

# Modal Characterization Theorems: from the classical to the quantitative coalgebraic

---

Paul Wild

October 25, 2022

Friedrich-Alexander-Universität Erlangen-Nürnberg

# A van Benthem theorem for fuzzy modal logic (LICS'18)

Paul Wild, Lutz Schröder, Dirk Pattinson and Barbara König

# The van Benthem Theorem

## Theorem

*Every bisimulation invariant first-order formula can be expressed by modal formulas.*

Goal: formalize and prove this theorem in terms of fuzzy logic.

# Fuzzy Modal Logic (1)

## Theorem

*Every bisimulation invariant first-order formula can be expressed by modal formulas*

Syntax:

$$\varphi, \psi ::= c \mid p \mid \neg\varphi \mid \varphi \ominus c \mid \varphi \wedge \psi \mid \Diamond\varphi$$

where  $c \in \mathbb{Q} \cap [0, 1]$  and  $p \in \text{At}$ .

Semantics: given over *fuzzy relational models*

$$\mathcal{A} = (A, (p^{\mathcal{A}})_{p \in \text{At}}, R^{\mathcal{A}})$$

- $A$  is the set of states.
- $p^{\mathcal{A}} : A \rightarrow [0, 1]$  is the interpretation for  $p \in \text{At}$ .
- $R^{\mathcal{A}} : A \times A \rightarrow [0, 1]$  is the transition relation.

## Fuzzy Modal Logic (2)

Formulas are interpreted as functions  $A \rightarrow [0, 1]$ :

- constants:  $c(a) = c$
- propositions:  $p(a) = p^A(a)$
- negation:  $(\neg\varphi)(a) = 1 - \varphi(a)$
- truncated subtraction:  $(\varphi \ominus c)(a) = \max(\varphi(a) - c, 0)$
- conjunction:  $(\varphi \wedge \psi)(a) = \min(\varphi(a), \psi(a))$
- modality:  $(\diamond\varphi)(a) = \sup_{a' \in A} \min(R^A(a, a'), \varphi(a'))$

Notations:  $a \wedge b = \min(a, b)$ ,  $a \vee b = \max(a, b)$ ,  $\bigvee = \sup$ ,  $\bigwedge = \inf$ .

## Theorem

*Every bisimulation invariant first-order formula can be expressed by modal formulas.*

Syntax:

$$\varphi, \psi ::= c \mid p(x) \mid R(x, y) \mid x = y \mid \neg\varphi \mid \varphi \ominus c \mid \varphi \wedge \psi \mid \exists x.\varphi$$

where  $c \in \mathbb{Q} \cap [0, 1]$ ,  $p \in \text{At}$ ,  $x, y$  variables.

Semantics:

Let  $\eta: \text{Var} \rightarrow A$ .  $\varphi(\eta)$  is defined inductively:

- Boolean connectives and equality as expected
- $p(x)(\eta) = p^A(\eta(x))$ ,  $R(x, y)(\eta) = R^A(\eta(x), \eta(y))$
- existential quantification:  $(\exists x.\varphi)(\eta) = \bigvee_{a \in A} \varphi(\eta[x \mapsto a])$

# Bisimulation Invariance

## Theorem

Every *bisimulation invariant first-order formula* can be expressed by modal formulas.



$$d(a, b) = 0.1$$

- In fuzzy logic we can *quantify* how similarly two states behave.
- This gives rise to *behavioural distance*  $d$ .
- Bisimilar states have distance 0.
- $\varphi$  *bisimulation invariant*  $\iff \varphi$  non-expansive wrt.  $d$ :

$$|\varphi(a) - \varphi(b)| \leq d(a, b) \quad \text{for all states } a, b.$$

## Theorem

*Every bisimulation invariant first-order formula can be expressed by modal formulas.*

- In classical modal logic, there are only finitely many modal formulas of fixed rank  $k$  (up to equivalence).
- In fuzzy modal logic, this is no longer true, because there are infinitely many truth constants  $c \in \mathbb{Q} \cap [0, 1]$ .
- Thus, instead of showing that the bisimulation invariant formula  $\varphi$  is *equivalent* to some modal  $\varphi$  of rank  $k$ , we show that it can be *approximated* by such formulas:

$$\forall \varepsilon > 0 \quad \exists \psi_\varepsilon \text{ modal of rank } k \quad \|\varphi - \psi_\varepsilon\|_\infty \leq \varepsilon.$$



# A Fuzzy van Benthem Theorem

## Theorem

*Every fuzzy first-order formula  $\varphi$  that is non-expansive wrt. behavioural distance  $d^G$  can be approximated by fuzzy modal formulas of some fixed rank  $k$ .*

Next: define behavioural distance  $d^G$  via a bisimulation game.

## Game-based Distance (1)

*Bisimulation game* for fuzzy logic:

- The game is parametrised by some  $\varepsilon \geq 0$
- Two players, spoiler  $S$  and duplicator  $D$
- Configurations: pairs of states  $(a, b)$
- Moves:
  - $S$  picks  $a'$  such that  $R(a, a') > \varepsilon$
  - $D$  picks  $b'$  such that  $R(b, b') \geq R(a, a') - \varepsilon$
  - New configuration:  $(a', b')$

$S$  may also swap the two sides before his move

- Whoever is unable to move, loses
- Winning condition for  $D$  before every round:

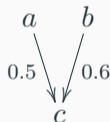
$$|p(a) - p(b)| \leq \varepsilon \text{ for all } p \in \text{At.}$$

## Game-based Distance (2)

The corresponding distances are:

$$d^G(a, b) = \bigwedge \{ \varepsilon \mid D \text{ wins the } \varepsilon\text{-game for } (a, b) \}$$

$$d_n^G(a, b) = \bigwedge \{ \varepsilon \mid D \text{ wins the } n\text{-round } \varepsilon\text{-game for } (a, b) \}$$



$$d^G(a, b) = 1$$

$D$  wins for  $\varepsilon = 0.1$ , but loses for  $\varepsilon < 0.1$ .

# Logic-based Distance

Using modal formulas, we can define:

$$d^L(a, b) = \bigvee_{\varphi \text{ modal}} |\varphi(a) - \varphi(b)|$$

$$d_n^L(a, b) = \bigvee_{\varphi \text{ modal, rk}\varphi \leq n} |\varphi(a) - \varphi(b)|$$



$$d_1^L(a, b) = 0.1 \text{ with } \varphi = \diamond 1$$

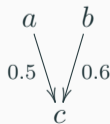
## Function-based Distance

Behavioural distance via a *Kantorovich* construction:

$$d_0^K(a, b) = 0$$

$$d_{n+1}^K(a, b) = \bigvee_{p \in \text{At}} |p(a) - p(b)| \vee \bigvee_{f: (A, d_n^K) \rightarrow [0,1] \text{ nonexp.}} |(\diamond f)(a) - (\diamond f)(b)|$$

$$(\diamond f)(a) = \bigvee_{a' \in A} R(a, a') \wedge f(a')$$



$$d_1^K(a, b) = 0.1 \text{ with } f = x \mapsto 1.$$

## Theorem

*Let  $\mathcal{A}$  be a model and  $n \geq 0$ . Then*

- 1.  $d_n^G = d_n^K = d_n^L =: d_n$  on  $\mathcal{A}$ .*
- 2.  $(A, d_n)$  is a totally bounded pseudometric space.*
- 3. The rank  $n$  formulas are a dense subset of the space of non-expansive maps  $(A, d_n) \rightarrow [0, 1]$ .*

## Coalgebraic View

Consider the set functors  $F$  and  $G$ :

$$FX = [0, 1]^X, \quad Ff(g)(y) = \bigvee_{f(x)=y} g(x)$$

where  $f: X \rightarrow Y, g \in [0, 1]^X, y \in Y$ .

$$GX = [0, 1]^{\text{At}} \times FX$$

Models  $\mathcal{A} = (A, (p^{\mathcal{A}})_{p \in \text{At}}, R^{\mathcal{A}})$  are *coalgebras*  $\alpha: A \rightarrow GA$ :

$$\alpha(a) = (\lambda p. p^{\mathcal{A}}(a), \lambda a'. R^{\mathcal{A}}(a, a')).$$

## Uniform Approximation

- $F_n := G^n(\{*\})$  is the set of all  $n$ -step behaviours.
- We can construct a model  $\mathcal{F}$  on the set  $F := \bigcup_{n \geq 0} F_n$ :

$$p^{\mathcal{F}}(h, g) = h(p), \quad R^{\mathcal{F}}((h, g), y) = \begin{cases} g(y), & \text{if } y \in F_n, \\ 0, & \text{otherwise.} \end{cases}$$

- For every model  $\mathcal{A}$ , there is a map  $\pi_n: A \rightarrow F$  such that

$$d_n(a, \pi_n(a)) = 0.$$

- Thus:

$$\|\varphi - \psi\|_{\infty} \leq \varepsilon \text{ on } \mathcal{F} \implies \|\varphi - \psi\|_{\infty} \leq \varepsilon \text{ on all models } \mathcal{A}.$$



# A characterization theorem for a modal description logic (IJCAI'19)

Paul Wild, Lutz Schröder, Dirk Pattinson and Barbara König

## The logic $\mathcal{ALC}(\mathbf{P})$ - Syntax

*Quantitative Probabilistic ALC:*

$$C, D ::= q \mid A \mid C \ominus q \mid \neg C \mid C \sqcap D \mid \mathbf{P} r. C$$

- rational constants  $q \in \mathbb{Q} \cap [0, 1]$
- basic concept names  $A \in \mathbf{N}_C$
- subtraction of constants  $\ominus$
- expected value over  $r$ -successors  $\mathbf{P}$  ( $r \in \mathbf{N}_R$ )

Loud  $\sqcap$   $\mathbf{P}$  hasSource. (Large  $\sqcap$   $\mathbf{P}$  hasMood. Angry)

## The logic $\mathcal{ALC}(\mathbf{P})$ - Semantics

Models:  $\mathcal{I} = (\Delta^{\mathcal{I}}, (A^{\mathcal{I}})_{A \in \mathbf{N}_C}, (r^{\mathcal{I}})_{r \in \mathbf{N}_R})$ , where

- $\Delta^{\mathcal{I}}$  is a set (the *domain*)
- $A^{\mathcal{I}}: \Delta^{\mathcal{I}} \rightarrow [0, 1]$
- $r^{\mathcal{I}}: \Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}} \rightarrow [0, 1]$

such that  $\sum_{a' \in \Delta^{\mathcal{I}}} r^{\mathcal{I}}(a, a') \in \{0, 1\}$  for each  $a \in \Delta^{\mathcal{I}}$ .

In other words, for role  $r$  each state  $a$  is either

- *r-blocking* –  $r_a := r^{\mathcal{I}}(a, \cdot)$  is zero; or
- *r-transient* –  $r_a$  is a discrete probability distribution on  $\Delta^{\mathcal{I}}$ .

## The logic $\mathcal{ALC}(\mathbf{P})$ - Semantics

Interpretations:  $C^{\mathcal{I}}: \Delta^{\mathcal{I}} \rightarrow [0, 1]$ , where

$$q^{\mathcal{I}}(a) = q$$

$$(C \ominus q)^{\mathcal{I}}(a) = \max(C^{\mathcal{I}}(a) - q, 0)$$

$$(\neg C)^{\mathcal{I}}(a) = 1 - C^{\mathcal{I}}(a)$$

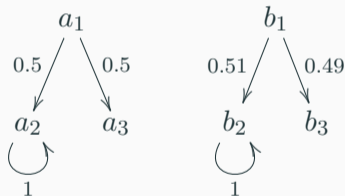
$$(C \sqcap D)^{\mathcal{I}}(a) = \min(C^{\mathcal{I}}(a), D^{\mathcal{I}}(a))$$

$$(\mathbf{P} r. C)^{\mathcal{I}}(a) = \mathbb{E}_{r_a}(C^{\mathcal{I}}) = \sum_{a' \in \Delta^{\mathcal{I}}} r_a(a') \cdot C^{\mathcal{I}}(a')$$

## Bisimulation Invariance

From now on, restrict to a single role  $\pi$ .

- Classically, bisimulations are used to tell whether two states exhibit the same behaviour.
- However, consider the following states:



$$d(a_1, b_1) = 0.01$$

- With a *behavioural distance*  $d$  we can *quantify* how similarly two states behave. Bisimilar states have distance 0.

## Towards a Characterization Theorem

- Defer the precise definition of *bisimulation distance*  $d$  for now.
- $\varphi$  *bisimulation invariant*  $\iff \varphi$  non-expansive wrt.  $d$ :

$$|\varphi(a) - \varphi(b)| \leq d(a, b) \quad \text{for all states } a, b.$$

- Characterize  $\mathcal{ALC}(\mathbf{P})$  using bisimulation invariance:
  - All  $\mathcal{ALC}(\mathbf{P})$ -concepts are bisimulation invariant.
  - Every bisimulation invariant property can be approximated by  $\mathcal{ALC}(\mathbf{P})$ -concepts.

# Correspondence Language

*Quantitative probabilistic first-order logic* (FO(**P**)):

$$\varphi, \psi ::= q \mid A(x) \mid x = y \mid \varphi \ominus q \mid \neg\varphi \mid \varphi \wedge \psi \mid \exists x. \varphi \mid x\mathbf{P}[y : \varphi]$$

Semantics:

$$A(x_i)(\bar{a}) = A^{\mathcal{I}}(a_i)$$

$$(\exists x_0. \varphi(x_0, x_1, \dots, x_n))(\bar{a}) = \sup\{\varphi(a_0, \bar{a}) \mid a_0 \in \Delta^{\mathcal{I}}\}$$

$$(x_i\mathbf{P}[y : \varphi(y, x_1, \dots, x_n)])(\bar{a}) = \mathbb{E}_{r_{a_i}}(\varphi(\cdot, \bar{a}))$$

Example:

$$\begin{aligned} x\mathbf{P}[z : z = y] &= \text{'the successor of } x \text{ is probably } y\text{'} \\ &= \text{probability of reaching } y \text{ from } x \text{ in one step} \end{aligned}$$

## Modal Approximation

- In classical  $\mathcal{ALC}$ , there are only finitely many modal formulas of fixed rank  $k$  (up to equivalence).
- In fuzzy modal logic, this is no longer true, because there are infinitely many truth constants  $c \in \mathbb{Q} \cap [0, 1]$ .
- Thus, instead of showing that the bisimulation invariant formula  $\varphi$  is *equivalent* to some modal  $\psi$  of rank  $k$ , we show that it can be *approximated* by such formulas:

$$\forall \varepsilon > 0 \quad \exists \psi_\varepsilon \text{ modal of rank } k \quad \|\varphi - \psi_\varepsilon\|_\infty \leq \varepsilon.$$



## Bisimulation game

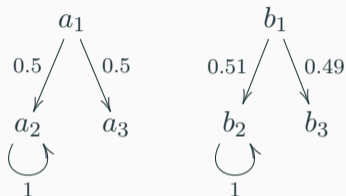
Game on models  $\mathcal{I}, \mathcal{J}$  played by *Spoiler* ( $S$ ) and *Duplicator* ( $D$ ):

- *Configurations*: triples  $(a, b, \varepsilon)$ ,  $a \in \Delta^{\mathcal{I}}$ ,  $b \in \Delta^{\mathcal{J}}$ ,  $\varepsilon \in [0, 1]$ .
- *Moves*:
  - $D$  picks  $\mu \in \text{Cpl}(\pi_a, \pi_b)$
  - $D$  picks a function  $\varepsilon': \Delta^{\mathcal{I}} \times \Delta^{\mathcal{J}} \rightarrow [0, 1]$  such that  $\mathbb{E}_{\mu}(\varepsilon') \leq \varepsilon$
  - $S$  picks  $(a', b')$  with  $\mu(a', b') > 0$
  - New configuration:  $(a', b', \varepsilon'(a', b'))$
- $D$  wins if both states are blocking or  $\varepsilon = 1$ .
- $S$  wins if exactly one state is blocking and  $\varepsilon < 1$ .
- Otherwise,  $D$  wins if she maintains the *winning condition*:  $|A^{\mathcal{I}}(a) - A^{\mathcal{J}}(b)| \leq \varepsilon$  for all  $A \in \mathbf{N}_{\mathcal{C}}$ .

$\text{Cpl}(\pi_a, \pi_b)$ : set of  $\mu: \Delta^{\mathcal{I}} \times \Delta^{\mathcal{J}} \rightarrow [0, 1]$  with marginals  $\pi_a$  and  $\pi_b$ :

$$\pi_a(a') = \sum_{b'} \mu(a', b') \qquad \pi_b(b') = \sum_{a'} \mu(a', b')$$

## Example game



- Initial configuration:  $(a_1, b_1, 0.01)$ .
- First turn:  $D$  picks  $\mu$  and  $\varepsilon'$  as follows:

$\mu$	$b_2$	$b_3$	
$a_2$	0.5	0	0.5
$a_3$	0.01	0.49	0.5
	0.51	0.49	

$\varepsilon'$	$b_2$	$b_3$
$a_2$	0	1
$a_3$	1	0

$$d^G(a, b) = \inf\{\varepsilon \mid D \text{ wins the game for } (a, b, \varepsilon)\}$$

$$d_n^G(a, b) = \inf\{\varepsilon \mid D \text{ wins the } n\text{-round game for } (a, b, \varepsilon)\}$$

### Lemma

*Each  $\mathcal{ALC}(\mathbf{P})$ -concept of rank  $n$  is depth- $n$  bisimulation-invariant, that is*

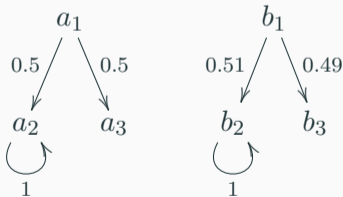
$$|C(a) - C(b)| \leq d_n^G(a, b).$$

# Logical Distance

Using modal formulas, we can define:

$$d^L(a, b) = \sup\{|C(a) - C(b)| \mid C \in \mathcal{ALC}(\mathbf{P})\}$$

$$d_n^L(a, b) = \sup\{|C(a) - C(b)| \mid C \in \mathcal{ALC}(\mathbf{P}), \text{rk}C \leq n\}$$



$$d_2^L(a, b) = 0.01 \text{ with } C = \mathbf{PP1}$$

## Pseudometric Liftings

Let  $(X, d)$  be a pseudometric space. We define two pseudometrics on the space  $\mathcal{D}X$  of discrete probability measures on  $X$ .

### Definition (Kantorovich distance)

$$d^\uparrow(\pi_1, \pi_2) = \sup\{|\mathbb{E}_{\pi_1}(f) - \mathbb{E}_{\pi_2}(f)| \mid f \in \text{Pred}(X, d)\}$$

where  $\text{Pred}(X, d)$  is the set of nonexpansive maps  $(X, d) \rightarrow [0, 1]$ .

### Definition (Wasserstein distance)

$$d^\downarrow(\pi_1, \pi_2) = \inf\{\mathbb{E}_\mu(d) \mid \mu \in \text{Cpl}(\pi_1, \pi_2)\}$$

These two *pseudometrics liftings* coincide:

### Theorem (Kantorovich-Rubinstein duality)

For all  $\pi_1, \pi_2$ ,  $d^\uparrow(\pi_1, \pi_2) = d^\downarrow(\pi_1, \pi_2)$ .

Behavioural distance via fixed point iteration:

$$d_0^K(a, b) = d_0^W(a, b) = 0$$

$$d_{n+1}^K(a, b) = \max\left(\sup_{A \in \mathcal{N}_c} |A^{\mathcal{I}}(a) - A^{\mathcal{I}}(b)|, (d_n^K)^{\uparrow}(\pi_a, \pi_b)\right)$$

$$d_{n+1}^W(a, b) = \max\left(\sup_{A \in \mathcal{N}_c} |A^{\mathcal{I}}(a) - A^{\mathcal{I}}(b)|, (d_n^W)^{\downarrow}(\pi_a, \pi_b)\right)$$

By Kantorovich-Rubinstein duality,  $d_n^K = d_n^W$  for all  $n$ .

# Equivalence of Distances and Density

## Theorem

Let  $\mathcal{I}$  be a model. Then for all  $n \geq 0$ :

- $d_n^G = d_n^W = d_n^K = d_n^L =: d_n$  on  $\mathcal{A}$ .
- The rank- $n$   $\mathcal{ALC}(\mathbf{P})$ -concepts form a dense subset of the space  $\text{Pred}(\Delta^{\mathcal{I}}, d_n)$  of non-expansive maps  $(\Delta^{\mathcal{I}}, d_n) \rightarrow [0, 1]$ .

This is proven by induction on  $n$ . Some intuition:

- $d_n^G = d_n^W$  because the game is built to model W. distance.
- $d_n^W = d_n^K$  by Kantorovich-Rubinstein duality.
- $d_n^K = d_n^L$  follows from the density claim for  $n - 1$ .

# The Characterization Theorem

## Theorem

*Every bisimulation-invariant FO( $\mathbf{P}$ )-formula of rank at most  $n$  can be approximated by  $\mathcal{ALC}(\mathbf{P})$ -concepts of rank at most  $3^n$ .*



# Characteristic logics for behavioural metrics via fuzzy lax extensions (CONCUR'20)

Paul Wild and Lutz Schröder

## Goal

Analyse the behaviour of transition systems involving quantitative data.

- Various system types can be modelled as **coalgebras**:
  - Labelled transition systems  $\alpha: A \rightarrow \mathcal{P}(L \times A)$
  - Markov chains  $\alpha: A \rightarrow DA$
  - ...

In general:  $\alpha: A \rightarrow TA$  for some set functor  $T$

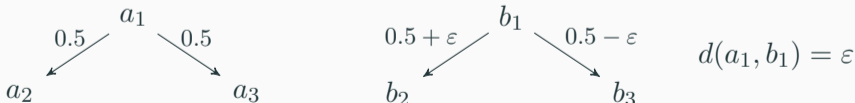
## Goal

Analyse the behaviour of transition systems involving quantitative data.

- Various system types can be modelled as **coalgebras**:
  - Labelled transition systems  $\alpha: A \rightarrow \mathcal{P}(L \times A)$
  - Markov chains  $\alpha: A \rightarrow DA$
  - ...

In general:  $\alpha: A \rightarrow TA$  for some set functor  $T$

- **Behavioural distances** allow for a quantitative measure of process equivalence:



## Goal

Analyse the behaviour of transition systems involving quantitative data.

- To define behavioural distances, we make use of **lax extensions**:
  - Lax extensions give a coalgebraic account of *bisimulation*.
  - Using a lax extension, *lift* the set functor  $T$  to a functor on pseudometrics.
  - Behavioural distance arises from a coalgebraic *fixpoint* construction.

## Goal

Analyse the behaviour of transition systems involving quantitative data.

- To define behavioural distances, we make use of **lax extensions**:
  - Lax extensions give a coalgebraic account of *bisimulation*.
  - Using a lax extension, *lift* the set functor  $T$  to a functor on pseudometrics.
  - Behavioural distance arises from a coalgebraic *fixpoint* construction.
- We extract **characteristic logics** for these behavioural distances:
  - Coalgebraic modal logics with *modalities* defined using  $L$ .
  - *Real-valued semantics* give rise to logical distance.
  - Logical distance = behavioural distance, amounting to a *Hennessy-Milner theorem*.

# Fuzzy Relations

## Definition

A **fuzzy relation** is a map  $R: A \times B \rightarrow [0, 1]$ , also written  $R: A \mapsto B$ .

Convention:  $a, b$  are related by  $R \iff R(a, b) = 0$ .

# Fuzzy Relations

## Definition

A **fuzzy relation** is a map  $R: A \times B \rightarrow [0, 1]$ , also written  $R: A \mapsto B$ .

Convention:  $a, b$  are related by  $R \iff R(a, b) = 0$ .

Let  $R: A \mapsto B, S: B \mapsto C$  and  $f: A \rightarrow B$ .

- *Converse relation*:  $R^\circ(b, a) = R(a, b)$ .
- *Graph of a function*:  $\text{Gr}_f(a, b) = \begin{cases} 0, & \text{if } f(a) = b, \\ 1, & \text{otherwise.} \end{cases}$

# Fuzzy Relations

## Definition

A **fuzzy relation** is a map  $R: A \times B \rightarrow [0, 1]$ , also written  $R: A \mapsto B$ .

Convention:  $a, b$  are related by  $R \iff R(a, b) = 0$ .

Let  $R: A \mapsto B, S: B \mapsto C$  and  $f: A \rightarrow B$ .

- *Converse relation*:  $R^\circ(b, a) = R(a, b)$ .
- *Graph of a function*:  $\text{Gr}_f(a, b) = \begin{cases} 0, & \text{if } f(a) = b, \\ 1, & \text{otherwise.} \end{cases}$
- *Composition of relations*:  $(R; S)(a, c) = \inf_{b \in B} R(a, b) \oplus S(b, c)$ .

$$x \oplus y = \min(x + y, 1)$$



# Fuzzy Relations

## Definition

A **fuzzy relation** is a map  $R: A \times B \rightarrow [0, 1]$ , also written  $R: A \mapsto B$ .

Convention:  $a, b$  are related by  $R \iff R(a, b) = 0$ .

Let  $R: A \mapsto B, S: B \mapsto C$  and  $f: A \rightarrow B$ .

- *Converse relation*:  $R^\circ(b, a) = R(a, b)$ .
- *Graph of a function*:  $\text{Gr}_f(a, b) = \begin{cases} 0, & \text{if } f(a) = b, \\ 1, & \text{otherwise.} \end{cases}$
- *Composition of relations*:  $(R; S)(a, c) = \inf_{b \in B} R(a, b) \oplus S(b, c)$ .
- $\varepsilon$ -*diagonal on a set*:  $\Delta_{\varepsilon, A}(a_1, a_2) = \begin{cases} \varepsilon, & \text{if } a_1 = a_2, \\ 1, & \text{otherwise.} \end{cases}$

$$x \oplus y = \min(x + y, 1)$$

## Definition

A **fuzzy lax extension** maps  $R: A \multimap B$  to  $LR: TA \multimap TB$  such that:

$$(L0) \quad L(R^\circ) = (LR)^\circ$$

$$(L1) \quad R_1 \leq R_2 \Rightarrow LR_1 \leq LR_2$$

$$(L2) \quad L(R; S) \leq LR; LS$$

$$(L3) \quad LGr_f \leq Gr_{Tf}$$

We say that  $L$  is **non-expansive**, if additionally

$$(L4) \quad L\Delta_{\varepsilon, A} \leq \Delta_{\varepsilon, TA}$$

where  $A, B, C$  are sets,  $R, R_1, R_2: A \multimap B$ ,  $S: B \multimap C$ ,  $f: A \rightarrow B$ ,  $\varepsilon > 0$ .

## Lemma

*$L$  satisfies Axiom (L4)  $\iff R \mapsto LR$  is non-expansive w.r.t. the supremum metric.*

## Properties of Fuzzy Lax Extensions

### Lemma

$L$  satisfies Axiom (L4)  $\iff R \mapsto LR$  is non-expansive w.r.t. the supremum metric.

### Lemma

If  $d: X \nrightarrow X$  is a pseudometric, then so is  $Ld: TX \nrightarrow TX$ .

Thus,  $L$  gives rise to a **functor lifting** of  $T: \text{Set} \rightarrow \text{Set}$  to a functor  $\bar{T}: \text{PMet} \rightarrow \text{PMet}$ .

category of pseudometric spaces and non-expansive maps



## The Hausdorff Lifting

Classically, bisimulations on Kripke frames arise via the *Egli-Milner extension*:

$$(U, V) \in \overline{\mathcal{P}}(R) \iff (\forall a \in U. \exists b \in V. (a, b) \in R) \wedge (\forall b \in V. \exists a \in U. (a, b) \in R).$$

$\overline{\mathcal{P}}$  is a two-valued lax extension of the powerset functor  $\mathcal{P}$ .

## The Hausdorff Lifting

Classically, bisimulations on Kripke frames arise via the *Egli-Milner extension*:

$$(U, V) \in \overline{\mathcal{P}}(R) \iff (\forall a \in U. \exists b \in V. (a, b) \in R) \wedge (\forall b \in V. \exists a \in U. (a, b) \in R).$$

$\overline{\mathcal{P}}$  is a two-valued lax extension of the powerset functor  $\mathcal{P}$ .

Replacing  $\forall$  with  $\sup$ ,  $\exists$  with  $\inf$ ,  $\wedge$  with  $\max$  gives the **Hausdorff lifting**  $H$ :

$$HR(U, V) = \max(\sup_{a \in U} \inf_{b \in V} R(a, b), \sup_{b \in V} \inf_{a \in U} R(a, b)).$$

$H$  is a non-expansive fuzzy lax extension of  $\mathcal{P}$ .

## Definition

Let  $L$  be a lax extension of  $T$ , and let  $\alpha: A \rightarrow TA$  and  $\beta: B \rightarrow TB$  be coalgebras.

1.  $R: A \rightarrow B$  is an  **$L$ -bisimulation** if  $LR \circ (\alpha \times \beta) \leq R$ .

## Definition

Let  $L$  be a lax extension of  $T$ , and let  $\alpha: A \rightarrow TA$  and  $\beta: B \rightarrow TB$  be coalgebras.

1.  $R: A \rightarrow B$  is an  $L$ -bisimulation if  $LR \circ (\alpha \times \beta) \leq R$ .
2.  $L$ -behavioural distance:  $d_{\alpha, \beta}^L = \inf\{R: A \rightarrow B \mid R \text{ is an } L\text{-bisimulation}\}$ .



## Definition

Let  $L$  be a lax extension of  $T$ , and let  $\alpha: A \rightarrow TA$  and  $\beta: B \rightarrow TB$  be coalgebras.

1.  $R: A \rightarrow B$  is an  $L$ -bisimulation if  $LR \circ (\alpha \times \beta) \leq R$ .
2.  $L$ -behavioural distance:  $d_{\alpha, \beta}^L = \inf\{R: A \rightarrow B \mid R \text{ is an } L\text{-bisimulation}\}$ .

Equivalently,  $d_{\alpha, \beta}^L$  is the least fixed point of  $R \mapsto LR \circ (\alpha \times \beta)$ .

## Definition

Let  $L$  be a lax extension of  $T$ , and let  $\alpha: A \rightarrow TA$  and  $\beta: B \rightarrow TB$  be coalgebras.

1.  $R: A \rightarrow B$  is an  **$L$ -bisimulation** if  $LR \circ (\alpha \times \beta) \leq R$ .
2.  **$L$ -behavioural distance**:  $d_{\alpha,\beta}^L = \inf\{R: A \rightarrow B \mid R \text{ is an } L\text{-bisimulation}\}$ .

Equivalently,  $d_{\alpha,\beta}^L$  is the least fixed point of  $R \mapsto LR \circ (\alpha \times \beta)$ .

$\implies$   $L$ -bisimulations can be used to prove upper bounds for behavioural distance.

## Definition

An  $n$ -ary (fuzzy) *predicate lifting* is a natural transformation

$$\lambda: Q^n \Rightarrow Q \circ T,$$

where  $QX = [0, 1]^X$  is the contravariant fuzzy powerset functor.

## Definition

An  $n$ -ary (fuzzy) *predicate lifting* is a natural transformation

$$\lambda: \mathbf{Q}^n \Rightarrow \mathbf{Q} \circ T,$$

where  $\mathbf{Q}X = [0, 1]^X$  is the contravariant fuzzy powerset functor.

- *Dual* of  $\lambda$ :  $\bar{\lambda}(f_1, \dots, f_n) = 1 - \lambda(1 - f_1, \dots, 1 - f_n)$ .
- $\lambda$  is *monotone* if  $f_1 \leq g_1, \dots, f_n \leq g_n \implies \lambda(f_1, \dots, f_n) \leq \lambda(g_1, \dots, g_n)$ .
- $\lambda$  is *nonexpansive* if

$$\|\lambda_X(f_1, \dots, f_n) - \lambda_X(g_1, \dots, g_n)\|_\infty \leq \max(\|f_1 - g_1\|_\infty, \dots, \|f_n - g_n\|_\infty).$$

## The Kantorovich Lifting

For  $\mu_1, \mu_2 \in \mathcal{D}X$  and  $d: X \rightarrow X$  a metric,

$$Kd(\mu_1, \mu_2) = \sup\{\mathbb{E} \mu_1(f) - \mathbb{E} \mu_2(f) \mid f: (X, d) \rightarrow ([0, 1], d_E) \text{ nonexpansive}\}.$$

# The Kantorovich Lifting

For  $\mu_1, \mu_2 \in DX$  and  $d: X \rightarrow X$  a metric,

$$Kd(\mu_1, \mu_2) = \sup\{\mathbb{E} \mu_1(f) - \mathbb{E} \mu_2(f) \mid f: (X, d) \rightarrow ([0, 1], d_E) \text{ nonexpansive}\}.$$

## Definition (Kantorovich Lifting)

cf. Baldan et al. 2018

Let  $\Lambda$  be a set of monotone predicate liftings that is closed under duals.

For  $R: A \rightarrow B$ ,  $K_\Lambda R: TA \rightarrow TB$  is given by

$$K_\Lambda R(t_1, t_2) = \sup\{\lambda_A(f)(t_1) - \lambda_B(g)(t_2) \mid \lambda \in \Lambda, (f, g) \text{ is } R\text{-nonexpansive}\},$$

where  $(f, g)$  is  $R$ -nonexpansive if  $f(a) - g(b) \leq R(a, b)$  for all  $a \in A, b \in B$ .

# The Kantorovich Lifting

For  $\mu_1, \mu_2 \in \mathcal{D}X$  and  $d: X \rightarrow X$  a metric,

$$Kd(\mu_1, \mu_2) = \sup\{\mathbb{E} \mu_1(f) - \mathbb{E} \mu_2(f) \mid f: (X, d) \rightarrow ([0, 1], d_E) \text{ nonexpansive}\}.$$

## Definition (Kantorovich Lifting)

cf. Baldan et al. 2018

Let  $\Lambda$  be a set of monotone predicate liftings that is closed under duals.

For  $R: A \rightarrow B$ ,  $K_\Lambda R: TA \rightarrow TB$  is given by

$$K_\Lambda R(t_1, t_2) = \sup\{\lambda_A(f)(t_1) - \lambda_B(g)(t_2) \mid \lambda \in \Lambda, (f, g) \text{ is } R\text{-nonexpansive}\},$$

where  $(f, g)$  is  $R$ -nonexpansive if  $f(a) - g(b) \leq R(a, b)$  for all  $a \in A, b \in B$ .

## Theorem

$K_\Lambda$  is a lax extension. If all  $\lambda \in \Lambda$  are nonexpansive, then  $K_\Lambda$  is nonexpansive.

# The Wasserstein Lifting

## Definition (Wasserstein lifting)

cf. Baldan et al. 2018, Hofmann 2007

Let  $\Lambda$  be a set of monotone predicate liftings.

For  $R: A \rightarrow B$ ,  $W_\Lambda R: TA \rightarrow TB$  is given by

$$W_\Lambda R(t_1, t_2) = \sup_{\lambda \in \Lambda} \inf\{\lambda_{A \times B}(R)(t) \mid t \in T(A \times B), T\pi_1(t) = t_1, T\pi_2(t) = t_2\}.$$



# The Wasserstein Lifting

## Definition (Wasserstein lifting)

cf. Baldan et al. 2018, Hofmann 2007

Let  $\Lambda$  be a set of monotone predicate liftings.

For  $R: A \leftrightarrow B$ ,  $W_\Lambda R: TA \leftrightarrow TB$  is given by

$$W_\Lambda R(t_1, t_2) = \sup_{\lambda \in \Lambda} \inf\{\lambda_{A \times B}(R)(t) \mid t \in T(A \times B), T\pi_1(t) = t_1, T\pi_2(t) = t_2\}.$$

Suppose  $T$  preserves weak pullbacks and for each  $\lambda \in \Lambda$ ,

$$\lambda_X(0_X) = 0_{TX} \quad \text{and} \quad \lambda_X(f \oplus g) \leq \lambda_X(f) \oplus \lambda_X(g).$$

## Theorem

$W_\Lambda$  is a lax extension. If all  $\lambda \in \Lambda$  are nonexpansive, then  $W_\Lambda$  is nonexpansive.

### Example (Wasserstein for distributions)

$\mathbb{D}$  has a nonexpansive fuzzy lax extension  $W = W_{\{\lambda\}}$ , where  $\lambda_X(f)(\mu) = \mathbb{E} \mu(f)$ .

# Wasserstein Examples

## Example (Wasserstein for distributions)

$D$  has a nonexpansive fuzzy lax extension  $W = W_{\{\lambda\}}$ , where  $\lambda_X(f)(\mu) = \mathbb{E} \mu(f)$ .

## Example (Hausdorff lifting)

For the Hausdorff lifting  $H$  of  $\mathcal{P}$ , we have  $H = W_{\{\lambda\}}$ , where  $\lambda_X(f)(A) = \sup f[A]$ .

# Wasserstein Examples

## Example (Wasserstein for distributions)

$D$  has a nonexpansive fuzzy lax extension  $W = W_{\{\lambda\}}$ , where  $\lambda_X(f)(\mu) = \mathbb{E} \mu(f)$ .

## Example (Hausdorff lifting)

For the Hausdorff lifting  $H$  of  $\mathcal{P}$ , we have  $H = W_{\{\lambda\}}$ , where  $\lambda_X(f)(A) = \sup f[A]$ .

## Example (Convex powersets)

$\mathcal{C}X =$  nonempty convex subsets of  $DX$ .

$\mathcal{C}$  has a nonexpansive fuzzy lax extension  $L = W_{\{\lambda\}}$ , where  $\lambda_X(f)(A) = \sup_{\mu \in A} \mathbb{E} \mu(f)$ .

# Wasserstein Examples

## Example (Wasserstein for distributions)

$D$  has a nonexpansive fuzzy lax extension  $W = W_{\{\lambda\}}$ , where  $\lambda_X(f)(\mu) = \mathbb{E} \mu(f)$ .

## Example (Hausdorff lifting)

For the Hausdorff lifting  $H$  of  $\mathcal{P}$ , we have  $H = W_{\{\lambda\}}$ , where  $\lambda_X(f)(A) = \sup f[A]$ .

## Example (Convex powersets)

$\mathcal{C}X$  = nonempty convex subsets of  $DX$ .

$\mathcal{C}$  has a nonexpansive fuzzy lax extension  $L = W_{\{\lambda\}}$ , where  
 $\lambda_X(f)(A) = \sup_{\mu \in A} \mathbb{E} \mu(f)$ .

One can show that in fact  $L = H \circ W = H \circ K$ . ← Mio/Vignudelli 2020

# Lax Extensions as Kantorovich Liftings

## Goal

Given a fuzzy lax extension  $L$ , find a set  $\Lambda$  such that  $L = K_\Lambda$ .

# Lax Extensions as Kantorovich Liftings

## Goal

Given a fuzzy lax extension  $L$ , find a set  $\Lambda$  such that  $L = K_\Lambda$ .

## Idea

If the functor  $T$  is *finitary*, it has a **finitary presentation**:

- a *signature*  $\Sigma$  of operations with given finite arities
- for each  $\sigma \in \Sigma$  of arity  $n$  a natural transformation  $\sigma: (-)^n \Rightarrow T$

such that every element of  $TX$  has the form  $\sigma_X(x_1, \dots, x_n)$  for some  $\sigma \in \Sigma$ .

## Definition

Let  $\sigma \in \Sigma$  be  $n$ -ary. The **Moss lifting**  $\mu^\sigma : Q^n \Rightarrow Q \circ T$  is defined as follows:

$$\mu_X^\sigma(f_1, \dots, f_n)(t) = Lev_X(\sigma_{QX}(f_1, \dots, f_n), t),$$

where  $ev_X : QX \rightarrow X$  is given by  $ev_X(f, x) = f(x)$ .



## Definition

Let  $\sigma \in \Sigma$  be  $n$ -ary. The **Moss lifting**  $\mu^\sigma : Q^n \Rightarrow Q \circ T$  is defined as follows:

$$\mu_X^\sigma(f_1, \dots, f_n)(t) = Lev_X(\sigma_{QX}(f_1, \dots, f_n), t),$$

where  $ev_X : QX \rightarrow X$  is given by  $ev_X(f, x) = f(x)$ .

## Theorem

*We have  $L = K_\Lambda$ , where  $\Lambda = \{\mu^\sigma \mid \sigma \in \Sigma\} \cup \{\overline{\mu^\sigma} \mid \sigma \in \Sigma\}$  is the set of all Moss liftings and their duals.*

*Moreover,  $L$  is nonexpansive iff all Moss liftings are nonexpansive.*

## Finitary Separability

### What about non-finitary functors?

Note that every set functor  $T$  has a **finitary part**  $T_\omega$  given by

$$T_\omega X = \bigcup \{Ti[TY] \mid Y \subseteq X \text{ finite, } i: Y \rightarrow X \text{ inclusion}\}.$$

## Finitary Separability

### What about non-finitary functors?

Note that every set functor  $T$  has a **finitary part**  $T_\omega$  given by

$$T_\omega X = \bigcup \{Ti[TY] \mid Y \subseteq X \text{ finite, } i: Y \rightarrow X \text{ inclusion}\}.$$

### Definition

A fuzzy lax extension  $L$  of  $T$  is **finitarily separable** if for every set  $X$ ,  $T_\omega X$  is a dense subset of  $TX$  wrt. to the pseudometric  $L\Delta_X$ .

### Example

The Kantorovich lifting  $K$  of  $D$  is finitarily separable.

# Finitary Separability

## What about non-finitary functors?

Note that every set functor  $T$  has a **finitary part**  $T_\omega$  given by

$$T_\omega X = \bigcup \{Ti[TY] \mid Y \subseteq X \text{ finite, } i: Y \rightarrow X \text{ inclusion}\}.$$

## Definition

A fuzzy lax extension  $L$  of  $T$  is **finitarily separable** if for every set  $X$ ,  $T_\omega X$  is a dense subset of  $TX$  wrt. to the pseudometric  $L\Delta_X$ .

## Example

The Kantorovich lifting  $K$  of  $D$  is finitarily separable.

## Theorem

*If  $L$  is finitarily separable, then the Moss liftings for  $T_\omega$  extend to a set  $\Lambda$  of predicate liftings for  $T$  such that  $L = K_\Lambda$ .*

# Real-valued Coalgebraic Modal Logic

## Syntax of $\mathcal{L}_\Lambda$

$$\varphi, \psi ::= c \mid \varphi \ominus c \mid \neg\varphi \mid \varphi \wedge \psi \mid \lambda(\varphi_1, \dots, \varphi_n) \quad (c \in [0, 1], \lambda \in \Lambda)$$

## Semantics over a coalgebra $\alpha: A \rightarrow TA$

## Syntax of $\mathcal{L}_\Lambda$

$$\varphi, \psi ::= c \mid \varphi \ominus c \mid \neg\varphi \mid \varphi \wedge \psi \mid \lambda(\varphi_1, \dots, \varphi_n) \quad (c \in [0, 1], \lambda \in \Lambda)$$

## Semantics over a coalgebra $\alpha: A \rightarrow TA$

$$\begin{aligned} \llbracket c \rrbracket(a) &= c & \llbracket \varphi \ominus c \rrbracket(a) &= \max(\llbracket \varphi \rrbracket(a) - c, 0) \\ \llbracket \neg\varphi \rrbracket(a) &= 1 - \llbracket \varphi \rrbracket(a) & \llbracket \varphi \wedge \psi \rrbracket(a) &= \min(\llbracket \varphi \rrbracket(a), \llbracket \psi \rrbracket(a)) \\ \llbracket \lambda(\varphi_1, \dots, \varphi_n) \rrbracket(a) &= \lambda_A(\llbracket \varphi_1 \rrbracket, \dots, \llbracket \varphi_n \rrbracket)(\alpha(a)) \end{aligned}$$

## Syntax of $\mathcal{L}_\Lambda$

$$\varphi, \psi ::= c \mid \varphi \ominus c \mid \neg\varphi \mid \varphi \wedge \psi \mid \lambda(\varphi_1, \dots, \varphi_n) \quad (c \in [0, 1], \lambda \in \Lambda)$$

## Semantics over a coalgebra $\alpha: A \rightarrow TA$

$$\begin{aligned} \llbracket c \rrbracket(a) &= c & \llbracket \varphi \ominus c \rrbracket(a) &= \max(\llbracket \varphi \rrbracket(a) - c, 0) \\ \llbracket \neg\varphi \rrbracket(a) &= 1 - \llbracket \varphi \rrbracket(a) & \llbracket \varphi \wedge \psi \rrbracket(a) &= \min(\llbracket \varphi \rrbracket(a), \llbracket \psi \rrbracket(a)) \\ \llbracket \lambda(\varphi_1, \dots, \varphi_n) \rrbracket(a) &= \lambda_A(\llbracket \varphi_1 \rrbracket, \dots, \llbracket \varphi_n \rrbracket)(\alpha(a)) \end{aligned}$$

## Definition

$\Lambda$ -logical distance:  $d^\Lambda(a, b) = \sup\{|\llbracket \varphi \rrbracket(a) - \llbracket \varphi \rrbracket(b)| \mid \varphi \in \mathcal{L}_\Lambda\}$ .

## A Hennessy-Milner Theorem

### Theorem (Fixpoint approximation)

Let  $L$  be a non-expansive and finitarily separable lax extension of  $T$  and let  $\alpha$  and  $\beta$  be  $T$ -coalgebras.

least fixpoint of  $R \mapsto LR \circ (\alpha \times \beta)$

Put  $d_0 = 0$  and  $d_{n+1} = Ld_n \circ (\alpha \times \beta)$  for  $n < \omega$ . Then  $d_{\alpha, \beta}^L \stackrel{\leftarrow}{=} \sup_{n < \omega} d_n$ .



## A Hennessy-Milner Theorem

### Theorem (Fixpoint approximation)

Let  $L$  be a non-expansive and finitarily separable lax extension of  $T$  and let  $\alpha$  and  $\beta$  be  $T$ -coalgebras.

least fixpoint of  $R \mapsto LR \circ (\alpha \times \beta)$

Put  $d_0 = 0$  and  $d_{n+1} = Ld_n \circ (\alpha \times \beta)$  for  $n < \omega$ . Then  $d_{\alpha, \beta}^L \stackrel{\leftarrow}{=} \sup_{n < \omega} d_n$ .

For Kantorovich extensions  $K_\Lambda$ , this is known to imply  $d^{K_\Lambda} = d^\Lambda$ .

König/Mika-Michalski 2018

## A Hennessy-Milner Theorem

### Theorem (Fixpoint approximation)

Let  $L$  be a non-expansive and finitarily separable lax extension of  $T$  and let  $\alpha$  and  $\beta$  be  $T$ -coalgebras.

least fixpoint of  $R \mapsto LR \circ (\alpha \times \beta)$

Put  $d_0 = 0$  and  $d_{n+1} = Ld_n \circ (\alpha \times \beta)$  for  $n < \omega$ . Then  $d_{\alpha, \beta}^L \stackrel{\leftarrow}{=} \sup_{n < \omega} d_n$ .

For Kantorovich extensions  $K_\Lambda$ , this is known to imply  $d^{K_\Lambda} = d^\Lambda$ .

As a corollary, we get:

König/Mika-Michalski 2018

### Theorem (Hennessy-Milner Theorem for Lax Extensions)

Let  $L$  be a non-expansive finitarily separable fuzzy lax extension. Then there exists a set  $\Lambda$  of monotone non-expansive predicate liftings such that  $L = K_\Lambda$  and  $d^\Lambda = d^L$ .

$\implies \mathcal{L}_\Lambda$  is a **characteristic logic** for  $L$ .

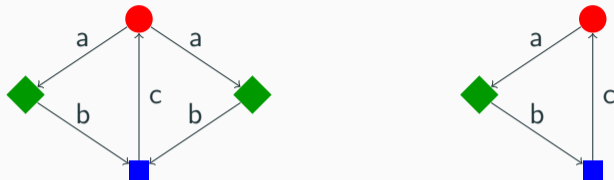
# A Quantified Coalgebraic van Benthem Theorem (FoSSaCS'21)

Paul Wild and Lutz Schröder

A **Quantified Quantitative** Coalgebraic van Benthem  
Theorem (FoSSaCS'21)

Paul Wild and Lutz Schröder

## Introduction – Bisimulation invariance



*Bisimilar states*: indistinguishable in terms of successor behaviour.

*Bisimulation invariant* properties:

$\diamond_a \varphi$  = there exists an  $a$ -successor satisfying  $\varphi$

$\square_b \varphi$  = all  $b$ -successors satisfy  $\varphi$

## Introduction – Modal logic

A *syntax* for bisimulation-invariant properties:

$$\varphi, \psi ::= \top \mid \varphi \wedge \psi \mid \neg\varphi \mid \Diamond_a\varphi \mid \Box_a\varphi \quad (a \text{ label})$$

### Lemma

*Every modal formula is bisimulation-invariant.*

## Introduction – Modal logic

A *syntax* for bisimulation-invariant properties:

$$\varphi, \psi ::= \top \mid \varphi \wedge \psi \mid \neg\varphi \mid \diamond_a\varphi \mid \square_a\varphi \quad (a \text{ label})$$

### Lemma

*Every modal formula is bisimulation-invariant.*

### Theorem (Hennessy-Milner Theorem)

*In **finitely branching systems**, two states agreeing on all modal formulae are bisimilar.*

# Introduction – Modal logic

A *syntax* for bisimulation-invariant properties:

$$\varphi, \psi ::= \top \mid \varphi \wedge \psi \mid \neg\varphi \mid \Diamond_a\varphi \mid \Box_a\varphi \quad (a \text{ label})$$

## Lemma

*Every modal formula is bisimulation-invariant.*

## Theorem (Hennessy-Milner Theorem)

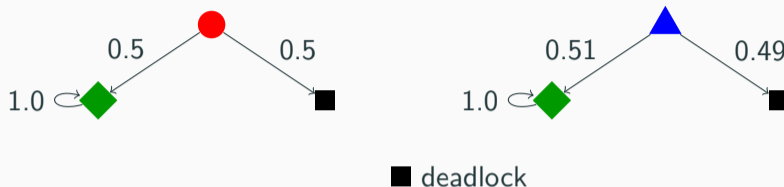
*In **finitely branching systems**, two states agreeing on all modal formulae are bisimilar.*

## Theorem (van Benthem Theorem)

*If a **first-order** property is bisimulation-invariant, it is equivalent to a modal formula.*



# Introduction – Markov chains



Behavioural distance  $d$  with  $d(\bullet, \blacktriangle) = 0.01$

Real-valued probabilistic modal logic with  $\llbracket \varphi \rrbracket(x) \in [0, 1]$ :

- $\mathbb{E} \varphi =$  expected truth value of  $\varphi$  over successors
- Modal formulae are *non-expansive* wrt.  $d$ :  $\llbracket \varphi \rrbracket(x) - \llbracket \varphi \rrbracket(y) \leq d(x, y)$

Probabilistic Hennessy-Milner Theorem: [van Breugel/Worrell 2005]

Probabilistic van Benthem Theorem: [Wild/Schröder/Pattinson/König 2019]

# Introduction – Simulations



■ simulates ●.

Syntax for properties *preserved under simulation*:

$$\varphi, \psi ::= \perp \mid \top \mid \varphi \wedge \psi \mid \varphi \vee \psi \mid \Diamond_a \varphi \quad (a \text{ label})$$

Hennessy-Milner Theorem for simulations: [van Glabbeek 2001]

van Benthem Theorem for simulations: [Lutz/Piro/Wolter 2010]

## Our Contribution – Overview

### Goal

General versions of the Hennessy-Milner and van Benthem Theorems that have all the previous examples as instances.

# Our Contribution – Overview

## Goal

General versions of the Hennessy-Milner and van Benthem Theorems that have all the previous examples as instances.

## Key Ingredients

- an algebra of truth values  $\rightsquigarrow$  *value co-quantale*  $\mathcal{V}$
- abstraction over system types  $\rightsquigarrow$  *T-coalgebras of a functor*  $T$
- a representation of the modalities  $\rightsquigarrow$  *set of predicate liftings*  $\Lambda$

# Our Contribution – Overview

## Goal

General versions of the Hennessy-Milner and van Benthem Theorems that have all the previous examples as instances.

## Key Ingredients

- an algebra of truth values  $\rightsquigarrow$  *value co-quantale*  $\mathcal{V}$
- abstraction over system types  $\rightsquigarrow$  *T-coalgebras of a functor*  $T$
- a representation of the modalities  $\rightsquigarrow$  *set of predicate liftings*  $\Lambda$

## Idea

Modal logic  $\mathcal{L}_\Lambda$  characterizes *non-expansiveness* wrt. behavioural distance  $d^K$ .

## Value co-quantale $\mathcal{V}$

[Flagg, 1997]

- Completely distributive lattice  $(V, \leq)$
- Monoid structure  $\oplus$  that distributes over meets:  $a \oplus \bigwedge_{i \in I} b_i = \bigwedge_{i \in I} a \oplus b_i$ .
- Subtraction  $a \ominus b \leq c \iff a \leq b \oplus c$ .
- Filter of positive elements  $\{\varepsilon \mid \varepsilon \gg 0\}$ .

## Key properties

$$0 = \bigwedge \{\varepsilon \mid \varepsilon \gg 0\} \quad \text{and} \quad \varepsilon \gg 0 \implies \exists \delta \gg 0. \delta \oplus \delta \leq \varepsilon$$

## Value co-quantale $\mathcal{V}$

[Flagg, 1997]

- Completely distributive lattice  $(V, \leq)$
- Monoid structure  $\oplus$  that distributes over meets:  $a \oplus \bigwedge_{i \in I} b_i = \bigwedge_{i \in I} a \oplus b_i$ .
- Subtraction  $a \ominus b \leq c \iff a \leq b \oplus c$ .
- Filter of positive elements  $\{\varepsilon \mid \varepsilon \gg 0\}$ .

## Key properties

$$0 = \bigwedge \{\varepsilon \mid \varepsilon \gg 0\} \quad \text{and} \quad \varepsilon \gg 0 \implies \exists \delta \gg 0. \delta \oplus \delta \leq \varepsilon$$

## Main Examples

$$2 = \{0, 1\} \quad [0, 1] \quad \{[a, b] \mid 0 \leq a \leq b \leq 1\}$$

$$\gamma: X \rightarrow TX \quad (T \text{ endofunctor on Set})$$

Some choices of  $T$ :

- LTS with edge labels in  $A$ :  $TX = \mathcal{P}(A \times X)$
- Markov chains with deadlocks:  $TX = 1 + \mathcal{D}X$
- Metric transition systems with state labels in  $(S, d_S)$ :  $TX = S \times \mathcal{P}X$



## Predicate Lifting

$\lambda_X: (X \rightarrow V) \rightarrow (TX \rightarrow V)$ , natural, monotone and non-expansive

## Predicate Lifting

$\lambda_X: (X \rightarrow V) \rightarrow (TX \rightarrow V)$ , natural, monotone and non-expansive

## Syntax of QCML

$$\varphi, \psi ::= c \mid \varphi \oplus c \mid \varphi \ominus c \mid \varphi \wedge \psi \mid \varphi \vee \psi \mid \lambda \varphi \quad (c \in V, \lambda \in \Lambda).$$

## Predicate Lifting

$\lambda_X: (X \rightarrow V) \rightarrow (TX \rightarrow V)$ , natural, monotone and non-expansive

## Syntax of QCML

$\varphi, \psi ::= c \mid \varphi \oplus c \mid \varphi \ominus c \mid \varphi \wedge \psi \mid \varphi \vee \psi \mid \lambda\varphi \quad (c \in V, \lambda \in \Lambda).$

## Semantics over $\gamma: X \rightarrow TX$

$\llbracket \varphi \rrbracket_\gamma: X \rightarrow V$  recursively defined with  $\llbracket \lambda\varphi \rrbracket_\gamma = \lambda_X(\llbracket \varphi \rrbracket_\gamma) \circ \gamma.$

# Quantitative Coalgebraic Modal Logic

## Predicate Lifting

$\lambda_X: (X \rightarrow V) \rightarrow (TX \rightarrow V)$ , natural, monotone and non-expansive

## Syntax of QCML

$\varphi, \psi ::= c \mid \varphi \oplus c \mid \varphi \ominus c \mid \varphi \wedge \psi \mid \varphi \vee \psi \mid \lambda\varphi \quad (c \in V, \lambda \in \Lambda).$

## Semantics over $\gamma: X \rightarrow TX$

$\llbracket \varphi \rrbracket_\gamma: X \rightarrow V$  recursively defined with  $\llbracket \lambda\varphi \rrbracket_\gamma = \lambda_X(\llbracket \varphi \rrbracket_\gamma) \circ \gamma$ .

## Example

Probabilistic modal logic:  $\mathbb{E} X(f)(\mu) =$  expected value of  $f$  under  $\mu$

# Behavioural Distance via Relation Lifting

## Kantorovich Lifting

$$(R: A \times B \rightarrow V) \quad \mapsto \quad (K_{\Lambda}(R): TA \times TB \rightarrow V)$$

# Behavioural Distance via Relation Lifting

## Kantorovich Lifting

$$(R: A \times B \rightarrow V) \quad \mapsto \quad (K_\Lambda(R): TA \times TB \rightarrow V)$$

$$K_\Lambda(R)(t_1, t_2) = \bigvee \{ \lambda_A(f)(t_1) \ominus \lambda_B(g)(t_2) \mid \lambda \in \Lambda, \forall a, b. f(a) \ominus g(b) \leq R(a, b) \}$$

# Behavioural Distance via Relation Lifting

## Kantorovich Lifting

$$(R: A \times B \rightarrow V) \quad \mapsto \quad (K_\Lambda(R): TA \times TB \rightarrow V)$$

$$K_\Lambda(R)(t_1, t_2) = \bigvee \{ \lambda_A(f)(t_1) \ominus \lambda_B(g)(t_2) \mid \lambda \in \Lambda, \forall a, b. f(a) \ominus g(b) \leq R(a, b) \}$$

## Behavioural distance as least fixed point

$$d^K = K_\Lambda(d^K) \circ (\gamma \times \gamma)$$

$$d^K(x, x) = 0 \quad \text{and} \quad d^K(x, z) \leq d^K(x, y) \oplus d^K(y, z)$$

# Behavioural Distance via Relation Lifting

## Kantorovich Lifting

$$(R: A \times B \rightarrow V) \quad \mapsto \quad (K_\Lambda(R): TA \times TB \rightarrow V)$$

$$K_\Lambda(R)(t_1, t_2) = \bigvee \{ \lambda_A(f)(t_1) \ominus \lambda_B(g)(t_2) \mid \lambda \in \Lambda, \forall a, b. f(a) \ominus g(b) \leq R(a, b) \}$$

## Behavioural distance as least fixed point

$$d^K = K_\Lambda(d^K) \circ (\gamma \times \gamma)$$

$$d^K(x, x) = 0 \quad \text{and} \quad d^K(x, z) \leq d^K(x, y) \oplus d^K(y, z)$$

$$d^K(x, y) = d^K(y, x) \text{ if } \Lambda \text{ closed under duals}$$



# Quantitative Hennessy-Milner Theorem

## Theorem (Quantitative Hennessy-Milner theorem)

Let  $\Lambda$  be finite and  $\mathcal{V}$  totally bounded and continuous from below.

If  $T$  is *finitary*, then

$$d^K(a, b) = \bigvee \{ \llbracket \varphi \rrbracket(a) \ominus \llbracket \varphi \rrbracket(b) \mid \varphi \text{ a modal formula} \}.$$

# Quantitative van Benthem Theorem

## Quantitative Coalgebraic *Predicate* Logic

$$\varphi, \psi ::= c \mid x = y \mid \varphi \oplus c \mid \varphi \ominus c \mid \varphi \wedge \psi \mid \varphi \vee \psi \mid \exists x. \varphi \mid \forall x. \varphi \mid x \lambda [y: \varphi]$$

## Theorem (Quantitative van Benthem theorem)

Let  $\Lambda$  be finite and  $\mathcal{V}$  totally bounded. Let  $\varphi \in \text{QCPL}$  be non-expansive wrt.  $d^K$ .

For every  $\varepsilon \gg 0$  there exists a modal formula  $\psi$  such that for all  $\gamma, x$ :

$$\llbracket \varphi \rrbracket_\gamma(x) \ominus \llbracket \psi \rrbracket_\gamma(x) \leq \varepsilon \quad \text{and} \quad \llbracket \psi \rrbracket_\gamma(x) \ominus \llbracket \varphi \rrbracket_\gamma(x) \leq \varepsilon$$

Existing instances of Hennessy-Milner and van Benthem theorems we cover:

- Classical modal logic with  $\mathcal{V} = 2$  and  $TX = \mathcal{P}(A \times X)$
- Probabilistic modal logic with  $\mathcal{V} = [0, 1]$  and  $TX = 1 + \mathcal{D}X$
- Two-valued ( $\mathcal{V} = 2$ ) coalgebraic modal logic [Schröder/Pattinson/Litak 2017]

Existing instances of Hennessy-Milner and van Benthem theorems we cover:

- Classical modal logic with  $\mathcal{V} = 2$  and  $TX = \mathcal{P}(A \times X)$
- Probabilistic modal logic with  $\mathcal{V} = [0, 1]$  and  $TX = 1 + \mathcal{D}X$
- Two-valued ( $\mathcal{V} = 2$ ) coalgebraic modal logic [Schröder/Pattinson/Litak 2017]

New instances include:

- Metric modal logic with  $TX = S \times \mathcal{P}X$  and modalities based on  $d_S$ .
- For  $\mathcal{V} = \{[a, b] \mid 0 \leq a \leq b \leq 1\}$ : convex-nondeterministic metric modal logic.
- Simulation-based versions of all the above.