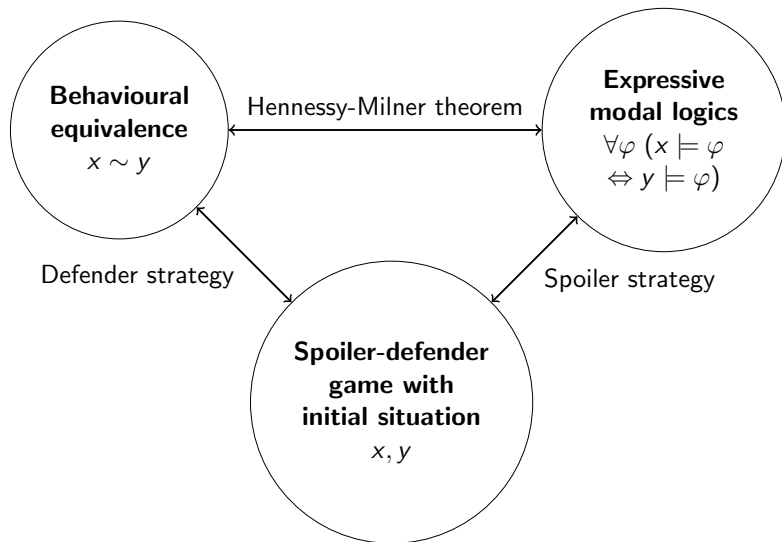


# Metric Bisimulation Games and Real-Valued Modal Logics for Coalgebras (CONCUR '18)

Barbara König and Christina Mika-Michalski  
Universität Duisburg-Essen

## Behavioural Equivalences, Modal Logics &amp; Games



# Behavioural Equivalences, Modal Logics & Games

Games are interesting since . . .

- they explain the essence of behavioural equivalences.
- they can be used to explain why two states are not behaviourally equivalent ( $x \not\sim y$ ) by giving the spoiler strategy.
- they can be used to prove interesting theorems (for instance the van Benthem/Rosen theorem: “the bisimulation-invariant fragment of first-order logic corresponds to propositional modal logics”).

There is little work on games that goes beyond labelled transition systems [Desharnais et al., Fijalkow et al.] and only few paper on games [Baltag, Kupke] in a coalgebraic setting.

# Behavioural Metrics

Find a **quantitative notion of behavioural equivalence** ...

- Do not insist on the exact same behaviour.
- Measure the **behavioural distance** between two states.
- Make statements such as “the behaviour of two states differs only by  $\varepsilon$ ”.

↪ **Behavioural metrics**

**Applications:** differential privacy [Palamidessi], conformance testing for cyber-physical systems [Khakpour, Mousavi]

**Here:** work out the triad of behavioural distance, modal logics and spoiler-defender games for behavioural metrics.

# Behavioural distances

## Pseudo-metric space

Let  $X$  be a set. A **pseudo-metric** is a function  $d : X \times X \rightarrow [0, 1]$  where for all  $x, y, z \in X$ :

1.  $d(x, x) = 0$  (reflexivity) (**metric** if  $d(x, y) = 0 \Rightarrow x = y$ )
2.  $d(x, y) = d(y, x)$  (symmetry)
3.  $d(x, z) \leq d(x, y) + d(y, z)$  (triangle inequality)

A **(pseudo-)metric space** is a pair  $(X, d)$  where  $X$  is a set and  $d$  is a (pseudo-)metric.

A **directed pseudo-metric** must only satisfy reflexivity and the triangle inequality.

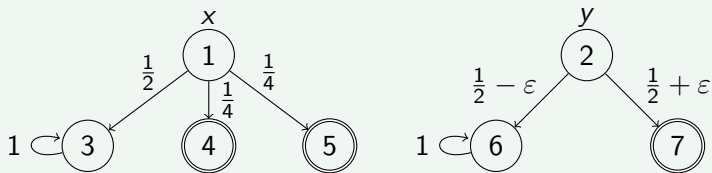
## Example: Probabilistic Transition System

### Probabilistic transition system

A **probabilistic transition system** is a tuple  $P = (X, T, p)$ , where  $X$  is a set of states,  $T \subseteq X$  is the set of terminal states and every state  $x \notin T$  is assigned a probability distribution  $p_x : X \rightarrow [0, 1]$ .

Similar models studied by [Larsen, Skou], [van Breugel, Worrel]

# Example: Probabilistic Transition System



States 4, 5, 7 are terminal.

What is the distance between states  $x, y$ ?

# Example: Probabilistic Transition System

Distance of states in a probabilistic transition system

Compute the smallest fixed-point of

$$d(x, y) = \begin{cases} 1 & \text{if } x \in T, y \notin T \text{ or } x \notin T, y \in T \\ 0 & \text{if } y \in T, y \in T \\ d^P(p_x, p_y) & \text{otherwise} \end{cases}$$

What does it mean to compute the distance between two probability distributions  $p_x, p_y$  on a metric space (i.e. to lift the distance to probability distributions)?



## Example: Probabilistic Transition System

### Non-expansive function

A **non-expansive function**  $f: X \rightarrow Y$  between two (pseudo-) metric spaces  $(X, d_X), (Y, d_Y)$  satisfies for  $x, y \in X$ :

$$d_X(x, y) \geq d_Y(f(x), f(y))$$

The category of pseudo-metric spaces and non-expansive functions is denoted by **PMet**.

## Example: Probabilistic Transition System

We obtain the distance for two probability distributions  $p, q : X \rightarrow [0, 1]$  by lifting a metric  $d : X \times X \rightarrow [0, 1]$  as follows:

### Kantorovich lifting for probability distributions

The **supremum** of  $|\sum_{x \in X} f(x) \cdot p(x) - \sum_{x \in X} f(x) \cdot q(x)|$  for all non-expansive functions  $f : (X, d) \rightarrow ([0, 1], d_e)$  (where  $d_e(a, b) = |a - b|$ )

**Intuition:** we measure the cost of transporting supply  $p$  to demand  $q$  from the point of view of a logistics company [Villani]

- $f$  is a price function, which specifies buying/selling prices at each location
- Non-expansiveness means  $f(x) - f(y) \leq d(x, y)$  for all  $x, y$ . If this is not satisfied the company is not hired, since it is cheaper for the customer to perform the transport himself.

# Coalgebras

## Coalgebras & Coalgebra Homomorphism

Let  $F$  be a functor. A **coalgebra** is a function  $\alpha : X \rightarrow FX$  (where  $X$  is the state set).

We call  $f : X \rightarrow Z$  a coalgebra homomorphism from  $\alpha$  to  $\beta : Y \rightarrow FY$  whenever  $Ff \circ \alpha = \beta \circ f$ .

$$\begin{array}{ccc}
 X & \xrightarrow{\alpha} & FX \\
 f \downarrow & & \downarrow Ff \\
 Y & \xrightarrow{\beta} & FY
 \end{array}$$

Two states  $x, y \in X$  are behaviourally equivalent ( $x \sim y$ ) if there exists a coalgebra homomorphism  $f$  with  $f(x) = f(y)$ .

Probabilistic transition system:  $\alpha : X \rightarrow \mathcal{D}X + 1$  where  $\mathcal{D}$  is the probability distribution functor and  $1 = \{\sqrt{\cdot}\}$ .

# Coalgebras & Behavioural Metrics

**Aim:** Lift the functor  $F$  from **Set** to **PMet**

## Evaluation maps

We need one parameter: a set  $\Gamma$  of evaluation maps

$$\gamma: F[0, 1] \rightarrow [0, 1].$$

## Example: Probabilistic transition systems, $FX = \mathcal{D}X + 1$

We define two evaluation maps (one for probability distributions, one for termination):

- $\gamma_{\mathcal{D}}: \mathcal{D}[0, 1] + 1 \rightarrow [0, 1]$  with
  - $\gamma_{\mathcal{D}}(p) = \sum_{r \in [0, 1]} r \cdot p(r)$ ,  $p: X \rightarrow [0, 1]$  (expectation)
  - $\gamma_{\mathcal{D}}(\surd) = 0$
- $\gamma_{\bullet}: \mathcal{D}[0, 1] + 1 \rightarrow [0, 1]$  with
  - $\gamma_{\bullet}(p) = 0$
  - $\gamma_{\bullet}(\surd) = 1$

# Functor Lifting via Kantorovich

## Kantorovich Lifting (for a functor $F$ )

Let  $d : X \times X \rightarrow [0, 1]$  be a pseudo-metric. We define  $d^{\uparrow\Gamma} : FX \times FX \rightarrow [0, 1]$ . Let  $t_1, t_2 \in FX$ :

$$d^{\uparrow\Gamma}(t_1, t_2) = \sup\{d_e(\gamma(Ff(t_1)), \gamma(Ff(t_2))) \mid f : (X, d) \rightarrow ([0, 1], d_e) \text{ non-expansive}, \gamma \in \Gamma\}$$

Lifting **directed pseudo-metrics**: exchange  $d_e(a, b) = |a - b|$  with  $d_{\ominus}(a, b) = \max\{a - b, 0\}$ .

## Functor Lifting via Kantorovich

Every evaluation map  $\gamma$  induces a (real-valued) predicate lifting

$$(f: X \rightarrow [0, 1]) \quad \mapsto \quad (\gamma \circ Ff: FX \rightarrow [0, 1])$$

We require that this predicate lifting is **non-expansive wrt. the supremum metric** (local non-expansiveness, see also [Turi, Rutten]):

$$d_e^\infty(\gamma \circ Ff, \gamma \circ Fg) \leq d_e^\infty(f, g)$$

for all  $f, g: X \rightarrow [0, 1]$ . (The same holds if we replace  $d_e$  by  $d_{\ominus}$ .)

**Supremum metric:**  $d^\infty(f, g) = \sup_{x \in X} d(f(x), g(x))$

In the directed case this condition generalizes **monotonicity**.

# Defining Distances in Coalgebras

Coalgebraic behavioural metrics [Baldan, Bonchi, Kerstan, K.]

Given a set of evaluation maps  $\Gamma$  and coalgebra  $\alpha : X \rightarrow FX$  in **Set**, compute its associated **behavioural metric**  $d : X \times X \rightarrow [0, 1]$  as the smallest fixed-point of:

$$d = d^{\uparrow\Gamma} \circ (\alpha \times \alpha)$$

where  $d^{\uparrow\Gamma}$  is the Kantorovich lifting.

This instantiates to the earlier definition of probabilistic behavioural metric.

## Real-Valued Modal Logics

Behavioural metrics in the probabilistic case are often associated with **real-valued modal logics** with formulas that assign real numbers to states (instead of truth values  $0, 1$ ), indicating the “degree” up to which a formula holds.

Given a coalgebra  $\alpha: X \rightarrow FX$ , the semantics of a formula  $\varphi$  is given by a function  $\llbracket \varphi \rrbracket_\alpha: X \rightarrow [0, 1]$ .

$\varphi:$	$\mathbf{1}$	$[\gamma]\psi, \gamma \in \Gamma$	$\min(\psi, \psi')$
$\llbracket \varphi \rrbracket_\alpha:$	$\mathbf{1}$	$\gamma \circ F\llbracket \psi \rrbracket_\alpha \circ \alpha$	$\min\{\llbracket \psi \rrbracket_\alpha, \llbracket \psi' \rrbracket_\alpha\}$

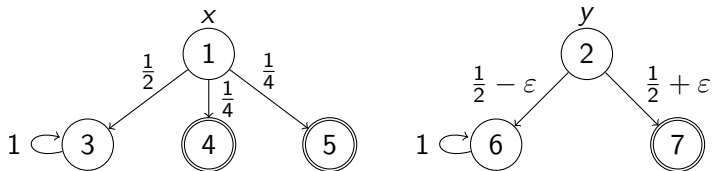
$\varphi:$	$\neg\psi$	$\psi \ominus \mathbf{q}$
$\llbracket \varphi \rrbracket_\alpha:$	$\mathbf{1} - \llbracket \psi \rrbracket_\alpha$	$\llbracket \psi \rrbracket_\alpha \ominus \mathbf{q}$

The modalities  $[\gamma]\psi$  are based on the evaluation maps  $\gamma \in \Gamma$ .

Every function  $\llbracket \varphi \rrbracket_\alpha$  is **non-expansive** (analogue of bisimulation-invariant).



## Real-Valued Modal Logics



The formula  $\varphi = [\gamma_{\mathcal{D}}][\gamma_{\bullet}]1$  distinguishes the states  $x, y$ .

- $\psi = [\gamma_{\bullet}]1$  assigns 1 to terminating states and 0 to non-terminating states.
- $x$  makes a transition to a terminating state with probability  $\frac{1}{2}$   
 $\Rightarrow \llbracket \varphi \rrbracket(x) = \gamma_{\mathcal{D}}(\mathcal{D}[\llbracket \psi \rrbracket](\alpha(x))) = \frac{1}{2}$ .  
 Similarly:  $\llbracket \varphi \rrbracket(y) = \frac{1}{2} + \varepsilon$ .

Hence  $d_{\alpha}^L(x, y) \geq d_e(\llbracket \varphi \rrbracket(x), \llbracket \varphi \rrbracket(y)) = \varepsilon$ . (In fact we have equality.)

# A Hennessy-Milner Theorem Real-Valued Modal Logics

## Logical Distance

Let  $\alpha: X \rightarrow FX$ ,  $x, y \in X$ :

$$d_\alpha^L(x, y) = \sup\{d_e(\llbracket\varphi\rrbracket_\alpha(x), \llbracket\varphi\rrbracket_\alpha(y)) \mid \varphi\}.$$

## Hennessy-Milner Theorem

Assume that the fixpoint  $d_\alpha$  is reached in  $\omega$  steps.

Then behavioural distance equals logical distance:  $d_\alpha = d_\alpha^L$ .

**Proof strategy** (adapted from [van Breugel/Worrel]):

- Show that the lifted functor preserves total boundedness.
- Show that the set  $\{\llbracket\varphi\rrbracket: X \rightarrow [0, 1] \mid \varphi\}$  is dense in the non-expansive functions.

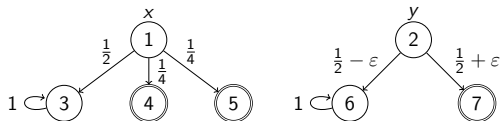
# Games for behavioural distances

Metric bisimulation game for a coalgebra  $\alpha: X \rightarrow FX$

- **Initial situation:**  $(x, y, \varepsilon)$  where  $\varepsilon \in [0, 1]$ .
- **Step 1:** Spoiler (S) chooses  $s \in \{x, y\}$  and a (real-valued) predicate  $p_1: X \rightarrow [0, 1]$ .
- **Step 2:** Defender (D) takes  $t \in \{x, y\} \setminus \{s\}$  and has to answer with a predicate  $p_2: X \rightarrow [0, 1]$ , which satisfies
 
$$d_{\ominus}(\gamma(Fp_1(\alpha(s))), \gamma(Fp_2(\alpha(t)))) \leq \varepsilon \text{ for all } \gamma \in \Gamma.$$
- **Step 3:** S chooses  $p_i$  with  $i \in \{1, 2\}$  and a state  $x' \in X$ .
- **Step 4:** D chooses  $y' \in X$  with  $p_i(x') \leq p_j(y')$  where  $j \neq i$
- **Next round:**  $(x', y', \varepsilon')$  with  $\varepsilon' = p_j(y') - p_i(x')$ .

D wins the game if the game continues forever or if S has no move at Step 3. If D has no move at Step 2 or Step 4, S wins.

## Games for behavioural distances



- **Initial situation:**  $(x, y, \varepsilon)$
- **Step 1:** S chooses  $x$  with  $p_1(4) = 1$  (0 otherwise).
- **Step 2:** D plays  $p_2$  with  $p_2(4) = p_2(5) = p_2(7) = 1$  (0 otherwise).

$$d_{\ominus}(\gamma_{\mathcal{D}}(\mathcal{D}p_1(\alpha(x))), \gamma_{\mathcal{D}}(\mathcal{D}p_2(\alpha(y)))) = d_{\ominus}\left(\frac{1}{4}, \frac{1}{2} + \varepsilon\right) = 0 \leq \varepsilon$$

(Similar for  $\gamma_{\bullet}$ .)

- **Step 3/4:** If S picks a terminating state  $x'$  and  $p_i$ , D can also pick a terminating state  $y'$  and  $p_j$  with  $p_j(y') - p_i(x') = 0$  (similarly for non-terminating states).  
New situation:  $(x', y', 0)$  where  $x' \sim y'$ .

# Games for behavioural distances

## Game Distance

For a coalgebra  $\alpha: X \rightarrow FX$  the game distance is defined as:

$$d_{\alpha}^G(x, y) = \inf\{\varepsilon \mid D \text{ has a winning strategy for } (x, y, \varepsilon)\}$$

## Behavioural Distance and Game Distance

The behavioural distance equals the game distance:  $d_{\alpha} = d_{\alpha}^G$ .

# Games for behavioural distances

## Spoiler Strategy

Take a modal formula  $\varphi \neq 1$  with  $d_{\ominus}(\llbracket \varphi \rrbracket(x), \llbracket \varphi \rrbracket(y)) > \varepsilon$ .

- $\varphi = [\gamma]\psi$ : S chooses  $x$ ,  $p_1 = \llbracket \psi \rrbracket$  at Step 1 and can play in such a way that  $d_e(\llbracket \psi \rrbracket(x'), \llbracket \psi \rrbracket(y')) > \varepsilon'$ .  
 $\leadsto$  the game continues in the situation  $(x', y', \varepsilon')$  with the formula  $\psi$ .
- $\varphi = \min(\psi, \psi')$ : either  $d_e(\llbracket \psi \rrbracket(x), \llbracket \psi \rrbracket(y)) > \varepsilon$  or  $d_e(\llbracket \psi' \rrbracket(x), \llbracket \psi' \rrbracket(y)) > \varepsilon$  and S picks  $\psi$  or  $\psi'$ .
- $\varphi = \neg\psi$ : S takes  $\psi$ .
- $\varphi = \psi \ominus q$ : S takes  $\psi$ .

# Open Questions & Future Work

## Open Questions

- Hennessy-Milner theorem if the fixpoint  $d_\alpha$  is not reached in  $\omega$  steps.
- Does the Kantorovich lifting preserve completeness of metrics?
- Can we replace  $[0, 1]$  by  $[0, \infty]$ ?

## Current & Future Work

- Work out the real-valued modal logics for various functors (in the paper: metric transition systems [de Alfaro, Faella, Stoelinga], functor  $FX = [0, 1] \times \mathcal{P}X$ )
- van Benthem/Rosen theorem(s)  
 $\rightsquigarrow$  Fuzzy case: [Wild, Schröder, Pattinson, K., LICS '18]
- Up-to techniques for behavioural metrics [Bonchi, K., Petrisan, CONCUR '18]