# FOPPS Lectures: Probabilistic Bisimulation Metrics and Their Applications to Representation Learning <br> Lecture 1: The logical characterization of bisimulation 

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## Overview

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- Labelled Markov processes: probabilistic transition systems with continuous state spaces.


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- Logical characterization.


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- Quantitative equational logic [LICS 2016, 2017, 2018, 2021, CALCO 2021]
- Diffusion and continuous-time processes [MFPS 2019, 2020]


## Collaborators

Giorgio Bacci, Philippe Chaput, Linan Chen, Florence Clerc, Vincent Danos, Josée Desharnais, Abbas Edalat, Norm Ferns, Nathanaël Fijalkow, Robert Furber, Vineet Gupta, Radha Jagadeesan, Bartek Klin, Dexter Kozen, Kim Larsen, François Laviolette, Radu Mardare, Gordon Plotkin and Doina Precup.

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- a set of labels or actions, $L$ or $\mathcal{A}$ and
- a transition relation $\subseteq S \times \mathcal{A} \times S$, usually written

$$
\rightarrow_{a} \subseteq S \times S
$$

The transitions could be indeterminate (nondeterministic).

## Markov Chains

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- A discrete-time Markov chain is a finite set $S$ (the state space) together with a transition probability function $T: S \times S \rightarrow[0,1]$.
- The key property is that the transition probability from $s$ to $s^{\prime}$ only depends on $s$ and $s^{\prime}$ and not on the past history of how it got there. This is what allows the probabilistic data to be given as a single matrix $T$.


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\left(S, \mathrm{~L}, \forall a \in \mathrm{~L} T_{a}: S \times S \rightarrow[0,1]\right)
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- The model is reactive: All probabilistic data is internal - no probabilities associated with environment behaviour.


## Examples of PTSs



## Bisimulation for PTS: Larsen and Skou

- Consider

$P_{1}$

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## Bisimulation for PTS: Larsen and Skou

- Consider

- Should $s_{0}$ and $t_{0}$ be bisimilar?
- Yes, but we need to add the probabilities.


## The Official Definition

- Let $\mathcal{S}=\left(S, \mathrm{~L}, T_{a}\right)$ be a PTS. An equivalence relation $R$ on $S$ is a bisimulation if whenever $s R s^{\prime}$, with $s, s^{\prime} \in S$, we have that for all $a \in \mathcal{A}$ and every $R$-equivalence class, $A, T_{a}(s, A)=T_{a}\left(s^{\prime}, A\right)$.


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- The notation $T_{a}(s, A)$ means "the probability of starting from $s$ and jumping to a state in the set $A$."
- Two states are bisimilar if there is some bisimulation relation $R$ relating them.


## What are labelled Markov processes?

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- Labelled Markov processes are probabilistic versions of labelled transition systems. Labelled transition systems where the final state is governed by a probability distribution - no other indeterminacy.
- All probabilistic data is internal - no probabilities associated with environment behaviour.
- We observe the interactions - not the internal states.
- In general, the state space of a labelled Markov process may be a continuum.


## The Need for Measure Theory

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- Basic fact: There are subsets of $\mathbf{R}$ for which no sensible notion of size can be defined.
- More precisely, there is no non-trivial translation-invariant measure defined on all the subsets of the reals.


## Stochastic Kernels

- A stochastic kernel (Markov kernel) is a function $h: S \times \Sigma \rightarrow[0,1]$ with (a) $h(s, \cdot): \Sigma \rightarrow[0,1]$ a (sub)probability measure and (b) $h(\cdot, A): S \rightarrow[0,1]$ a measurable function.


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- Though apparantly asymmetric, these are the stochastic analogues of binary relations
- and the uncountable generalization of a matrix.
- They are the Kleisli arrows of a monad: the Giry monad.


## Formal Definition of LMPs

- An LMP is a tuple $\left(S, \Sigma, \mathrm{~L}, \forall \alpha \in \mathrm{~L} . \tau_{\alpha}\right)$ where $\tau_{\alpha}: S \times \Sigma \rightarrow[0,1]$ is a transition probability function such that


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- $\forall s: S . \lambda A: \Sigma . \tau_{\alpha}(s, A)$ is a subprobability measure and
$\forall A: \Sigma . \lambda s: S . \tau_{\alpha}(s, A)$ is a measurable function.


## Probabilistic Bisimulation

## Desharnais et al.

Let $\mathcal{S}=(S, i, \Sigma, \tau)$ be a labelled Markov process. An equivalence relation $R$ on $S$ is a bisimulation if whenever $s R s^{\prime}$, with $s, s^{\prime} \in S$, we have that for all $a \in \mathcal{A}$ and every $R$-closed measurable set $A \in \Sigma$, $\tau_{a}(s, A)=\tau_{a}\left(s^{\prime}, A\right)$.

Two states are bisimilar if they are related by a bisimulation relation.

## A game for bisimulation

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- Duplicator responds by saying that $C$ is not bisimulation-closed and that exhibits $x^{\prime} \in C$ and $y^{\prime} \notin C$ and claims that $x^{\prime}, y^{\prime}$ are bisimilar.


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- A player loses when he or she cannot make a move. Note that if $C$ is all of the state space, duplicator loses. Duplicator wins if she can play forever.
- We prove that $x$ is bisimilar to $y$ iff Duplicator has a winning strategy starting from $(x, y)$.


## Logical Characterization

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- Two systems are bisimilar iff they obey the same formulas of $\mathcal{L}$. [DEP 1998 LICS, I and C 2002]


## That cannot be right?



Two processes that cannot be distinguished without negation. The formula that distinguishes them is $\langle a\rangle(\neg\langle b\rangle \top)$.

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- If $p+q<r$ or $p+q>r$ we can easily distinguish them.
- If $p+q=r$ and $p>0$ then $q<r$ so $\langle a\rangle r\langle b\rangle 1 \top$ distinguishes them.


## Proof idea

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- Use Dynkin's lemma to show that we get a well defined measure on the $\sigma$-algebra generated by such sets and the above equality holds.
- Use special properties of analytic spaces to show that this $\sigma$-algebra is the same as the original $\sigma$-algebra.


## Simulation

Let $\mathcal{S}=(S, \Sigma, \tau)$ be a labelled Markov process. A preorder $R$ on $S$ is a simulation if whenever $s R s^{\prime}$, we have that for all $a \in \mathcal{A}$ and every $R$-closed measurable set $A \in \Sigma, \tau_{a}(s, A) \leq \tau_{a}\left(s^{\prime}, A\right)$. We say $s$ is simulated by $s^{\prime}$ if $s R s^{\prime}$ for some simulation relation $R$.

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- One can show that if $s$ simulates $s^{\prime}$ then $s$ satisfies all the formulas of $\mathcal{L}$ that $s^{\prime}$ satisfies.
- What about the converse?


## Counter example!

In the following picture, $t$ satisfies all formulas of $\mathcal{L}$ that $s$ satisfies but $t$ does not simulate $s$.


All transitions from $s$ and $t$ are labelled by $a$.

## Counter example (contd.)

- A formula of $\mathcal{L}$ that is satisfied by $t$ but not by $s$.

$$
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## Counter example (contd.)

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$$
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- A formula with disjunction that is satisfied by $s$ but not by $t$ :

$$
\langle a\rangle_{\frac{3}{4}}\left(\langle a\rangle_{0} \mathbf{T} \vee\langle b\rangle_{0} \mathbf{T}\right)
$$

## A logical characterization for simulation

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- With this logic we have:

An LMP $s_{1}$ simulates $s_{2}$ if and only if for every formula $\phi$ of $\mathcal{L}_{V}$ we have

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- The original proof uses domain theory and approximation.
- New development (2017 ICALP) we can prove logical characterization for simulation and bisimulation in almost the same way.


## Digression on Analytic Spaces

- An analytic set $A$ is the image of a Polish space $X$ (or a Borel subset of $X$ ) under a continuous (or measurable) function $f: X$ $\rightarrow Y$, where $Y$ is Polish. If $(S, \Sigma)$ is a measurable space where $S$ is an analytic set in some ambient topological space and $\Sigma$ is the Borel $\sigma$-algebra on $S$.


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- Analytic sets do not form a $\sigma$-algebra but they are in the completion of the Borel algebra under any measure. [Universally measurable.]


## Amazing Facts about Analytic Spaces

- Given $A$ an analytic space and $\sim$ an equivalence relation such that there is a countable family of real-valued measurable functions $f_{i}: S \rightarrow \mathbf{R}$ such that

$$
\forall s, s^{\prime} \in S . s \sim s^{\prime} \Longleftrightarrow \forall f_{i} \cdot f_{i}(s)=f_{i}\left(s^{\prime}\right)
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then the quotient space ( $Q, \Omega$ ) - where $Q=S / \sim$ and $\Omega$ is the finest $\sigma$-algebra making the canonical surjection $q: S \rightarrow Q$ measurable - is also analytic.

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then the quotient space $(Q, \Omega)$ - where $Q=S / \sim$ and $\Omega$ is the finest $\sigma$-algebra making the canonical surjection $q: S \rightarrow Q$ measurable - is also analytic.

- If an analytic space $(S, \Sigma)$ has a sub- $\sigma$-algebra $\Sigma_{0}$ of $\Sigma$ which separates points and is countably generated then $\Sigma_{0}$ is $\Sigma$ ! The Unique Structure Theorem (UST).


## Some more measure theory

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- $\lambda-\pi$ theorem: If $\Pi$ is a $\pi$-system and $\Lambda$ is a $\lambda$-system and $\Pi \subset \Lambda$ then $\sigma(\Pi) \subset \Lambda$.


## Some more measure theory

- A $\pi$-system is a family of sets closed under finite intersections.
- A $\lambda$-system is a family of sets closed under complements and countable disjoint unions.
- $\lambda-\pi$ theorem: If $\Pi$ is a $\pi$-system and $\Lambda$ is a $\lambda$-system and $\Pi \subset \Lambda$ then $\sigma(\Pi) \subset \Lambda$.
- Corollary: If two measures agree on the sets of a $\pi$-system then they agree on the generated $\sigma$-algebra.


## Bisimulation proof I

- Given $\left(S, \Sigma, \tau_{a}\right)$ an LMP, we define $x \simeq y$ if $x$ and $y$ obey exactly the same formulas of $\mathcal{L}_{0}$.


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- We claim that $\simeq$ is a bisimulation relation.


## Bisimulation proof I

- Given $\left(S, \Sigma, \tau_{a}\right)$ an LMP, we define $x \simeq y$ if $x$ and $y$ obey exactly the same formulas of $\mathcal{L}_{0}$.
- We claim that $\simeq$ is a bisimulation relation.
- Suppose that $x, y \in S$ and for some $a$ and some $\simeq$-closed set $C$, $\tau_{a}(x, C) \neq \tau_{a}(y, C)$.


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- We need to show there is a formula on which $x, y$ disagree.
- Let $\delta=\tau_{a}(x, \cdot)$ and $\gamma=\tau_{a}(y, \cdot)$.
- If $\delta(S)>\gamma(S)$ then choose rational $q$ such that $\delta(S)>q>\gamma(S)$. Now $x \neq\langle a\rangle_{q} \top$ and $y \not \vDash\langle a\rangle_{q} \top$.


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- Suppose $\delta(\llbracket \phi \rrbracket)>\gamma(\llbracket \phi \rrbracket)$ choose $q$ rational in between and we have
- $x \models\langle a\rangle_{q} \phi$ and $y \not \vDash\langle a\rangle_{q} \phi$.


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- $x \preceq y$ iff Duplicator has a winning strategy starting from $x, y$.


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- With these in place the proof of the logical characterization of simulation follows the same pattern.


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- However, if you require the transition functions to be continuous instead of measurable then logical characterization is restored.
- For simulation as well as bisimulation.
- We heavily use topological ideas in this proof.


## What is next?

## Metrics!

