

**Tao Gu**, Fabio Zanasi  
University College London

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# Functorial Semantics as a Unifying Perspective on Logic Programming

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- ▶ Background
  - ▶ A string diagram perspective of PLP
  - ▶ Application: transformation between PLP and BN
  - ▶ Classical logic programs, Weighted logic program
  - ▶ Future work

**Background**

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## Probabilistic logic programs (PLP)

- ▶ A PLP clause  $\psi$  based on  $At$  is  $p :: A \leftarrow L_1, \dots, L_m .$ , where:
  - ▶  $p \in [0,1]$
  - ▶  $A \in At$  is the **head**
  - ▶  $\{L_1, \dots, L_m\} \subseteq At \cup \neg At$  is the **body**
- ▶ A PLP program  $\mathbb{P} = \{\psi_1, \dots, \psi_n\}$ .

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## PLP example

► **Example.**  $\mathbb{P}_{wet}$  consists of the following 8 clauses:

0.25 ::	Winter	← .	( $\psi_1$ )	0.9 ::	WetGrass	← Sprinkler.	( $\psi_5$ )
0.2 ::	Sprinkler	← Winter.	( $\psi_2$ )	0.8 ::	WetGrass	← Rain.	( $\psi_6$ )
0.6 ::	Rain	← Winter.	( $\psi_3$ )	0.7 ::	SlipperyRoad	← Rain.	( $\psi_7$ )
0.1 ::	Rain	← $\neg$ Winter.	( $\psi_4$ )	0.1 ::	SlipperyRoad	← $\neg$ Rain.	( $\psi_8$ )

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# Semantics of acyclic PLP

- ▶ Idea: **distribution semantics** + **negation as failure**
- ▶ **Distribution semantics**
  - ▶  $\mathbb{P}$  determines a distribution  $\mu_{\mathbb{P}}$  on  $\mathcal{P}(|\mathbb{P}|)$
  - ▶  $Pr_{\mathbb{P}}(A) = \sum \{\mu_{\mathbb{P}}(Q) \mid Q \subseteq |\mathbb{P}|, Q \models A\}$
- ▶ **Negation as failure**
  - ▶  $\neg A$  is true in  $\mathbb{P}$  if it is *not* finitely derivable/provable in  $\mathbb{P}$
- ▶ We focus on **acyclic PLP**:
  - ▶ Relatively simple semantics (imprecise probabilities needed).
  - ▶ Expressive power: equivalent as boolean-valued Bayesian networks.

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## Functorial semantics

- ▶ F. W. Lawvere, *Functorial Semantics of Algebraic Theories*, Ph.D. thesis, Columbia University, 1963.
- ▶ Idea: algebras are functors (models) from syntax categories (e.g. Lawvere theories) to semantics categories (e.g. **Set**).
- ▶ **Example.** Monoids  $\simeq \text{Prod}(\mathcal{L}_{Mon}^{op}, \mathbf{Set})$ , where  $\mathcal{L}_{Mon}$  has natural numbers as objects, and morphisms  $n \rightarrow m$  are  $\langle t_1, \dots, t_n \rangle$  of free monoids on  $\{x_1, \dots, x_m\}$ .

# Functorial semantics of PLP



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# Categorical perspective of PLP

- ▶ We separate the syntax and semantics:
- ▶ The **syntax** describes the inferential structure
  - ▶ A definite logic program  $\mathbb{L} := [\mathbb{P}]$  describes the inferential structure of  $\mathbb{P}$ : the inferential structure of  $p :: A \leftarrow L_1, \dots, L_m.$  is clause  $A \leftarrow B_1, \dots, B_m.,$  where  $B_i$  is the atom in literal  $L_i.$
  - ▶ **Example.** The inferential structure of  $A \leftarrow B_1, \neg B_2.$  is  $A \leftarrow B_1, B_2. ([A \leftarrow B_1, \neg B_2.] = A \leftarrow B_1, B_2.)$
- ▶ The **semantics** determines the meaning of clauses
- ▶ Both are symmetric monoidal categories with additional structure.
- ▶ *Functorial* semantics: PLPs are exactly structure-preserving functors

$$F : \text{SynCat} \rightarrow \text{SemCat}$$

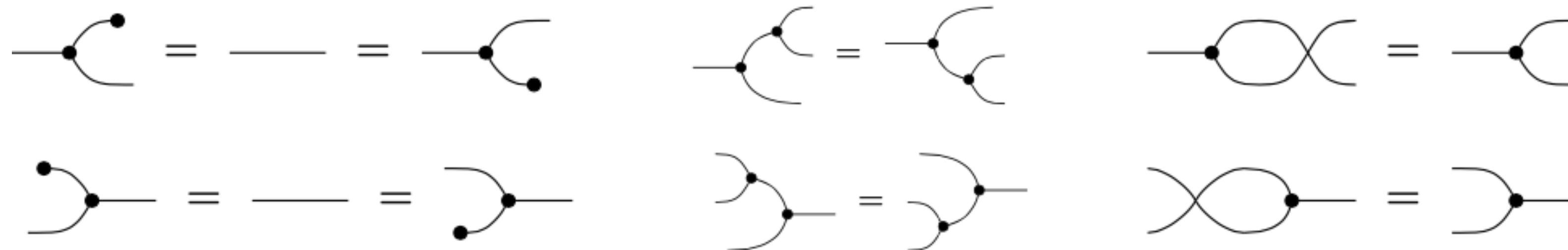
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## CDMU category

- ▶ **Definition.** Every object has **copier**, **discarder**, **multiplication**, **unit** morphisms



- ▶ satisfying:



- ▶ Both the syntax and semantics categories of PLP are CDMU categories.

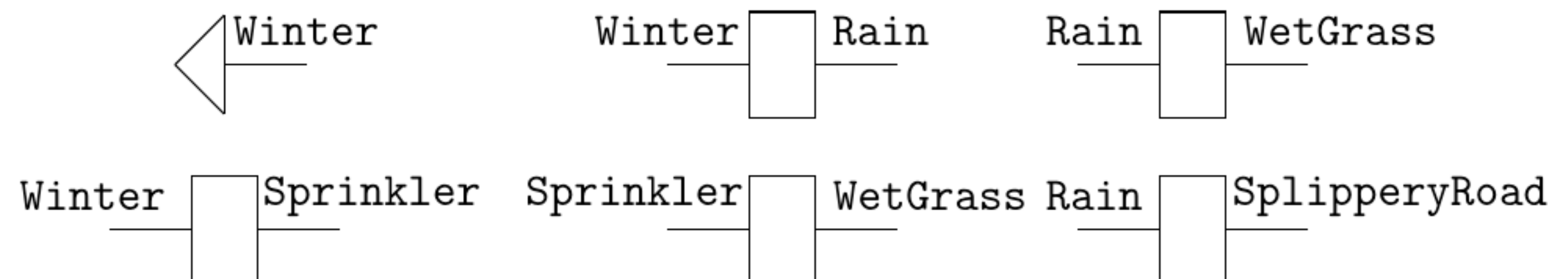
# Syntax category

- ▶  $\mathbb{L} := [\mathbb{P}]$  determines a syntax category  $\text{SynPLP}_{\mathbb{L}} := \text{freeCDMU}(At, \Sigma_{\mathbb{L}})$ , where  $\Sigma_{\mathbb{L}}$  is the set of generating morphisms, one for each  $\mathbb{L}$ -clause:

$$\Sigma_{\mathbb{L}} := \left\{ \begin{array}{c} B_1 \\ \vdots \\ B_m \end{array} \begin{array}{|c} \hline \varphi \\ \hline \end{array} \begin{array}{c} A \end{array} \mid \varphi \equiv A \leftarrow B_1, \dots, B_m \text{ is a clause in } \mathbb{L} \right\}$$

- ▶ Objects: finite lists over  $At$ .
- ▶ Morphisms: freely composed string diagrams using  $\Sigma_{\mathbb{L}}$  and CDMU structure, modulo CDMU equations.

- ▶ **Example.**  $\Sigma_{\mathbb{L}_{wet}}$  consists of 6 boxes:



# Syntax category

- ▶  $\mathbb{L} := [\mathbb{P}]$  determines a syntax category  $\text{SynPLP}_{\mathbb{L}} := \text{freeCDMU}(At, \Sigma_{\mathbb{L}})$ , where  $\Sigma_{\mathbb{L}}$  is the set of generating morphisms, one for each  $\mathbb{L}$ -clause:

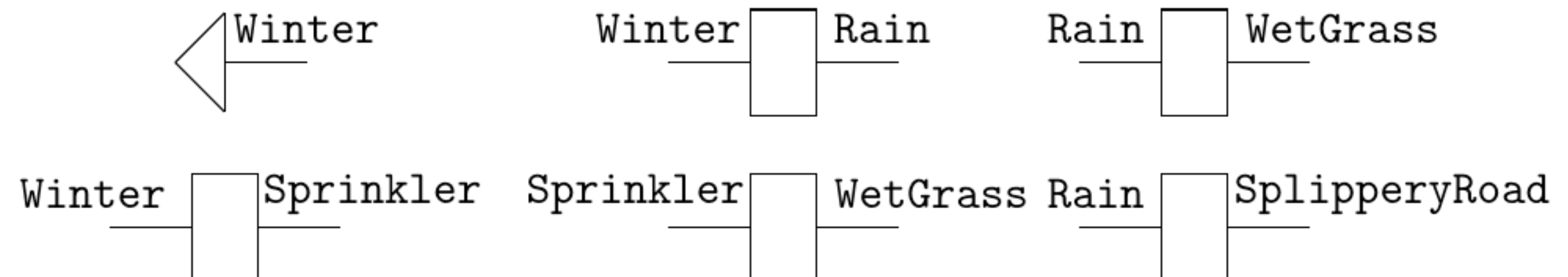
$$\Sigma_{\mathbb{L}} := \left\{ \begin{array}{c} B_1 \\ \vdots \\ B_m \end{array} \begin{array}{|c} \hline \varphi \\ \hline \end{array} \begin{array}{c} A \end{array} \mid \varphi \equiv A \leftarrow B_1, \dots, B_m \text{ is a clause in } \mathbb{L} \right\}$$

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- ▶ Morphisms: freely composed string diagrams using  $\Sigma_{\mathbb{L}}$  and CDMU structure, modulo CDMU equations.

- ▶ **Example.**  $\Sigma_{\mathbb{L}_{wet}}$  consists of 6 boxes:



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## Semantics category

▶  $Kl(\mathcal{D})(\mathbf{2})$

- ▶ objects: finite products of  $\mathbf{2}_A = \{0_A, 1_A\}$ .
- ▶ morphisms  $X \rightarrow Y$  are functions  $X \rightarrow \mathcal{D}(Y)$ .

▶  $Kl(\mathcal{D})(\mathbf{2})$  is a CDMU category:

$$\begin{array}{cccc} [[\neg \langle_A]]_{\mathbb{P}} : \mathbf{2}_A \rightarrow \mathbf{2}_A \times \mathbf{2}_A & [[\neg \bullet_A]]_{\mathbb{P}} : \mathbf{2}_A \rightarrow \mathbf{1} & [[\neg \bullet_A]]_{\mathbb{P}} : \mathbf{2}_A \times \mathbf{2}_A \rightarrow \mathbf{2}_A & [[\bullet \neg_A]]_{\mathbb{P}} : \mathbf{1} \rightarrow \mathbf{2}_A \\ x \mapsto 1|(x, x)\rangle & x \mapsto 1|*\rangle & (x, y) \mapsto 1|x \vee y\rangle & * \mapsto 1|\neg A\rangle \end{array}$$

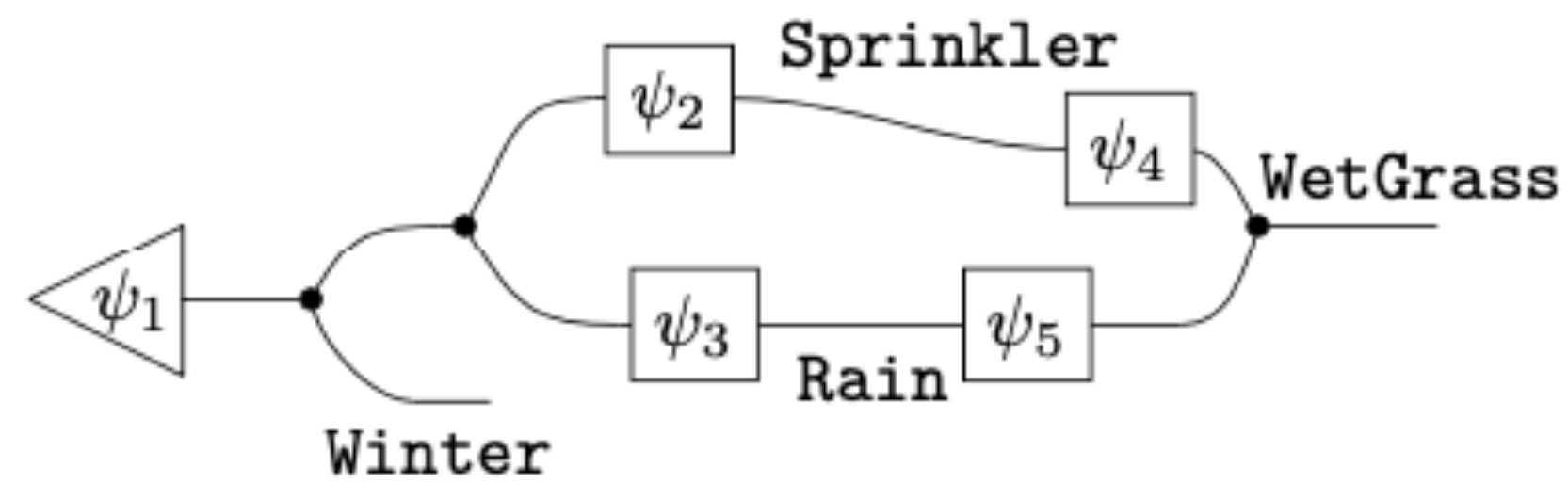
# Functorial semantics of PLP

- ▶ **Proposition.** PLP programs based on  $\mathbb{L} \simeq$  structure-preserving functors  $[[ - ]]: \text{SynPLP}_{\mathbb{L}} \rightarrow \text{Kl}(\mathcal{D})(\mathbf{2})$  mapping generating objects to generating objects.
- ▶ Idea:  $\mathbb{P}$  determines the functor  $[[ - ]]$ 's action on the generating morphisms/boxes, namely the probability labels of the clauses.
- ▶ **Example.**  $\mathbb{P}_{wet}$  defines the functor  $[[ - ]]$  such that:

$$\begin{array}{ccc} \text{Rain} \text{---} \boxed{\phantom{0}} \text{---} \text{WetGrass} & \xrightarrow{[[ - ]]_{\mathbb{P}_{wet}}} & \begin{array}{l} 0_{\text{Rain}} \mapsto 1 | 0_{\text{WetGrass}} \rangle \\ 1_{\text{Rain}} \mapsto 0.8 | 1_{\text{WetGrass}} \rangle + 0.2 | 0_{\text{WetGrass}} \rangle \end{array} \end{array}$$

# Atom probability

- ▶ A compositional diagrammatic presentation of atom probabilities.
- ▶ **Example.** the  $[[ - ]]$  <sub>$\mathbb{P}_{wet}$</sub> -image of the following string diagram is  $Pr(\text{Winter}, \text{WetGrass})$  in  $\mathcal{D}(2_{\text{Winter}} \times 2_{\text{WetGrass}})$ .



**Application: transformations  
between PLP and BN**



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# Functorial semantics of BN

▶ Idea:

▶ **Syntax:** DAGs  $\rightarrow$  string diagrams.  $\text{SynBN}_G$  for a DAG  $G$  is  $\text{FreeCD}(V_G, \Sigma_G)$ ,

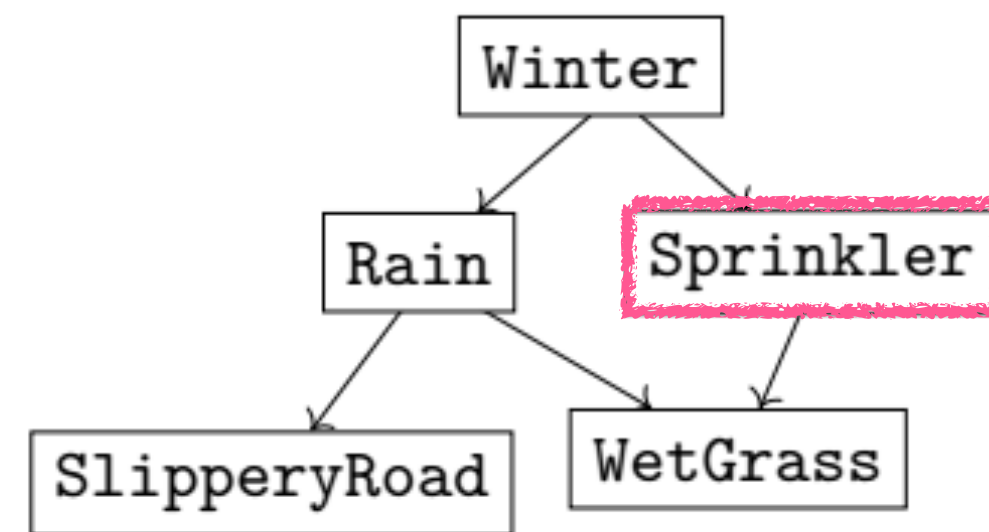
where  $\Sigma_G = \left\{ \begin{array}{c} B_1 \text{---} \\ \vdots \\ B_m \text{---} \end{array} \boxed{\phantom{A}} \text{---} A \mid pa(A) = \{B_1, \dots, B_m\} \right\}$

▶ **Semantics:** category of stochastic processes.

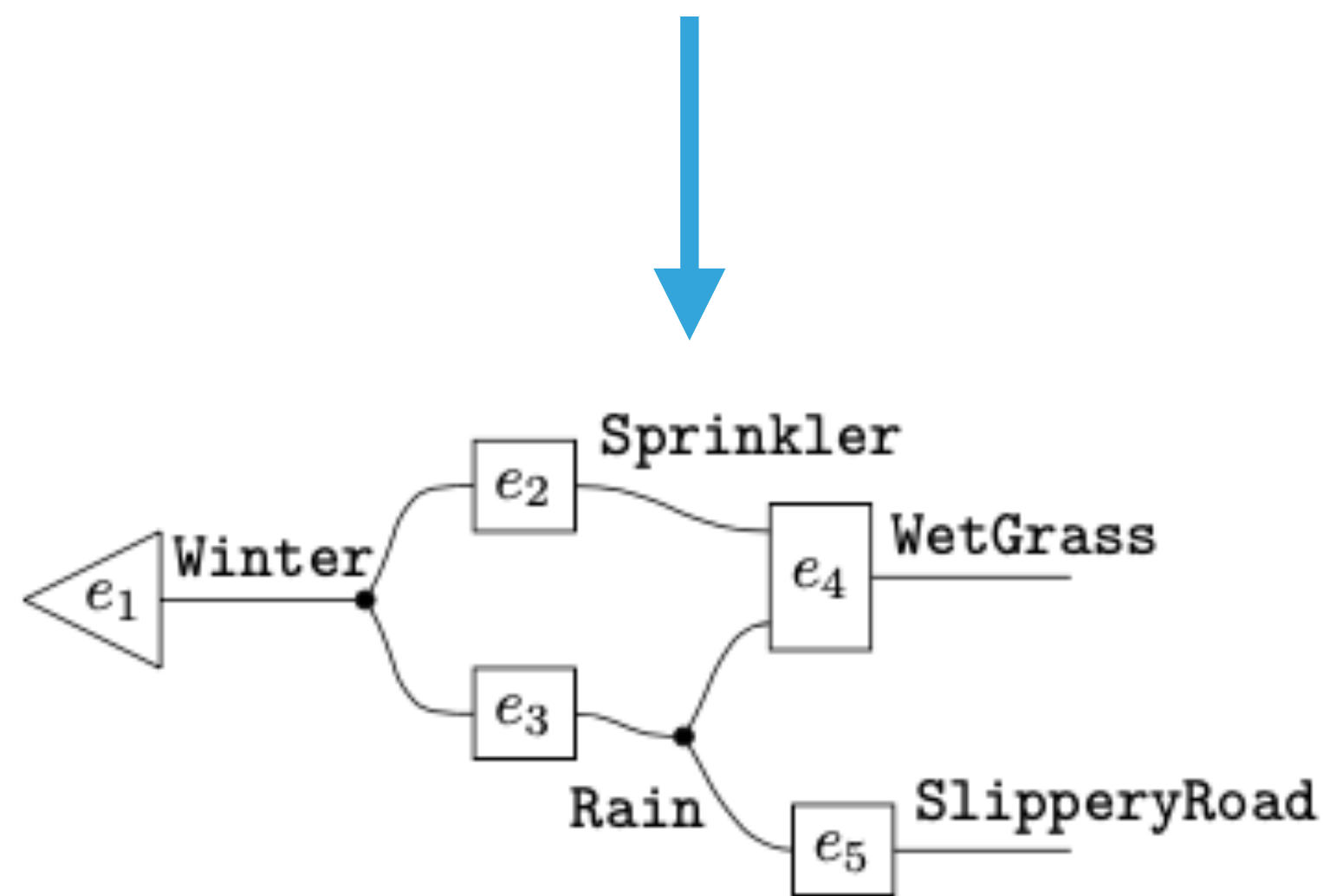
▶ **Proposition ([JKZ2019]).** Boolean-valued Bayesian networks based on  $G \simeq$  structure-preserving functors  $[[ - ]]_{\mathbb{B}} : \text{SynBN}_G \rightarrow \text{Kl}(\mathcal{D})(\mathbf{2})$ .

# BN example

► BN  $\mathbb{B}_{wet}$ :



		Pr(WetGrass)
$\neg$ Sprinkler	$\neg$ Rain	0
$\neg$ Sprinkler	Rain	0.8
Sprinkler	$\neg$ Rain	0.9
Sprinkler	Rain	0.98



$G_{wet}$  as string diagram

$2_{Sprinkler} \times 2_{Rain} \rightarrow \mathcal{D}(2_{WetGrass})$  such that:

$$(0_S, 0_R) \mapsto 1 |0_W\rangle$$

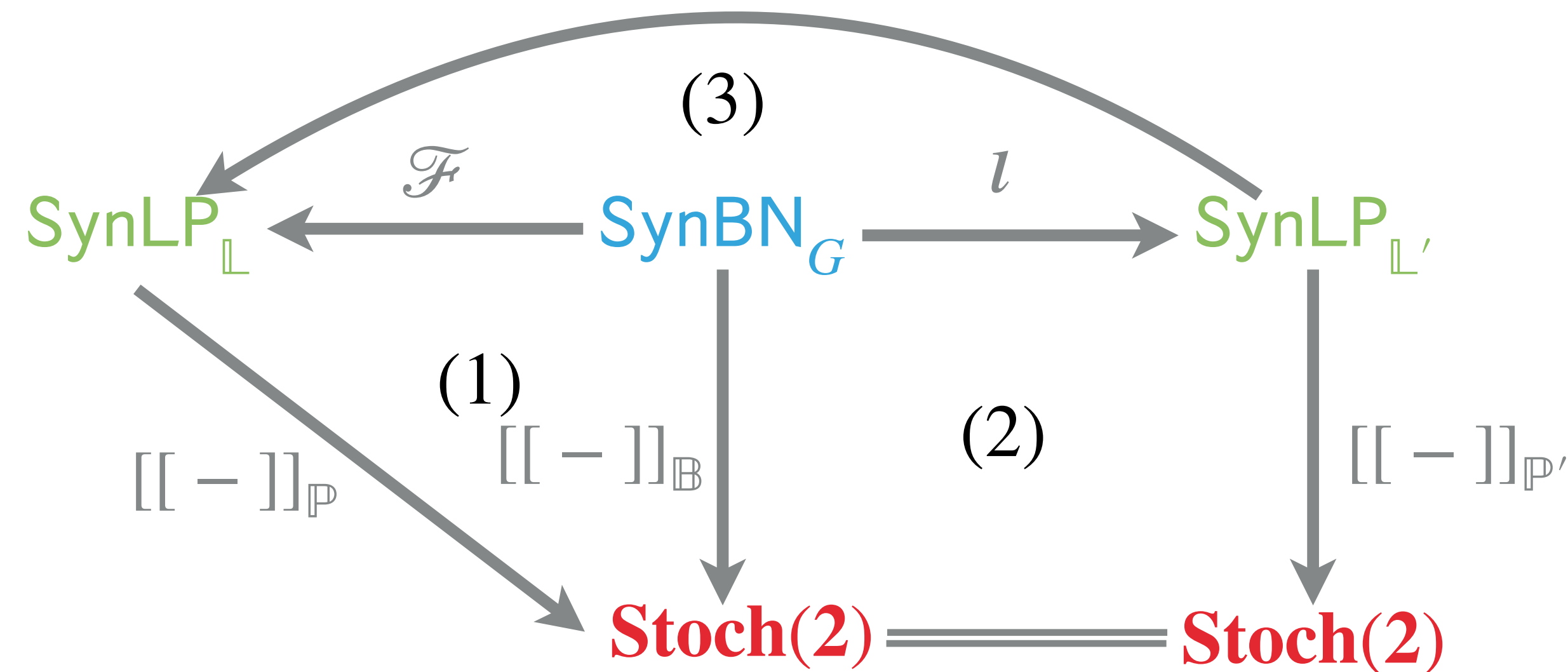
$$(0_S, 1_R) \mapsto 0.8 |0_W\rangle + 0.2 |0_W\rangle$$

$$(1_S, 0_R) \mapsto 0.9 |0_W\rangle + 0.1 |0_W\rangle$$

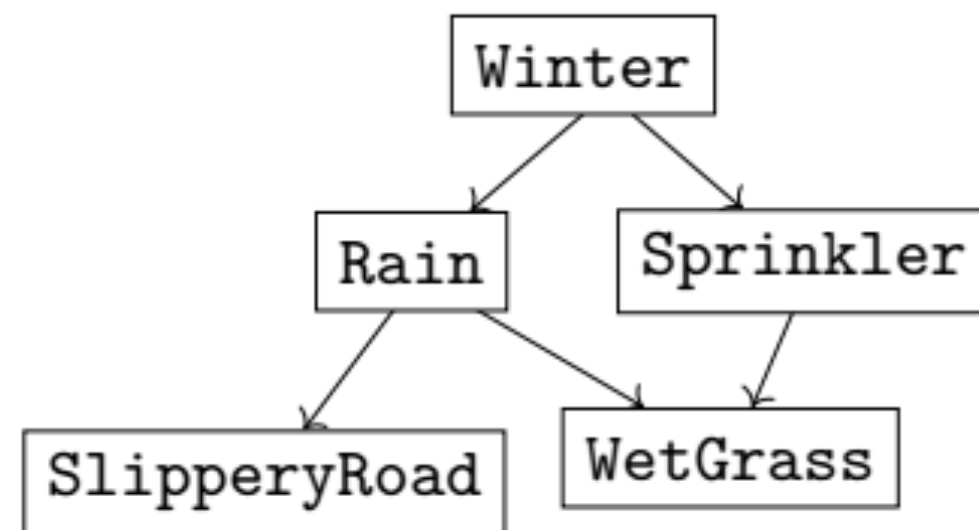
$$(1_S, 1_R) \mapsto 0.98 |0_W\rangle + 0.02 |0_W\rangle$$

## PLP and BN

- ▶ **Fact.** Every boolean-valued Bayesian network can be encoded as an acyclic ground PLP, and vice versa.
- ▶ Categorically,



# From BN to PLP

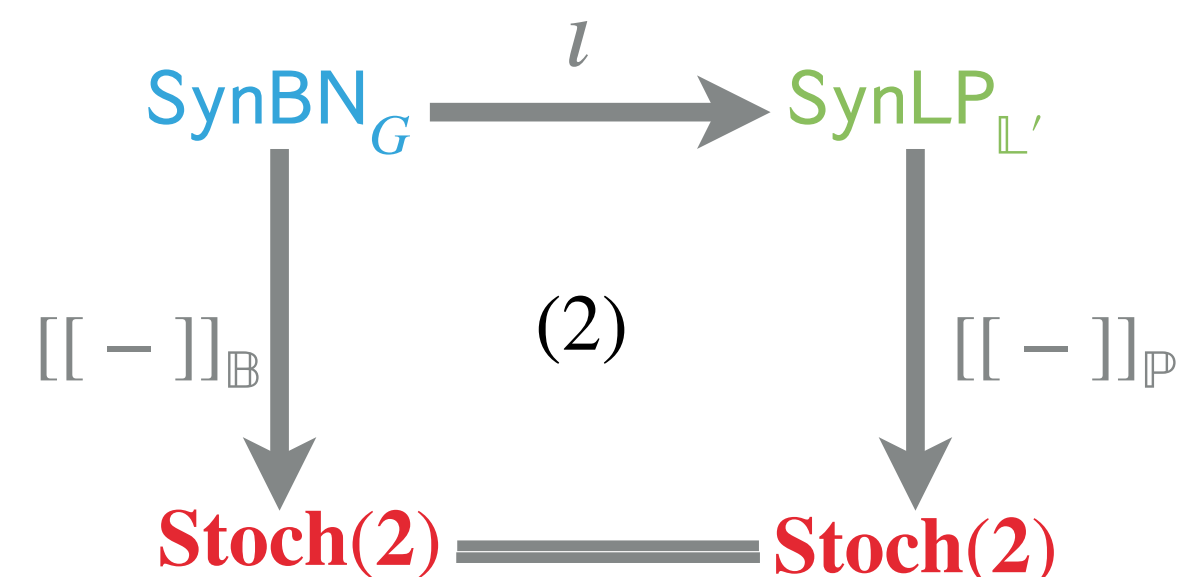


		Pr(WetGrass)
$\neg$ Sprinkler	$\neg$ Rain	0
$\neg$ Sprinkler	Rain	0.8
Sprinkler	$\neg$ Rain	0.9
Sprinkler	Rain	0.98



$0:: \text{WetGrass} \leftarrow \neg\text{Sprinkler}, \neg\text{Rain} .$   
 $0.8:: \text{WetGrass} \leftarrow \neg\text{Sprinkler}, \text{Rain} .$   
 $0.9:: \text{WetGrass} \leftarrow \text{Sprinkler}, \neg\text{Rain} .$   
 $0.98:: \text{WetGrass} \leftarrow \text{Sprinkler}, \text{Rain} .$

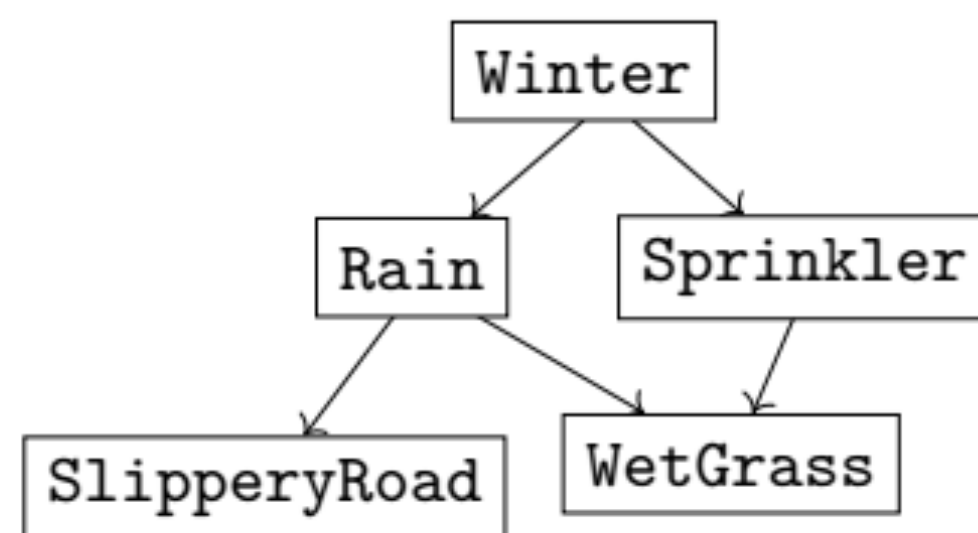
- ▶ Start from  $\text{SynBN}_G = \text{freeCD}(V_G, \Sigma_G)$ , define  $\text{SynPLP}_{\mathbb{L}}$  as  $\text{freeCDMU}(V_G, \Sigma_G)$ .
- ▶  $[[ - ]]$ <sub>P</sub> is  $[[ - ]]$ <sub>B</sub> plus (routine) interpretation of multiplication and unit.



# From PLP to BN

0.25 :: Winter ← . ( $\psi_1$ )  
0.2 :: Sprinkler ← Winter. ( $\psi_2$ )  
0.6 :: Rain ← Winter. ( $\psi_3$ )  
0.1 :: Rain ←  $\neg$ Winter. ( $\psi_4$ )

0.9 :: WetGrass ← Sprinkler. ( $\psi_5$ )  
0.8 :: WetGrass ← Rain. ( $\psi_6$ )  
0.7 :: SlipperyRoad ← Rain. ( $\psi_7$ )  
0.1 :: SlipperyRoad ←  $\neg$ Rain. ( $\psi_8$ )



		Pr(WetGrass)
$\neg$ Sprinkler	$\neg$ Rain	0
$\neg$ Sprinkler	Rain	0.8
Sprinkler	$\neg$ Rain	0.9
Sprinkler	Rain	0.98

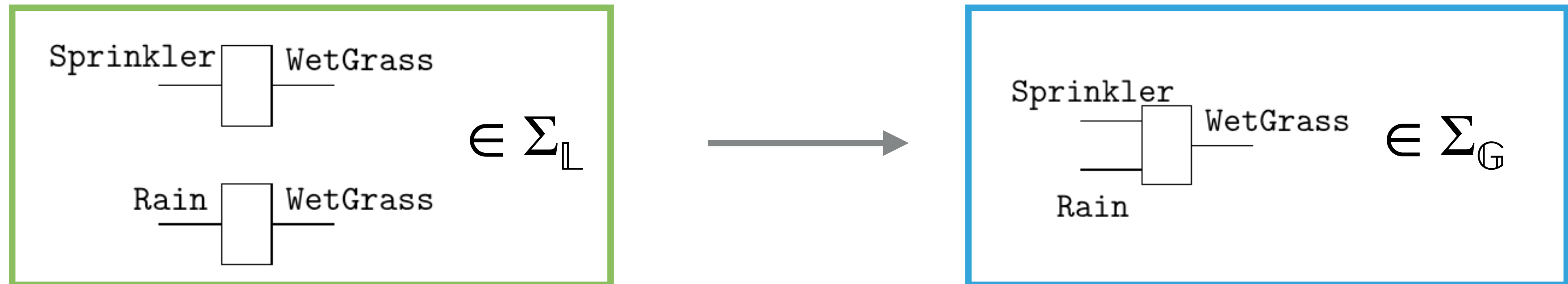
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## From PLP to BN

- ▶ **Observation.** The PLP-to-BN transformation is syntactical.
- ▶ In words, to obtain the corresponding BN model (functor) of a PLP model (functor), it suffices to define a suitable functor between the two syntax categories.

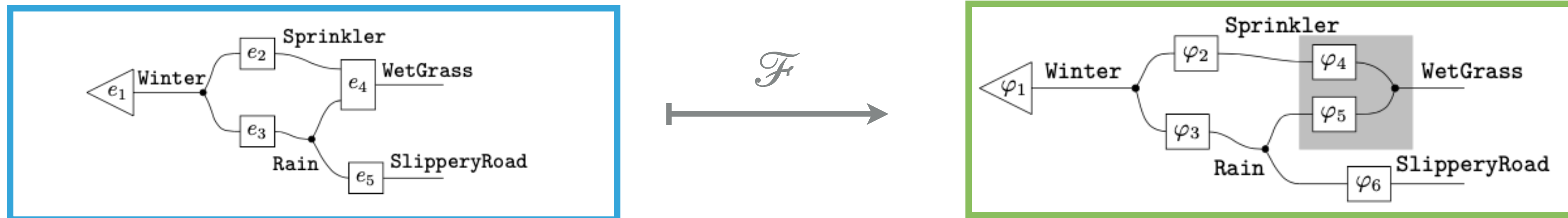
# Step 1

- ▶ We define a BN-syntax category from the PLP-syntax category, by combining all generating boxes in the latter with the same codomain into a single generating box in the former.
- ▶ **Example.**

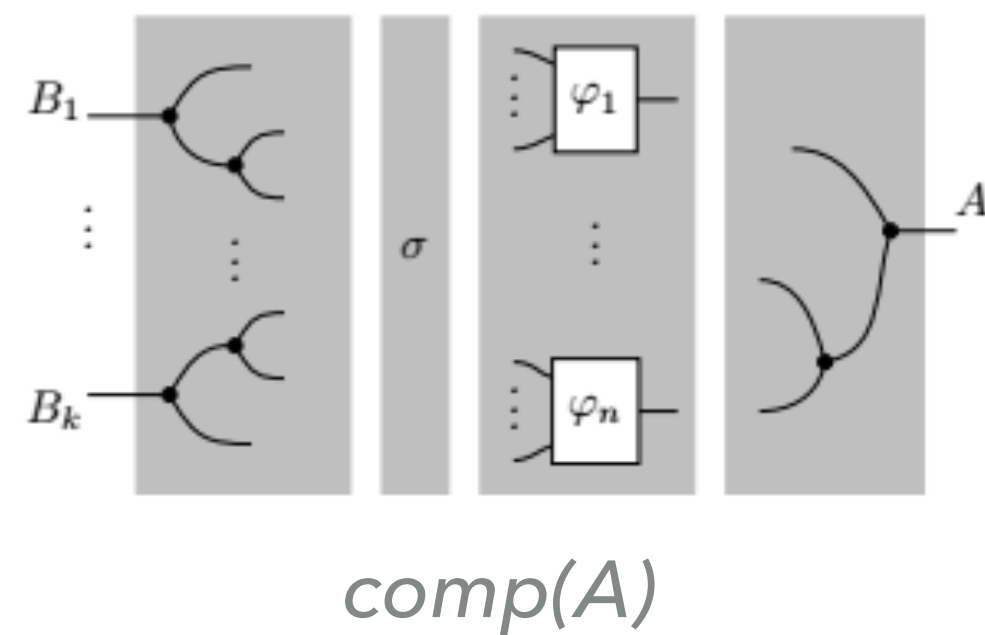


## Step 2

- ▶ We inductively define a functor  $\mathcal{F}$  between the two syntax categories.
- ▶ Example.



- ▶ In general,  $pa(A) \rightarrow A$  is mapped to  $comp(A)$ :





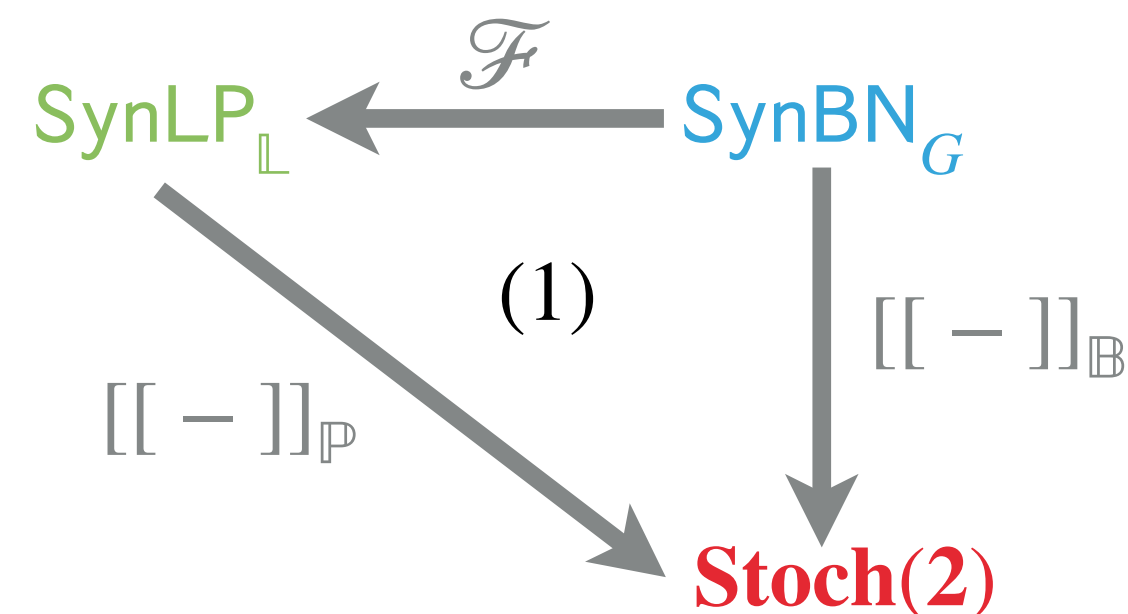
# Step 3

▶ We obtain the BN model by composition:  $[[ - ]]_{\mathbb{B}} = [[ - ]]_{\mathbb{P}} \circ \mathcal{F}$

▶ **Example.** 
$$\left[ \begin{array}{c} \text{Sprinkler} \\ \text{Rain} \end{array} \begin{array}{c} \text{WetGrass} \end{array} \right]_{\mathbb{B}} = \left[ \mathcal{F} \left( \begin{array}{c} \text{Sprinkler} \\ \text{Rain} \end{array} \begin{array}{c} \text{WetGrass} \end{array} \right) \right]_{\mathbb{P}} = \left[ \begin{array}{c} \text{Sprinkler} \\ \text{Rain} \end{array} \begin{array}{c} \text{WetGrass} \end{array} \right]_{\mathbb{P}}$$

$$= \left( \left[ \begin{array}{c} \text{Sprinkler} \\ \text{WetGrass} \end{array} \right]_{\mathbb{P}} \times \left[ \begin{array}{c} \text{Rain} \\ \text{WetGrass} \end{array} \right]_{\mathbb{P}} \right); \left[ \begin{array}{c} \text{WetGrass} \end{array} \right]_{\mathbb{P}}$$

▶ For instance,  $(0_{\text{Sprinkler}}, 1_{\text{Rain}}) \mapsto 0.8 | 1_{\text{WetGrass}} \rangle + 0.2 | 0_{\text{WetGrass}} \rangle$



# Classical, weighted LP

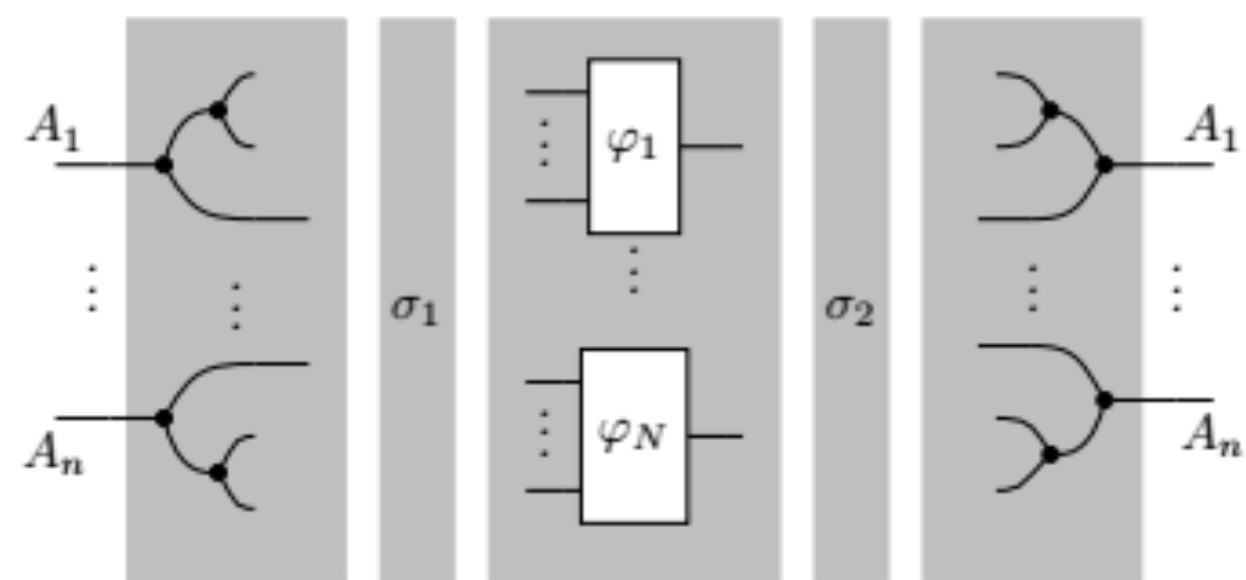
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# Idea

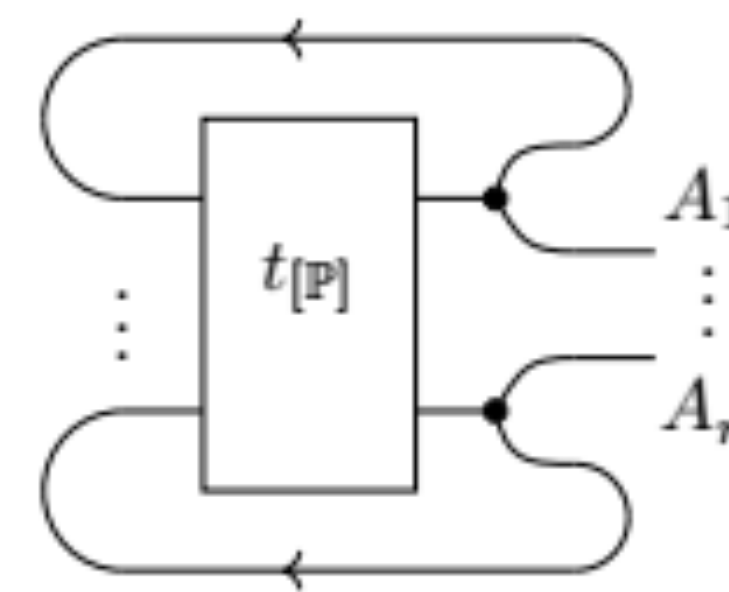
- ▶ Syntax categories: remains the same.
- ▶ Semantics categories:
  - ▶ Classical LP: **Set(2)**
  - ▶ Probabilistic LP:  $Kl(\mathcal{D})(\mathbf{2})$
  - ▶ Weighted LP: **Set( $\mathcal{K}$ )**
- ▶ Interestingly,  $Kl(\mathcal{D})(\mathbf{2})$  and **Set( $\mathcal{K}$ )** are both variants of **Set(2)**, but in different flavour:  $Kl(\mathcal{D})(\mathbf{2})$  is some 'Kleisli category' (morphisms are different), **Set( $\mathcal{K}$ )** changes the 'base' (objects are different).

# Fixed-point style semantics

- ▶ We enrich the syntax categories to include 'feedback wires'.
- ▶ Consequently the semantics categories move from functions to relations.
- ▶ Example.



immediate consequence operator  $T_P$



supported models

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## Future work

- ▶ General case: variables may occur.
- ▶ Other logic-programming formalisms, e.g. CP-logic, LPAD.
- ▶ Diagrammatically represent PLP inference tasks, e.g. maximum a posteriori (MAP), most likely explanation (MEP), probabilistic inductive logic programming (PILP).

**Thank you!**

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## Reference

- ▶ F. W. Lawvere, *Functorial Semantics of Algebraic Theories*, Ph.D. thesis, Columbia University, 1963.
- ▶ Brendon Fong. Causal theories: A categorical perspective of on bayesian networks. 2013. arXiv:1301.6201.
- ▶ Bart Jacobs, Aleks Kissinger, and Fabio Zanasi. Causal inference by string diagram surgery. CoRR, abs/1811.08338, 2018.