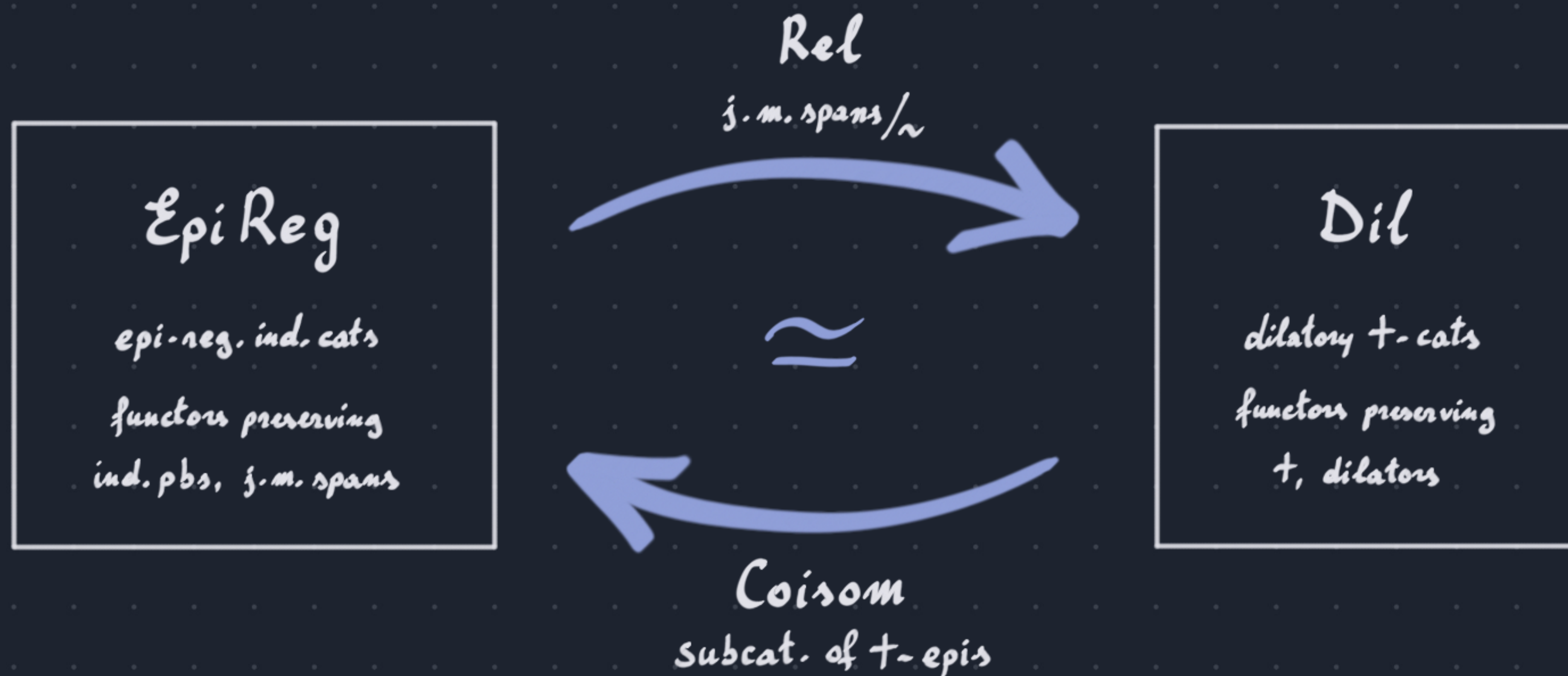


$\mathcal{C}$



3. THE EQUIVALENCE • Main theorems

Theorem 1. We have an adjoint equivalence of categories



3. THE EQUIVALENCE • Main theorems

Theorem 2. We have an adjoint equivalence of 2-categories

Epi Reg

epi-reg. ind. cats
functors preserving ind. pbs, j.m. spans

Def. Independent nat. transf.:

$$\begin{array}{ccc} \mathcal{C} & \xrightarrow{F} & \mathcal{D} \\ \downarrow \alpha & & \downarrow \alpha \\ \mathcal{C} & \xrightarrow{G} & \mathcal{D} \end{array} : \begin{array}{ccc} FA & \xrightarrow{Ff} & FB \\ \alpha_A \downarrow & \perp & \downarrow \alpha_B \\ GA & \xrightarrow{Gf} & GB \end{array}$$

Rel

j.m. spans / $\sim$

$\cong$

Coisom

subcat. of $\dagger$ -epis

Dil

dilatory $\dagger$ -cats
functors preserving $\dagger$, dilators

Def. Natural coisometries:

$$\begin{array}{ccc} \mathcal{C} & \xrightarrow{F} & \mathcal{D} \\ \downarrow \alpha & & \downarrow \alpha \\ \mathcal{C} & \xrightarrow{G} & \mathcal{D} \end{array} : \begin{array}{ccc} FA & \xrightarrow{Ff} & FB \\ \alpha_A \downarrow & \dagger\text{-epi} & \downarrow \alpha_B \\ GA & \xrightarrow{Gf} & GB \end{array}$$

3. THE EQUIVALENCE • Main theorems

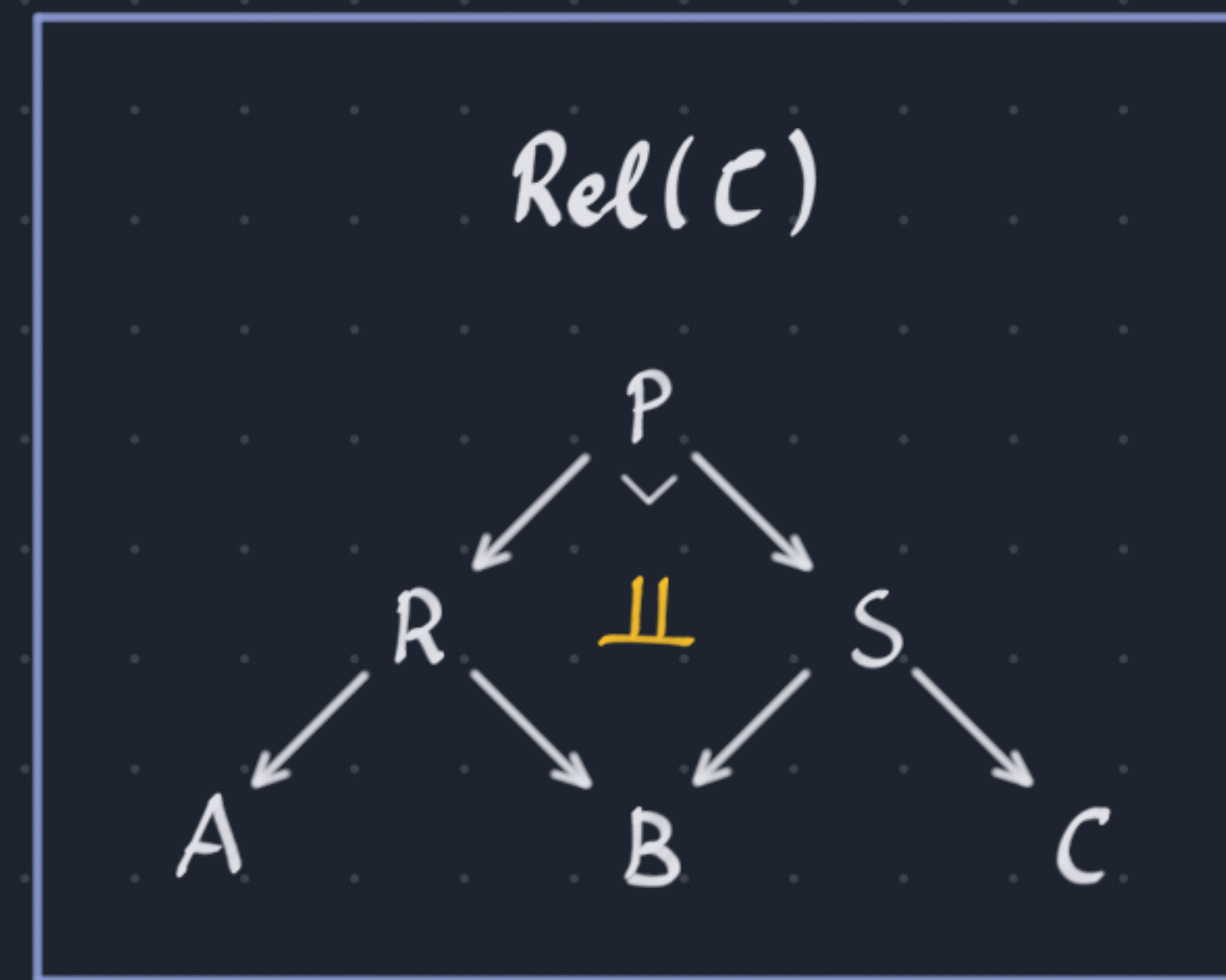
Examples.

• $\mathbf{Krn} \approx$ "category of couplings"

• Same for $\mathbf{ProbStoch}(\mathcal{C})$, where

$\mathcal{C} =$ Markov cat. w. conditionals.

• $\mathbf{Hilb}_{\leq 1}, \mathbf{Mat}_{\leq 1} \approx$ cospans of isometries



3. THE EQUIVALENCE • Main theorems

Examples.

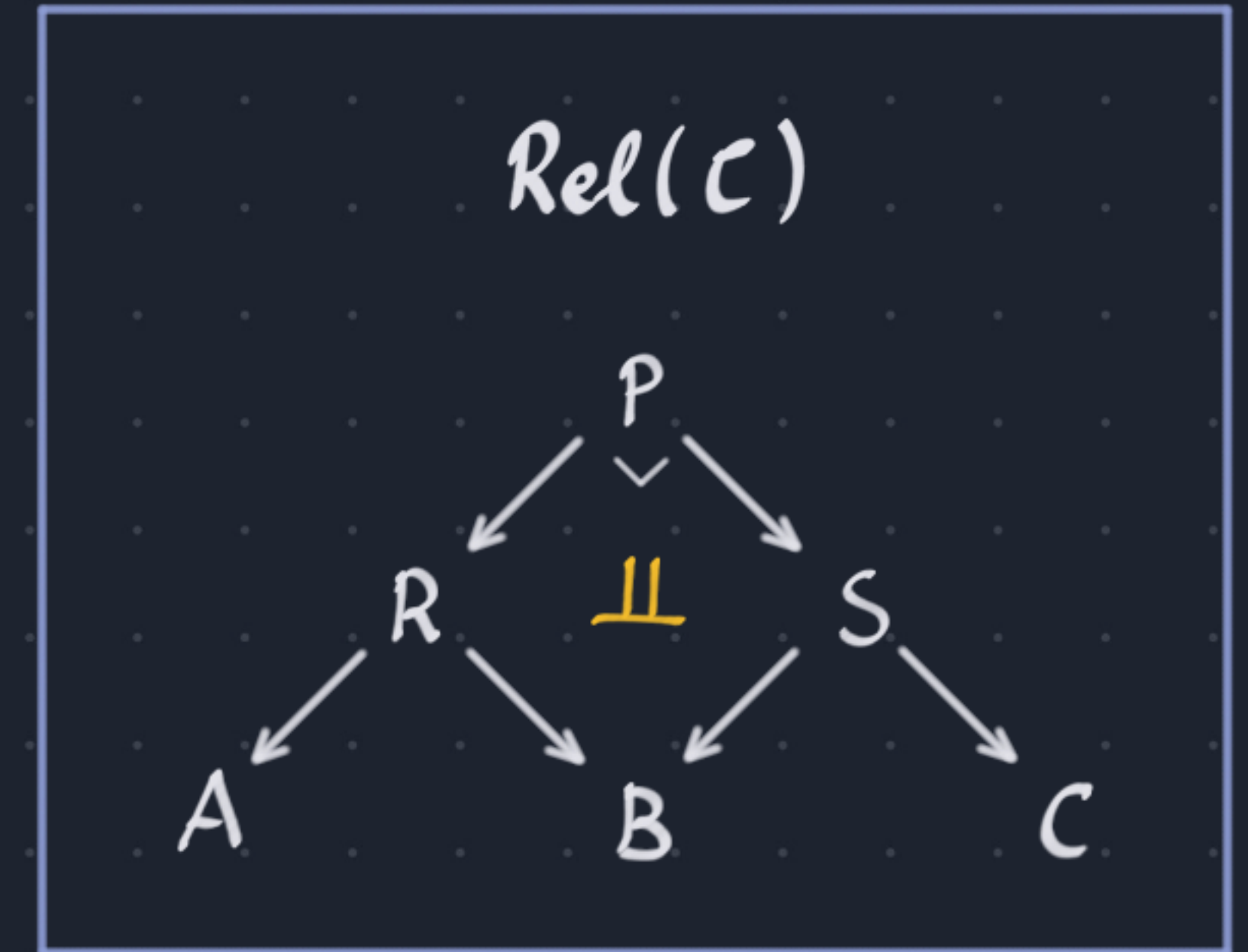
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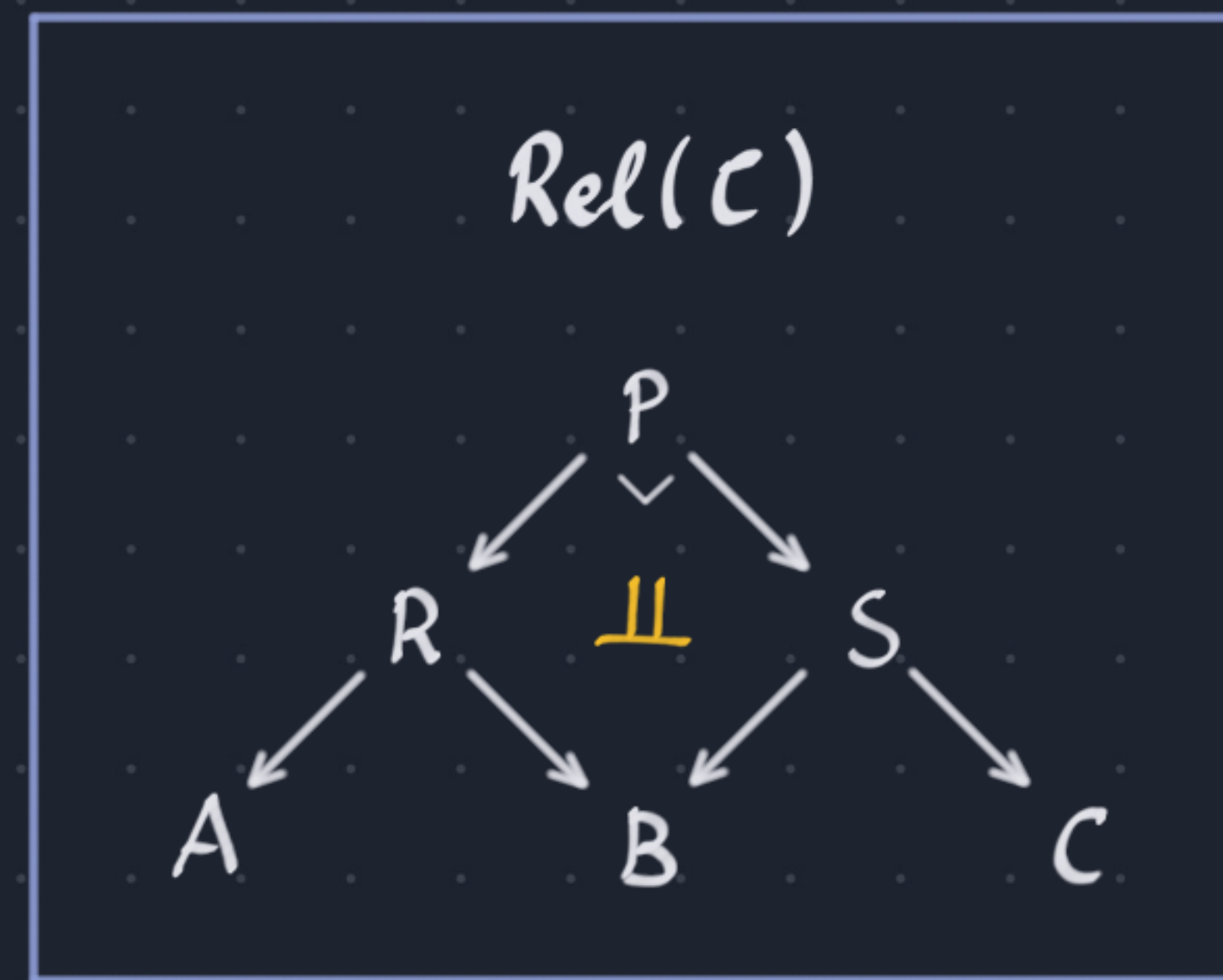
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• $\mathbf{PInj} \approx$ cospans of inj. functions $\rightarrow$ True in every restriction cat. (Ask JS!)



3. THE EQUIVALENCE · Summary

CONCEPT → SETTING ↓	CATEGORY OF "RELATIONS"	
TRADITIONAL	ALLEGORY	
OURS	+ - CATEGORY	

3. THE EQUIVALENCE · Summary

CONCEPT → SETTING ↓	CATEGORY OF "RELATIONS"	"FUNCTION"	
TRADITIONAL	ALLEGORY	MAP (adjoint) $f \mapsto f^+$	
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OURS	$\vdash$ - CATEGORY	$\vdash$ - EPI (coisometry) $f f^+ = \text{id}$	DILATION, DILATOR	INDEPENDENT PULLBACK

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OURS	$\dagger$ - CATEGORY	$\dagger$ - EPI (coisometry) $ff^+ = \text{id}$	DILATION, DILATOR	INDEPENDENT PULLBACK	$\dagger$ - IDEMPOTENT $ff = f$ $f = f^+$

3. THE EQUIVALENCE · Summary

Open Questions

1) Common generalization?

1a. Are there ~~epi~~-regular independence categories?

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2) [Recall: $+ \text{-epi} \Leftrightarrow \text{coequalizer of independent kernel pair}$]

"Independent descent theory"? (Simpson & Stein have studied the probability case.)

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2) [Recall: $\perp$ -epi $\Leftrightarrow$ coequalizer of independent kernel pair]

"Independent descent theory"? (Simpson & Stein have studied the probability case.)

3) "Independent regular logic"? $\Rightarrow$ Would apply to probability!

References

Main results:

- M. Di Meglio, C. Heunen, JS Lemay, P. Perrone, D. Stein, Dagger Categories of Relations: The equivalence of dilatory dagger categories and epi-regular independence categories. **COMING SOON.**

Further references:

- D. Stein, Random Variables, Conditional Independence, and Categories of Abstract Sample Spaces. LICS 2025. [arxiv:2503.02477](#)
- A. Simpson, Category-Theoretic Structure for Independence & Conditional Independence. ENTCS 2018.