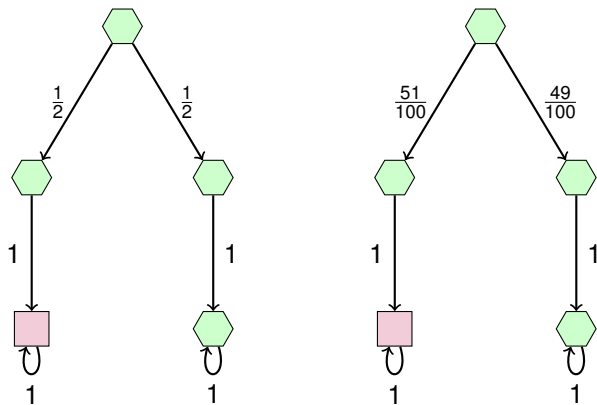


How to Explain Probabilistic Bisimilarity Distances?

Franck van Breugel
York University, Toronto

Probabilistic bisimilarity is *not* robust



Probabilistic bisimilarity is *not* robust



Giacalone, Jou, and Smolka, PROCOMET 1990.

Probabilistic bisimilarity is *not* robust

Fundamental problem

Probabilistic bisimilarity is *not* robust.

Probabilistic bisimilarity is *not* robust

Fundamental problem

Probabilistic bisimilarity is *not* robust.

Robust alternative

Instead of an equivalence relation

$$\sim : \mathcal{S} \times \mathcal{S} \rightarrow \{\text{true}, \text{false}\}$$

use **distances**

$$d : \mathcal{S} \times \mathcal{S} \rightarrow [0, 1].$$

Probabilistic bisimilarity is *not* robust

Fundamental problem

Probabilistic bisimilarity is *not* robust.

Robust alternative

Instead of an equivalence relation

$$\sim : S \times S \rightarrow \{\text{true}, \text{false}\}$$

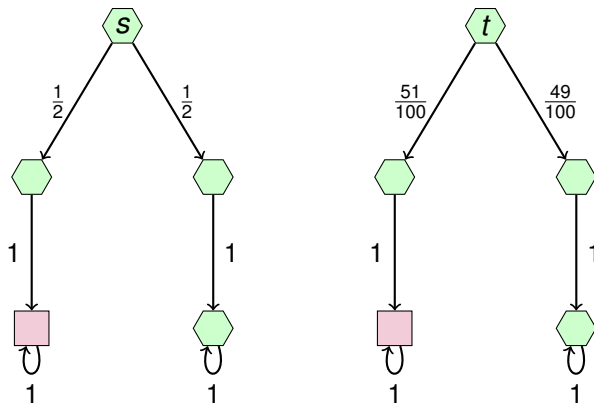
use **distances**

$$d : S \times S \rightarrow [0, 1].$$

Generalization

For all states s and t , $s \sim t$ iff $d(s, t) = 0$.

Probabilistic bisimilarity distances



Question

How do we explain that the distance of s and t is $\frac{1}{100}$?

Question

How do we explain why states of a labelled transition system are *not* bisimilar?

Question

How do we explain why states of a labelled transition system are *not* bisimilar?

- Game
Ehrenfeucht and Fraïssé, 1950 and 1961.
- Logic
Hennessy and Milner, 1980.
- ...

Theorem

States s and t of a labelled transition system are bisimilar iff duplicator has a winning strategy from (s, t) .

Theorem

States s and t of a labelled transition system are bisimilar iff duplicator has a winning strategy from (s, t) .

Question

How do we explain that states of a labelled transition system are *not* bisimilar?

Theorem

States s and t of a labelled transition system are bisimilar iff duplicator has a winning strategy from (s, t) .

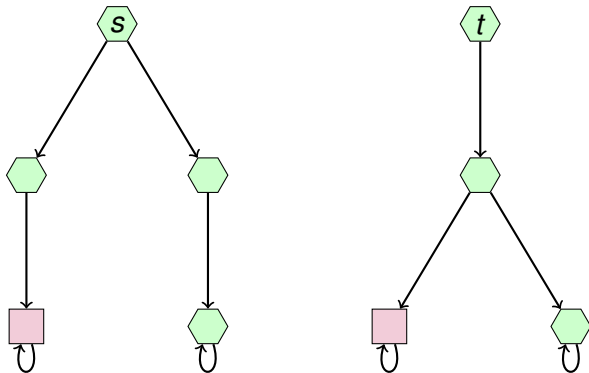
Question

How do we explain that states of a labelled transition system are *not* bisimilar?

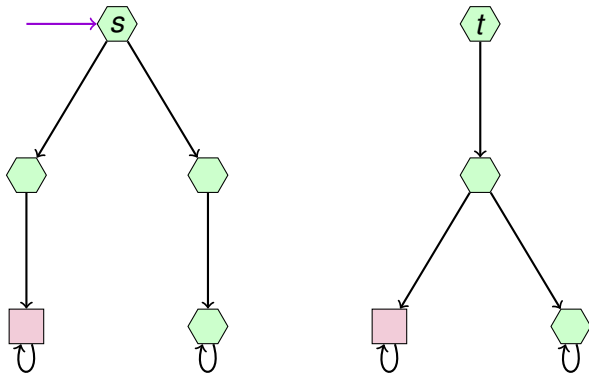
Answer

Find a strategy for spoiler that is winning from (s, t) .

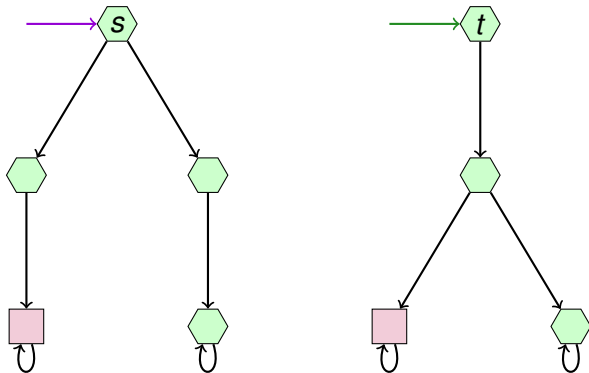
Bisimilarity



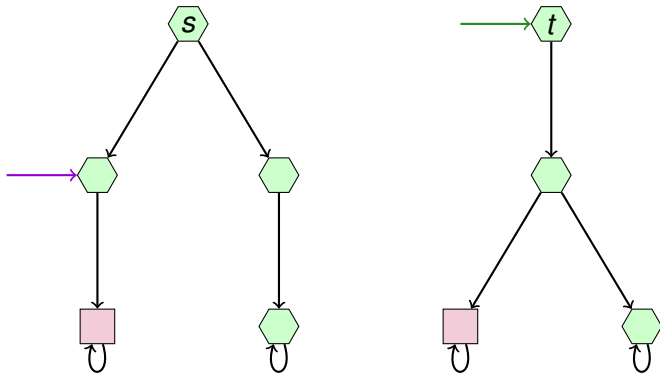
Bisimilarity



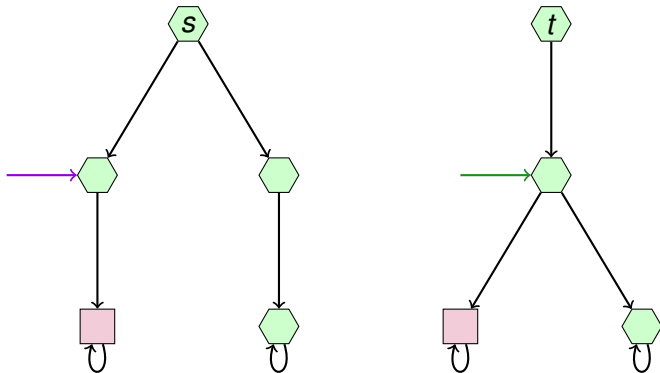
Bisimilarity



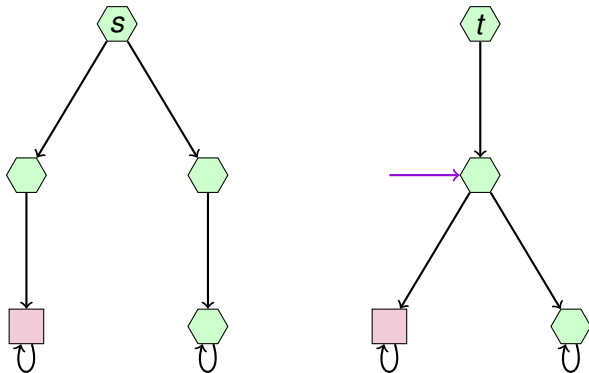
Bisimilarity



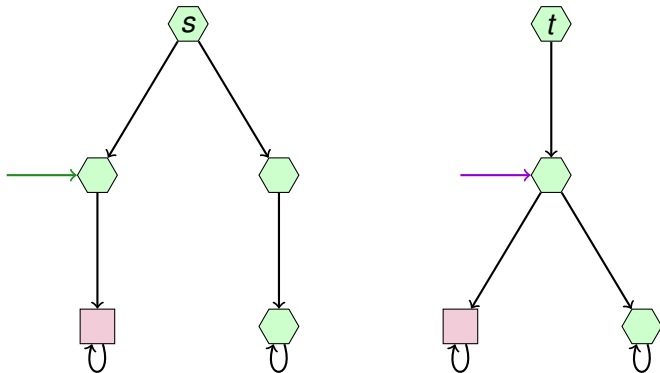
Bisimilarity



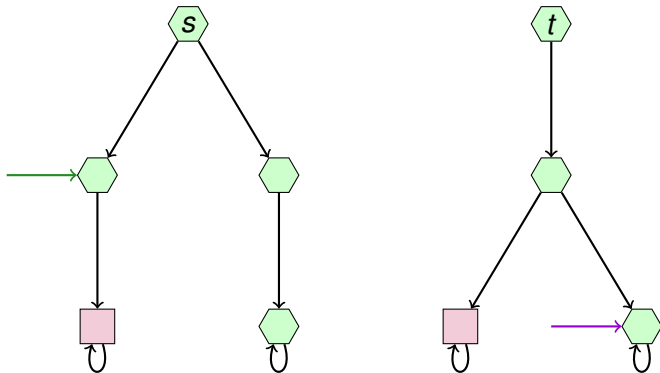
Bisimilarity



Bisimilarity



Bisimilarity



Hennessy-Milner theorem

Theorem

States of a labelled transition system are bisimilar iff they satisfy the same formulas of the Hennessy-Milner logic.



Hennessy and Milner, ICALP 1980.

Question

How do we explain that states of a labelled transition system are *not* bisimilar?

Question

How do we explain that states of a labelled transition system are *not* bisimilar?

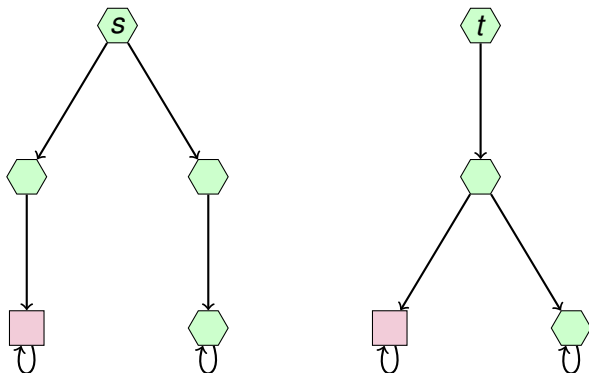
Answer

Find a formula of the Hennessy-Milner logic that is satisfied in one of the states but not the other.



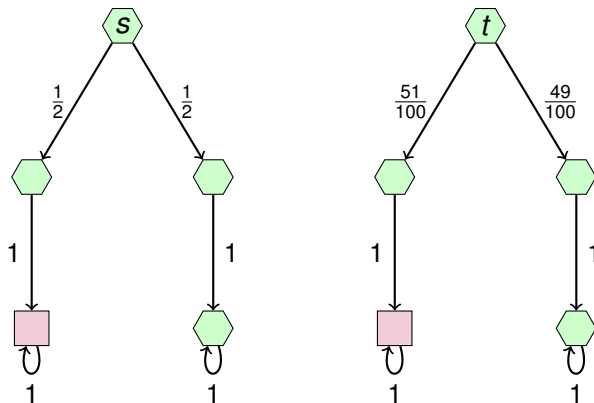
Cleveland, CAV 1990.

Bisimilarity



State t satisfies $\bigcirc((\bigcirc \text{hexagon}) \wedge (\bigcirc \text{square}))$, but state s does not.

Probabilistic bisimilarity distances



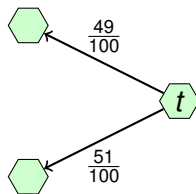
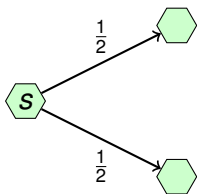
Question

How do we explain that the distance of s and t is $\frac{1}{100}$?

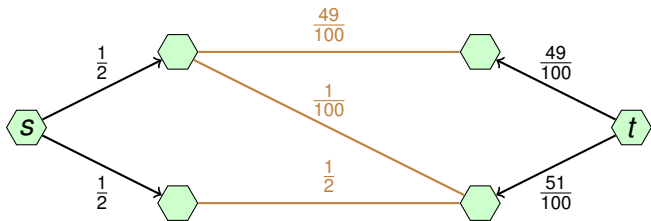
Probabilistic bisimilarity distances

- 1 Game.
- 2 Logic.

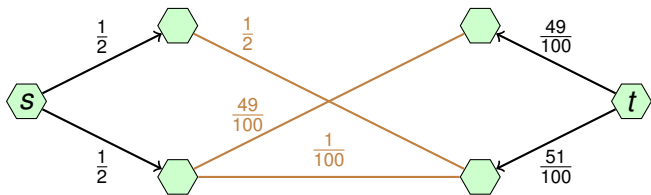
Matching probabilities



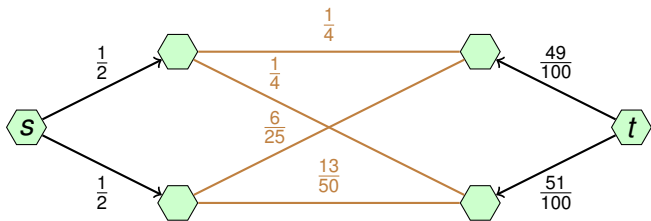
Matching probabilities



Matching probabilities

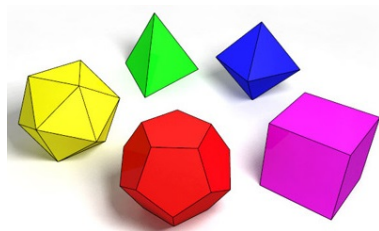


Matching probabilities



Matching probabilities = couplings

The couplings form a convex polytope



Doeblin, 1938.

A convex polytope is fully determined by its vertices

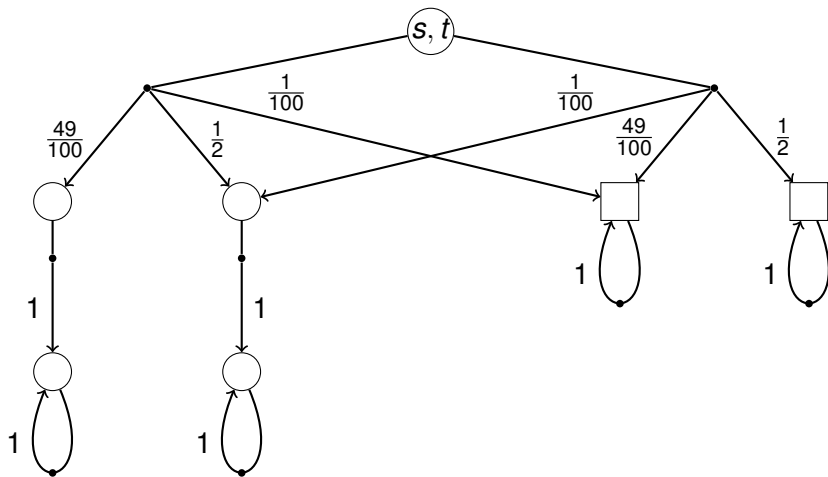
Probabilistic bisimilarity distance game

Labelled Markov chain $\langle S, \dots \rangle$

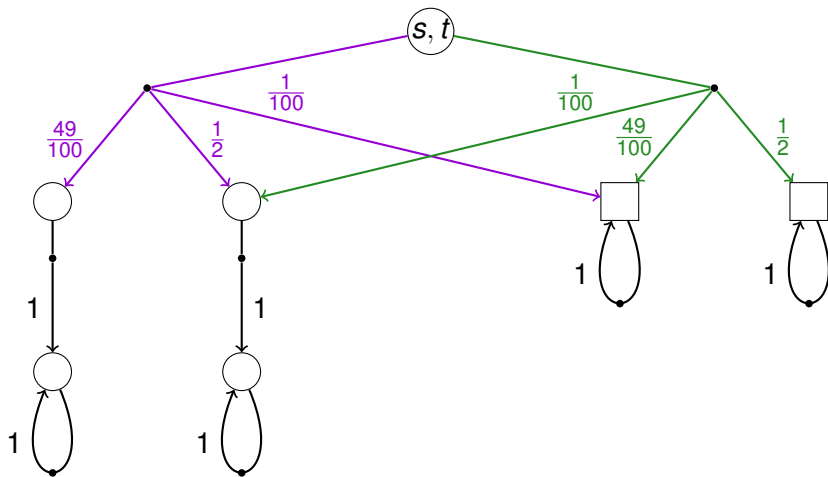


Markov decision process $\langle S \times S, \dots \rangle$

Markov decision process



Markov decision process



Probabilistic bisimilarity distances game

Theorem

$$d(s, t) = \min_{\text{strategy}} \text{probability of reaching } \square$$



Chen, Worrell, and vB, FoSSaCS 2012.

Question

How do we explain the probabilistic bisimilarity distances?

Question

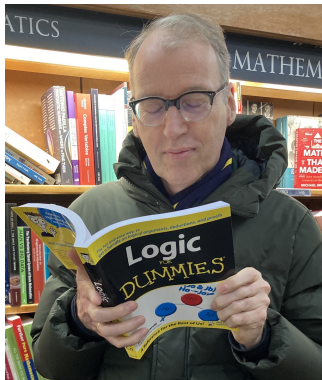
How do we explain the probabilistic bisimilarity distances?

Answer

Find a strategy that minimizes the probability of reaching \square .

Probabilistic bisimilarity distances

- 1 Game.
- 2 Logic.



A quantitative Hennessy-Milner theorem

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

The logic \mathcal{L} is defined by

$$f ::= a \mid \neg f \mid \bigcirc f \mid f \ominus q \mid f \wedge f$$

where a is a label and $q \in \mathbb{Q} \cap [0, 1]$.

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

$$\llbracket a \rrbracket(s) = \begin{cases} 1 & \text{if } a \text{ is label of } s \\ 0 & \text{otherwise} \end{cases}$$

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

$$\llbracket a \rrbracket(s) = \begin{cases} 1 & \text{if } a \text{ is label of } s \\ 0 & \text{otherwise} \end{cases}$$

$$\llbracket \neg f \rrbracket(s) = 1 - \llbracket f \rrbracket(s)$$

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

$$\llbracket a \rrbracket(s) = \begin{cases} 1 & \text{if } a \text{ is label of } s \\ 0 & \text{otherwise} \end{cases}$$

$$\llbracket \neg f \rrbracket(s) = 1 - \llbracket f \rrbracket(s)$$

$$\llbracket \bigcirc f \rrbracket(s) = \sum_{t \in S} P(s, t) \llbracket f \rrbracket(t)$$

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

$$\llbracket a \rrbracket(s) = \begin{cases} 1 & \text{if } a \text{ is label of } s \\ 0 & \text{otherwise} \end{cases}$$

$$\llbracket \neg f \rrbracket(s) = 1 - \llbracket f \rrbracket(s)$$

$$\llbracket \bigcirc f \rrbracket(s) = \sum_{t \in S} P(s, t) \llbracket f \rrbracket(t)$$

$$\llbracket f \ominus q \rrbracket(s) = \llbracket f \rrbracket(s) - q$$

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

$$\llbracket a \rrbracket(s) = \begin{cases} 1 & \text{if } a \text{ is label of } s \\ 0 & \text{otherwise} \end{cases}$$

$$\llbracket \neg f \rrbracket(s) = 1 - \llbracket f \rrbracket(s)$$

$$\llbracket \bigcirc f \rrbracket(s) = \sum_{t \in S} P(s, t) \llbracket f \rrbracket(t)$$

$$\llbracket f \ominus q \rrbracket(s) = \max(\llbracket f \rrbracket(s) - q, 0)$$

Theorem

For all states s, t ,

$$d(s, t) = \sup_{f \in \mathcal{L}} \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t).$$

$$\llbracket a \rrbracket(s) = \begin{cases} 1 & \text{if } a \text{ is label of } s \\ 0 & \text{otherwise} \end{cases}$$

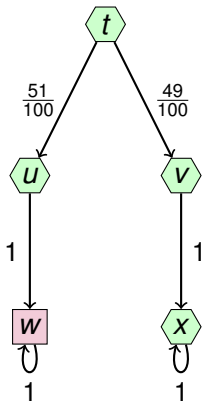
$$\llbracket \neg f \rrbracket(s) = 1 - \llbracket f \rrbracket(s)$$

$$\llbracket \bigcirc f \rrbracket(s) = \sum_{t \in S} P(s, t) \llbracket f \rrbracket(t)$$

$$\llbracket f \ominus q \rrbracket(s) = \max(\llbracket f \rrbracket(s) - q, 0)$$

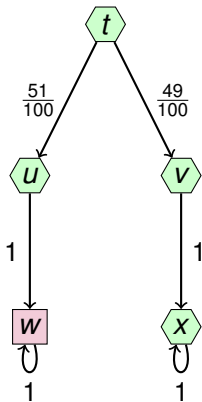
$$\llbracket f \wedge g \rrbracket(s) = \min(\llbracket f \rrbracket(s), \llbracket g \rrbracket(s))$$

A logical characterization



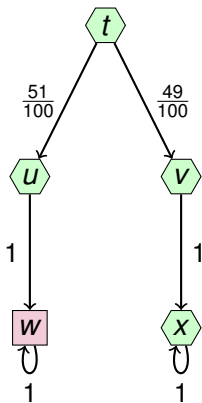
$$\llbracket \bigcirc(\text{hexagon} \wedge \bigcirc \text{hexagon}) \rrbracket (t)$$

A logical characterization



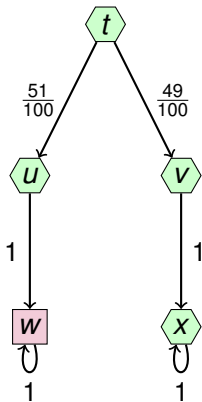
$$\begin{aligned} & \llbracket \bigcirc(\text{hexagon} \wedge \bigcirc \text{hexagon}) \rrbracket (t) \\ = & \frac{51}{100} \llbracket \text{hexagon} \wedge \bigcirc \text{hexagon} \rrbracket (u) + \frac{49}{100} \llbracket \text{hexagon} \wedge \bigcirc \text{hexagon} \rrbracket \end{aligned}$$

A logical characterization



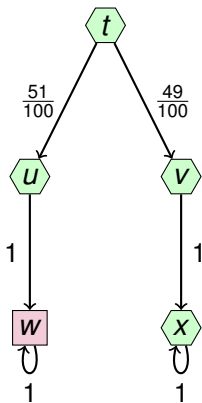
$$\begin{aligned} & \llbracket \bigcirc(\text{hexagon} \wedge \bigcirc \text{hexagon}) \rrbracket (t) \\ = & \frac{51}{100} \llbracket \text{hexagon} \wedge \bigcirc \text{hexagon} \rrbracket (u) + \frac{49}{100} \llbracket \text{hexagon} \wedge \bigcirc \text{hexagon} \rrbracket (v) \\ = & \frac{51}{100} \llbracket \bigcirc \text{hexagon} \rrbracket (u) + \frac{49}{100} \llbracket \bigcirc \text{hexagon} \rrbracket (v) \end{aligned}$$

A logical characterization



$$\begin{aligned} & \llbracket \bigcirc(\heptagon \wedge \bigcirc \heptagon) \rrbracket (t) \\ = & \frac{51}{100} \llbracket \heptagon \wedge \bigcirc \heptagon \rrbracket (u) + \frac{49}{100} \llbracket \heptagon \wedge \bigcirc \heptagon \rrbracket (v) \\ = & \frac{51}{100} \llbracket \bigcirc \heptagon \rrbracket (u) + \frac{49}{100} \llbracket \bigcirc \heptagon \rrbracket (v) \\ = & \frac{51}{100} \llbracket \heptagon \rrbracket (w) + \frac{49}{100} \llbracket \heptagon \rrbracket (x) \end{aligned}$$

A logical characterization



$$\begin{aligned} & \left[\left[\bigcirc (\bigcirc \wedge \bigcirc) \right] \right] (t) \\ &= \frac{51}{100} \left[\bigcirc \wedge \bigcirc \right] (u) + \frac{49}{100} \left[\bigcirc \wedge \bigcirc \right] (v) \\ &= \frac{51}{100} \left[\bigcirc \right] (u) + \frac{49}{100} \left[\bigcirc \right] (v) \\ &= \frac{51}{100} \left[\bigcirc \right] (w) + \frac{49}{100} \left[\bigcirc \right] (x) \\ &= \frac{49}{100} \end{aligned}$$

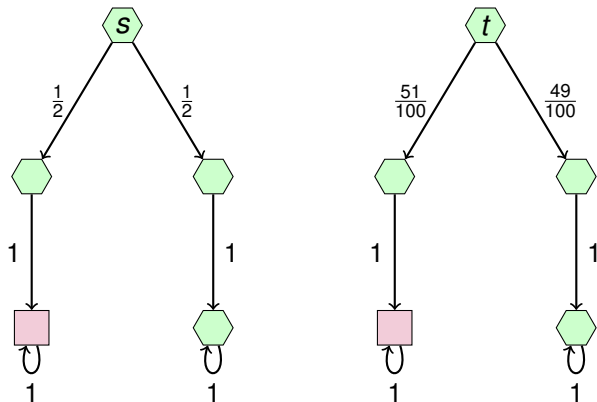
Question

How do we explain the probabilistic bisimilarity distances?

Attempt

Find a formula f of the logic \mathcal{L} with $d(s, t) = \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t)$.

Explainability



The distance of states s and t can be explained by the formula

$$\bigcirc(\text{green hexagon} \wedge \bigcirc \text{green hexagon})$$

Question

How do we explain the probabilistic bisimilarity distances?

Attempt fails

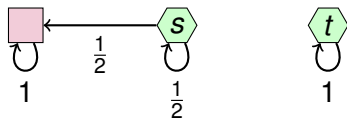
Find a formula f of the logic \mathcal{L} with $d(s, t) = \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t)$.

Question

How do we explain the probabilistic bisimilarity distances?

Attempt fails

Find a formula f of the logic \mathcal{L} with $d(s, t) = \llbracket f \rrbracket(s) - \llbracket f \rrbracket(t)$.



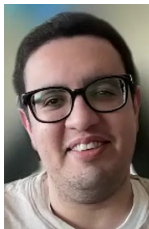
For every formula f , $\llbracket f \rrbracket(s) - \llbracket f \rrbracket(t) < d(s, t) = 1$.

Question

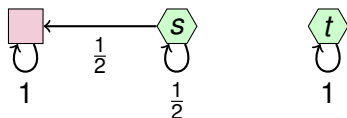
How do we explain the probabilistic bisimilarity distances?

Attempt succeeds







Find a sequence $(f_n)_n$ of formulas of the logic \mathcal{L} with
 $d(s, t) = \lim_n \llbracket f_n \rrbracket(s) - \llbracket f_n \rrbracket(t)$.

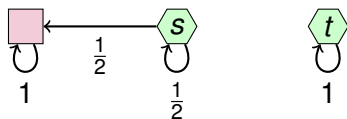


Rady and vB, FoSSaCS 2023.



The distance of states s and t can be explained by the sequence of formulas

- false
- \bigcirc 
- \bigcirc ( \vee \bigcirc )
- \bigcirc ( \vee \bigcirc ( \vee \bigcirc ))
- ...



The distance of states t and s can be explained by the sequence of formulas

- false
- $(\bigcirc \text{Hexagon}) \ominus \frac{1}{2}$
- $(\bigcirc (\text{Hexagon} \wedge ((\bigcirc \text{Hexagon}) \ominus \frac{1}{2} \oplus \frac{1}{2}))) \ominus \frac{1}{4}$
- $(\bigcirc (\text{Hexagon} \wedge ((\bigcirc (\text{Hexagon} \wedge ((\bigcirc \text{Hexagon}) \ominus \frac{1}{2} \oplus \frac{1}{2})))) \ominus \frac{1}{4} \oplus \frac{1}{4}))) \ominus \frac{1}{8}$
- ...

Objective

Construct a sequence $(f_{st}^n)_n$ of formulas of the logic \mathcal{L} with $d(s, t) = \lim_n \llbracket f_{st}^n \rrbracket(s) - \llbracket f_{st}^n \rrbracket(t)$.

Some technical details

Objective

Construct a sequence $(f_{st}^n)_n$ of formulas of the logic \mathcal{L} with $d(s, t) = \lim_n \llbracket f_{st}^n \rrbracket(s) - \llbracket f_{st}^n \rrbracket(t)$.

Kleene fixed point theorem

$$d = \lim_n d_n.$$



Some technical details

Objective

Construct a sequence $(f_{st}^n)_n$ of formulas of the logic \mathcal{L} with $d(s, t) = \lim_n \llbracket f_{st}^n \rrbracket(s) - \llbracket f_{st}^n \rrbracket(t)$.

Kleene fixed point theorem

$d = \lim_n d_n$.



Objective

Construct a formula f_{st}^n of the logic \mathcal{L} with $d_n(s, t) = \llbracket f_{st}^n \rrbracket(s) - \llbracket f_{st}^n \rrbracket(t)$.

Some technical details

Objective

Construct a formula f_{st}^n of the logic \mathcal{L} with
 $d_n(s, t) = \llbracket f_{st}^n \rrbracket(s) - \llbracket f_{st}^n \rrbracket(t)$.

Kantorovich-Rubinstein duality theorem

There exists $F_{st}^n : S \rightarrow [0, 1]$ such that

$$d_n(s, t) = \left(\sum_{u \in S} P(s, u) F_{st}^n(u) \right) - \left(\sum_{u \in S} P(t, u) F_{st}^n(u) \right).$$



Some technical details

The formula g_{stuv}^n is defined by

$$\begin{aligned} & \text{false} \oplus F_{st}^n(u) && \text{if } F_{st}^n(u) = F_{st}^n(v) \\ (f_{uv}^n \ominus (d_n(u, v) - (F_{st}^n(u) - F_{st}^n(v)))) \oplus F_{st}^n(v) && \text{if } F_{st}^n(u) > F_{st}^n(v) \\ (f_{vu}^n \ominus (d_n(u, v) - (F_{st}^n(v) - F_{st}^n(u)))) \oplus F_{st}^n(u) && \text{otherwise.} \end{aligned}$$

Stone-Weierstrass approximation theorem

$$f_{st}^n = \left[\bigwedge_{u \in S} \bigvee_{v \in S} g_{stuv}^n \right]$$



Some technical details

The formula f_{st}^n

Some technical details

The formula f_{st}^n

If s and t have a different label then $f_{st}^n = \text{label of } s$.

Some technical details

The formula f_{st}^n

If s and t have a different label then $f_{st}^n = \text{label of } s$.

Otherwise,

$$f_{st}^n = \left(\bigcirc \bigwedge_{u \in S} \bigvee_{v \in S} g_{stuv}^n \right) \ominus \left(\sum_{u \in S} P(t, u) F_{st}^n(u) \right).$$

- How to extend the logic \mathcal{L} and, for states s and t , find a formula $f_{st} \in \mathcal{L}$ such that

$$d(s, t) = \llbracket f_{st} \rrbracket(s) - \llbracket f_{st} \rrbracket(t)?$$

- How to extend the logic \mathcal{L} and, for state s , find a formula $f_s \in \mathcal{L}$ such that for all states t

$$d(s, t) = \llbracket f_s \rrbracket(s) - \llbracket f_s \rrbracket(t)?$$

- How to extend the logic \mathcal{L} and, for state s , find a formula $f_s \in \mathcal{L}$ such that for all states t

$$d(s, t) = \llbracket f_s \rrbracket(t)?$$

Slide 3

- Photo of Smolka:
www.concur2016.ulaval.ca/fr/program/speakers_html/
- Alessandro Giacalone, Chi-Chang Jou, and Scott Smolka. Algebraic reasoning for probabilistic concurrent systems. In *Proceedings of the IFIP WG 2.2/2.3 Working Conference on Programming Concepts and Methods*, pages 443–458, Sea of Gallilee, Israel, April 1990. North-Holland.

Slide 16

- Photo of Hennessy: www.scss.tcd.ie/Matthew.Hennessy/
- Photo of Milner: www.cl.cam.ac.uk/archive/rm135/
- Matthew Hennessy and Robin Milner. On observing nondeterminism and concurrency. In Jaco de Bakker and Jan van Leeuwen, editors, *Proceedings of the 7th Colloquium on Automata, Languages and Programming*, volume 85 of *Lecture Notes in Computer Science*, pages 299–309, Noordwijkerhout, The Netherlands, July 1980. Springer-Verlag.

Slide 17

- Photo of Cleaveland:
cmns.umd.edu/people/rance-cleaveland/
- Rance Cleaveland. On automatically distinguishing inequivalent processes. In Edmund Clarke and Robert Kurshan, editors, *Proceedings of a DIMACS Workshop on Computer Aided Verification*, volume 3 of *DIMACS Series in Discrete Mathematics and Theoretical Computer Science*, pages 463–476, New Brunswick, NJ, USA, June 1990. DIMACS/AMS.

Slide 25

- Photo of Doeblin:
mathshistory.st-andrews.ac.uk/Biographies/Doeblin/
- Wolfgang Doeblin. Exposé de la théorie des chaînes simples constantes de Markoff à un nombre fini d'états. *Revue Mathématiques de l'Union Interbalkanique*, 2: 77-105, 1938.
- Image of polytopes:
www.greatlittleminds.com/pages/maths/3d-platonic-solids.html

Slide 29

- Photo of Worrell:
www.cs.ox.ac.uk/people/james.worrell/home.html
- Di Chen, Franck van Breugel, and James Worrell. On the complexity of computing probabilistic bisimilarity. In Lars Birkedal, editor, *Proceedings of the 15th International Conference on Foundations of Software Science and Computational Structures*, volume 7213 of *Lecture Notes in Computer Science*, pages 437–451, Tallinn, Estonia, March/April 2012. Springer-Verlag.

Slide 38

- Amgad Rady and Franck van Breugel. Explainability of probabilistic bisimilarity distances for labelled Markov chains. In Orna Kupferman and Pawel Sobocinski, editors, *Proceedings of the 26th International Conference on Foundations of Software Science and Computational Structures*, volume 13992 of *Lecture Notes in Computer Science*, pages 285–307, Paris, France, April 2023. Springer-Verlag.

Slide 41

- Photo of Kleene:
www.computerhope.com/people/stephen_kleene.htm
- Stephen Kleene. *Introduction to Metamathematics*.

Slide 42

- Photo of Kantorovich:
armstrongeconomics.wordpress.com/research/economic-thought/by-author/kantorovich-leonid/
- Leonid Kantorovich and Gennadi Rubinstein. On the space of completely additive functions (in Russian). *Vestnik Leningradskogo Universiteta*, 3(2):52–59, 1958.

Slide 43

- Photo of Stone:
wildpeaches.xyz/blog/curve-fitting-with-julia/
- Photo of Weierstrass:
wildpeaches.xyz/blog/curve-fitting-with-julia/
- Marshall Stone. Applications of the theory of Boolean rings to general topology. *Transactions of the American Mathematical Society*, 41(3): 375–481, 1937.