Continuous Markovian Logic -From Complete Axiomatization to the Metric Space of Formulas

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Motivation

Complex systems are often modelled as stochastic processes

biological and ecological systems, physical systems, social systems, financial systems

- to encapsulate a lack of knowledge or inherent non-determinism, the information about real systems is based on approximations
- to model hybrid real-time and discrete-time interacting components, these systems are frequently studied in interaction with discrete controllers, or with interactive environments having continuous behavior
- to abstract complex continuous-time and continuous-space systems the real systems are reactive systems with continuous behaviour (in space and time)

Motivation

In this context, the <u>stochastic/probabilistic bisimulation</u> is a too strict concept

- the interest is to understand not whether two systems have identical behaviours, but when two systems have similar behaviours (up to an observational error)
- bisimulation => pseudometric that measures how similar two systems are from the point of view of their behaviours
- Model checking => property evaluation: instead of deciding whether "P⊨f", one measures "P⊨f" giving an observational error (granularity).



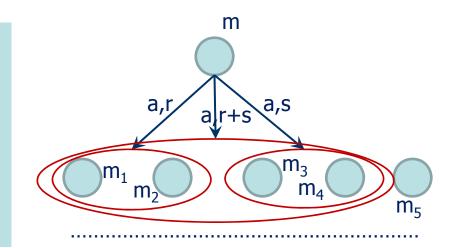
Overview

- We focus on continuous-time and continuous-space Markov processes (CMPs)
- We introduce the Continuous Markovian Logic (CML), a multimodal logic that <u>characterizes the stochastic bisimulation</u>. We provide complete Hilbert-style axiomatizations for CMLs and prove the <u>finite model property</u>
- We define an approximation of the satisfiability relation that induces:
 - a <u>bisimulation pseudodistance</u> on CMPs
 - a <u>syntactic pseudodistance</u> on logical formulas
- The pseudodistances are used to state the Strong Robustness Theorem and the finite model construction to approximate it in the form of the Weak Robustness Theorem
- The complete axiomatization allows <u>the transfer of topological properties</u> between the space of CMPs and the space of logical formulas.

Labelled Markov kernel

- A tuple $\mathcal{M} = (M, \Sigma, A, \{R_a | a \in A\})$ where
- (M,Σ) is an analytic set (measurable space)
- Σ is the Borel-algebra generated by the topology
- A is a set of labels
- for each $a \in A$, $R_a: M \times \Sigma \rightarrow [0,1]$ is such that $R_a(m,-)$ - (sub-)probability measure on (M,Σ) $R_a(-,S)$ - measurable function

(P. Panangaden, Labelled Markov Processes, 2009.)



Equivalent definition:

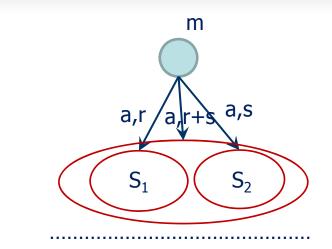
A tuple $\mathcal{M}=(\mathsf{M},\Sigma,\theta)$ where $\theta \in \llbracket \mathsf{M} \to \Pi(\mathsf{M},\Sigma) \rrbracket^{\mathsf{A}}$ $\theta_{\mathsf{a}}: \mathsf{M} \to \Pi(\mathsf{M},\Sigma), \quad \theta_{\mathsf{a}}(\mathsf{m}) \in \Pi(\mathsf{M},\Sigma), \quad \theta_{\mathsf{a}}(\mathsf{m})(\mathsf{S}) \in [0,1]$

$$\label{eq:product} \begin{split} \Pi(M,\Sigma) \text{ is a measurable space with the sigma-algebra generated, for arbitrary $S\in\Sigma$ and $r\in\mathbb{Q}$, by $\{\mu\in\Pi(M,\Sigma)\mid \mu(S)\leq r\}$. \end{split}$$

(E. Doberkat, Stochastic Relations, 2007.)

Continuous (Labelled) Markov kernel

- A tuple $\mathcal{M} = (M, \Sigma, A, \{R_a | a \in A\})$ where
- (M,Σ) is an analytic set (measurable space)
- A is a set of labels
- for each $a \in A$, $R_a: M \times \Sigma \rightarrow [0, \infty)$ is such that $R_a(m, -) - a$ measure on (M, Σ) $R_a(-, S) - a$ measurable function



- R_a(m,S)=r ∈[0,+∞) the rate of an exponentially distributed random variable that characterizes the time of a-transitions from m to arbitrary elements of S.
- the probability of the *transition within time t* is given by the cumulative distribution function P(t) = 1

 $P(t) = 1 - e^{-rt}$

Equivalent definition:

A tuple $\mathcal{M}=(M,\Sigma,\theta)$, where $\theta \in [[M \to \Delta(M,\Sigma)]]^A$

 $\theta_a : M \to \Delta(M, \Sigma), \quad \theta_a(m) {\in} \Delta(M, \Sigma), \quad \theta_a(m)(S) {\in} [0, +\infty)$

Continuous Markov process $(\mathcal{M}, m), m \in M$

Stochastic/Probabilistic Bisimulation

Given a **probabilistic/stochastic (Markovian) system** $\mathcal{M}=(M,\Sigma,\theta)$, a bisimulation relation is an equivalence relation $\sim \subseteq M \times M$ such that whenever $m_1 \sim m_2$, for arbitrary $S \in \Sigma(\sim)$ and $a \in A$

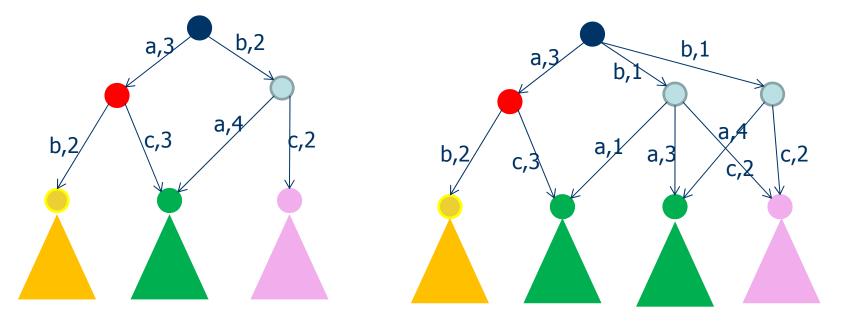
• If $m_1 \xrightarrow{a,p} S$, then $m_2 \xrightarrow{a,p} S$ and

 $\theta_a(m)(S) = \theta_a(m')(S)$

• If $m_2 a, p S$, then $m_1 a, p S$.

K. G. Larsen and A. Skou. *Bisimulation through probabilistic testing*, I&C 1991

P. Panangaden , Labelled Markov Processes, 2009.

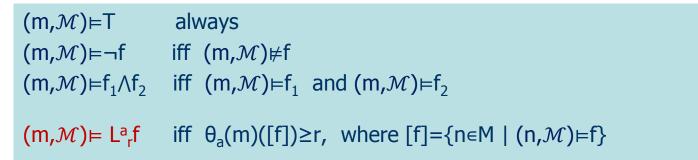


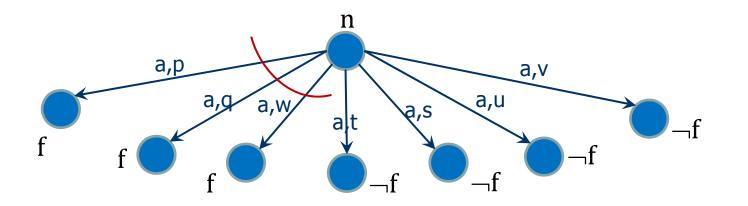
Continuous Markovian Logic

Syntax: CML(A)

 $f:= T | \neg f | f_1 \land f_2 | L^a_r f \qquad r \in \mathbb{Q}_+ a \in A$

Semantics: Let (m, \mathcal{M}) be an arbitrary CMP with $\mathcal{M} = (M, \Sigma, \theta)$.





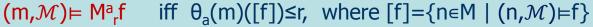
Continuous Markovian Logic

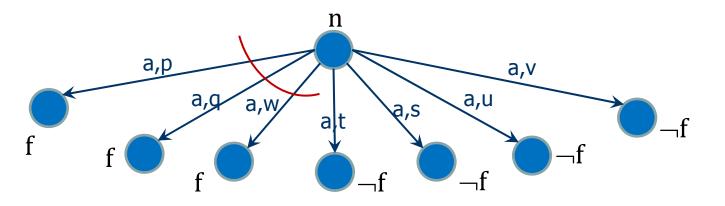
Syntax: CML+(A)

 $f:= T | \neg f | f_1 \land f_2 | L^a_r f | M^a_r f \qquad r \in \mathbb{Q}_+ a \in A$

Semantics: Let (m, \mathcal{M}) be an arbitrary CMP with $\mathcal{M} = (M, \Sigma, \theta)$.

 $\begin{array}{ll} (\mathsf{m},\mathcal{M}) \vDash \mathsf{T} & \text{always} \\ (\mathsf{m},\mathcal{M}) \vDash \neg \mathsf{f} & \text{iff } (\mathsf{m},\mathcal{M}) \nvDash \mathsf{f} \\ (\mathsf{m},\mathcal{M}) \vDash \mathsf{f}_1 \land \mathsf{f}_2 & \text{iff } (\mathsf{m},\mathcal{M}) \vDash \mathsf{f}_1 \text{ and } (\mathsf{m},\mathcal{M}) \vDash \mathsf{f}_2 \\ (\mathsf{m},\mathcal{M}) \vDash \mathsf{L}^{\mathsf{a}}_{\mathsf{r}} \mathsf{f} & \text{iff } \theta_{\mathsf{a}}(\mathsf{m})([\mathsf{f}]) \ge \mathsf{r} \end{array}$





Continuous Markovian Logic

Syntax: CML(A) & CML⁺(A)

 $f:= T | \neg f | f_1 \land f_2 | L^a_r f | M^a_r f \qquad r \in \mathbb{Q}_+ a \in A$

Semantics: Let (m, \mathcal{M}) be an arbitrary CMP with $\mathcal{M} = (M, \Sigma, \theta)$.

 $\begin{array}{ll} (m,\mathcal{M}) \vDash \mathsf{T} & \text{always} \\ (m,\mathcal{M}) \vDash \neg \mathsf{f} & \text{iff } (m,\mathcal{M}) \nvDash \mathsf{f} \\ (m,\mathcal{M}) \vDash \mathsf{f}_1 \land \mathsf{f}_2 & \text{iff } (m,\mathcal{M}) \vDash \mathsf{f}_1 \text{ and } (m,\mathcal{M}) \vDash \mathsf{f}_2 \\ (m,\mathcal{M}) \vDash \mathsf{L}^a{}_r \mathsf{f} & \text{iff } \theta_a(m)([\mathsf{f}]) \ge \mathsf{r} \\ (m,\mathcal{M}) \vDash \mathsf{M}^a{}_r \mathsf{f} & \text{iff } \theta_a(m)([\mathsf{f}]) \le \mathsf{r}, \text{ where } [\mathsf{f}] = \{\mathsf{n} \in \mathsf{M} \mid (\mathsf{n},\mathcal{M}) \vDash \mathsf{f}\} \end{array}$

<u>Theorem</u>: For arbitrary continuous Markov processes (m, \mathcal{M}) and (n, \mathcal{H}) , the following assertions are equivalent

(i) $(m,\mathcal{M}) \sim (n,\mathcal{H})$,

(ii) $\forall f \in CML(A), (m, \mathcal{M}) \models f \text{ iff } (n, \mathcal{H}) \models f$,

(iii) $\forall f \in CML^+(A), (m, \mathcal{M}) \models f \text{ iff } (n, \mathcal{H}) \models f.$

(P. Panangaden, Labelled Markov Processes, 2009.)

Modal Probabilistic Logic versus Continuous Markovian Logic

 $f := T \mid \neg f \mid f_1 \land f_2 \mid L^a_r f \mid M^a_r f$

MPL(A) for LMPs

 $\begin{array}{l} \mathcal{M} = (\mathsf{M}, \Sigma, \theta), \ \theta \in \llbracket \mathsf{M} \to \Pi(\mathsf{M}, \Sigma) \rrbracket^{\mathsf{A}} \\ S \in \Sigma, \ \theta_{\mathsf{a}}(\mathsf{m})(\mathsf{S}) \in [0, 1] \end{array}$

 $\vdash \mathsf{M}^{\mathsf{a}}_{r} \mathsf{f} \leftrightarrow \mathsf{L}^{\mathsf{a}}_{1\text{-}r} \neg \mathsf{f}$

 $\vdash L^{a}_{r}f \leftrightarrow \neg L^{a}_{s}\neg f, r+s>1$

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\vdash [If a \text{ is active}] \rightarrow L^{a}_{r}T\vdash L^{a}_{r}f \rightarrow L^{a}_{r}T
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For a fixed $q \in \mathbb{N}$ the set $\{p/q \in [0,1] \mid p \in \mathbb{N}\}$ is finite

CML(A) for CMPs

a∈A

 $\mathcal{M}=(\mathsf{M},\Sigma,\theta), \ \theta\in\llbracket\mathsf{M}\to\Delta(\mathsf{M},\Sigma)\rrbracket^{\mathsf{A}}$ $\mathsf{S}\in\Sigma, \ \theta_{\mathsf{a}}(\mathsf{m})(\mathsf{S})\in\llbracket0,+\infty)$

M^a_rf and L^a_sf are independent operators

$$- L^{a}_{s+r} f \rightarrow \neg M^{a}_{r} f , s>0$$

- M^{a}_{s+r} f \rightarrow \neg L^{a}_{r} f , s>0

 $\begin{array}{l} \vdash \neg L^{a}_{r}f \rightarrow M^{a}_{r}f \\ \vdash \neg M^{a}_{r}f \rightarrow L^{a}_{r}f \end{array}$

For a fixed $q \in \mathbb{N}$ the set $\{p/q \in [0, +\infty) \mid p \in \mathbb{N}\}$ is not finite

K.G. Larsen, A. Skou. *Bisimulation through probabilistic testing*, 1991.
R. Fagin, J.Y. Halpern, Reasoning about Knowledge and Probability, 1994
A. Heifetz, P. Mongin, Probability Logic for Type Spaces, 2001
C. Zhou, *A complete deductive system for probability logic with application to Harsanyi type spaces*, 2007.

Axiomatic Systems

CML(A)

 $\begin{array}{l} (A1) \vdash L^{a}_{0}f \\ (A2) \vdash L^{a}_{r+s}f \rightarrow L^{a}_{r}f \\ (A3) \vdash L^{a}_{r}(f \land g) \land L^{a}_{s}(f \land \neg g) \rightarrow L^{a}_{r+s}f \\ (A4) \vdash \neg L^{a}_{r}(f \land g) \land \neg L^{a}_{s}(f \land \neg g) \rightarrow \neg L^{a}_{r+s}f \end{array}$

CML+(A)

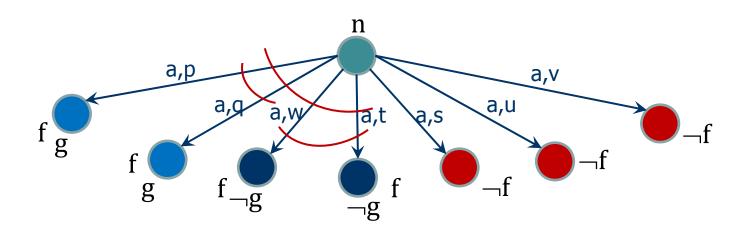
$$(B1) \vdash L^{a}_{0}f$$

$$(B2) \vdash L^{a}_{r+s}f \rightarrow \neg M^{a}_{r}f, s>0$$

$$(B3) \vdash \neg L^{a}_{r}f \rightarrow M^{a}_{r}f$$

$$(B4) \vdash \neg L^{a}_{r}(f \land g) \land \neg L^{a}_{s}(f \land \neg g) \rightarrow \neg L^{a}_{r+s}f$$

$$(B5) \vdash \neg M^{a}_{r}(f \land g) \land \neg M^{a}_{s}(f \land \neg g) \rightarrow \neg M^{a}_{r+s}f$$



Axiomatic Systems

CML(A)

$$\begin{split} & (A1) \vdash L^a{}_0 f \\ & (A2) \vdash L^a{}_{r+s} f \rightarrow L^a{}_r f \\ & (A3) \vdash L^a{}_r (f \land g) \land L^a{}_s (f \land \neg g) \rightarrow L^a{}_{r+s} f \\ & (A4) \vdash \neg L^a{}_r (f \land g) \land \neg L^a{}_s (f \land \neg g) \rightarrow \neg L^a{}_{r+s} f \end{split}$$

(R1) If $\vdash f \rightarrow g$, then $\vdash L^{a}_{r}f \rightarrow L^{a}_{r}g$ (R2) If $\forall r < s, \vdash f \rightarrow L^{a}_{r}g$, then $\vdash f \rightarrow L^{a}_{s}g$ (R3) If $\forall r > s, \vdash f \rightarrow L^{a}_{r}g$, then $\vdash f \rightarrow \neg T$ CML⁺(A)

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 \begin{array}{l} (B1) \vdash L^{a}{}_{0}f \\ (B2) \vdash L^{a}{}_{r+s}f \rightarrow \neg M^{a}{}_{r}f , s>0 \\ (B3) \vdash \neg L^{a}{}_{r}f \rightarrow M^{a}{}_{r}f \\ (B4) \vdash \neg L^{a}{}_{r}(f \land g) \land \neg L^{a}{}_{s}(f \land \neg g) \rightarrow \neg L^{a}{}_{r+s}f \\ (B5) \vdash \neg M^{a}{}_{r}(f \land g) \land \neg M^{a}{}_{s}(f \land \neg g) \rightarrow \neg M^{a}{}_{r+s}f \\ \end{array} 
 \begin{array}{l} (S1) If \vdash f \rightarrow g , \text{ then } \vdash L^{a}{}_{r}f \rightarrow L^{a}{}_{r}g \\ (S2) If \forall r < s, \vdash f \rightarrow L^{a}{}_{r}g , \text{ then } \vdash f \rightarrow L^{a}{}_{s}g \end{array}
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(S2) If \forall r < s, \vdash f \rightarrow L^{a}_{r}g, then \vdash f \rightarrow L^{a}_{s}g
(S3) If \forall r > s, \vdash f \rightarrow M^{a}_{r}g, then \vdash f \rightarrow M^{a}_{s}g
(S4) If \forall r > s, \vdash f \rightarrow L^{a}_{r}g, then \vdash f \rightarrow \neg T
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A. Heifetz, P. Mongin, Probability Logic for Type Spaces, 2001C. Kupke, D. Pattinson. On Modal Logics of Linear Inequalities, AiML 2010.

Metaproperties

<u>Metatheorem [Small model property]:</u>

If f is consistent (in CML(A) or CML⁺(A)), there exists a CMP (m, \mathcal{M}_{f}^{e}) that satisfies f. The support of \mathcal{M}_{f}^{e} is finite of cardinality bound by the dimension of f; the construction of \mathcal{M}_{f}^{e} is parametric (e>0) and depends on the *granularity* of f.

The granularity of a set $S \subseteq \mathbb{Q}^+$ is the least common denominator of the elements of S.

Metatheorem [Soundness & Weak Completeness]:

The axiomatic system of CML(A) and CML⁺(A) are sound and complete w.r.t. the Markovian semantics,

⊢f iff ⊨f.

- Stochastic bisimulation equates CMPs with identical stochastic behaviours
- CMLs are multimodal logics that characterize stochastic bisimulation
- CMLs are completely axiomatized for CMP-semantics
- We have a clear intuition of what a distance between CMPs should be



Classical Logic	Generalization
Truth values {0,1}	Interval [0,1]
Propositional function	Measurable function
State	Measure
The satisfiability relation ⊨	Integration ∫

D. Kozen, A Probabilistic PDL, 1985.

The satisfiability relation is replaced by a pseudometric over the space of CMPs.

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\begin{split} &d((m,\mathcal{M}),T)=0\\ &d((m,\mathcal{M}),\neg f)=1-d((m,\mathcal{M}),f)\\ &d((m,\mathcal{M}),f_1\wedge f_2)=\max\{d((m,\mathcal{M}),f_1),d((m,\mathcal{M}),f_2)\}\\ &d((m,\mathcal{M}),\ L^a_r f)=<r,\ \theta_a(m)([f])>\\ &d((m,\mathcal{M}),\ M^a_r f)=<\theta_a(m)([f]),\ r> \end{split}
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<r,s>= \begin{cases} (r-s)/r , if r>s \\ 0, otherwise \end{cases}
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(m, <i>M</i>)⊨T	always
(m,ℋ)⊨¬f	iff (m,ℋ)⊭f
(m, \mathcal{M})⊨f ₁ ∧f ₂	iff $(m, \mathcal{M}) \models f_1, (m, \mathcal{M}) \models f_2$
(m,ℋ)⊨ Lª _r f	iff θ _a (m)([f])≥r
(m,ℋ)⊨ Mª _r f	iff θ _a (m)([f])≤r,

Example:

$$\begin{split} (m,\mathcal{M}) &\vDash L^{a}{}_{r}f \implies \theta_{a}(m)([f]) \geq r \implies d((m,\mathcal{M}), \ L^{a}{}_{r}f) = 0 \\ (m,\mathcal{M}) &\nvDash L^{a}{}_{r}f \implies \theta_{a}(m)([f]) < r \implies d((m,\mathcal{M}), \ L^{a}{}_{r}f) > 0 \end{split}$$

 $\begin{aligned} d: & \mathscr{D} \times \mathcal{L} \rightarrow [0,1] \\ d((m,\mathcal{M}),T) = 0 \\ d((m,\mathcal{M}),\neg f) = 1 - d((m,\mathcal{M}),f) \\ d((m,\mathcal{M}),f_1 \wedge f_2) = \max\{d((m,\mathcal{M}),f_1),d((m,\mathcal{M}),f_2)\} \\ d((m,\mathcal{M}), L^a_r f) = <r, \theta_a(m)([f]) > \\ d((m,\mathcal{M}), M^a_r f) = <\theta_a(m)([f]), r > \end{aligned}$

$$< r, s > =$$
 $\begin{bmatrix} (r-s)/r , if r > s \\ 0, otherwise \end{bmatrix}$

$$\begin{split} \mathsf{D}: & \mathcal{D} \times & \mathcal{D} \to [0,1], \\ & \mathsf{D}((\mathsf{m},\mathcal{M}),(\mathsf{m}',\mathcal{M}')) = \sup\{|\mathsf{d}((\mathsf{m},\mathcal{M}),\mathsf{f}) - \mathsf{d}((\mathsf{m}',\mathcal{M}'),\mathsf{f})|, \ \mathsf{f} \in \mathcal{L}\} \end{split}$$

$$\begin{split} \delta: \mathcal{L} \times \mathcal{L} &\to [0, 1], \\ \delta(f, f') = \sup\{|d((m, \mathcal{M}), f) - d((m, \mathcal{M}), f')|, (m, \mathcal{M}) \in \mathcal{D}\} \end{split}$$

Metaproperties

Theorem [Strong Robustness]:

For arbitrary $f, f' \in \mathcal{L}$, and arbitrary $(m, \mathcal{M}) \in \mathcal{D}$, $d((m, \mathcal{M}), f') \leq d((m, \mathcal{M}), f) + \delta(f, f')$

 $\delta^*: \mathcal{L} \times \mathcal{L} \rightarrow [0,1],$

 $\delta^*(\mathbf{f},\mathbf{f}') = \sup\{|d((\mathbf{m},\mathcal{M}^{\mathbf{e}}_{\mathsf{f}\wedge\mathsf{f}'}),\mathbf{f}) - d((\mathbf{m},\mathcal{M}_{\mathsf{f}\wedge\mathsf{f}'}),\mathbf{f}')|, \ \mathbf{m}\in \sup(\mathcal{M}_{\mathsf{f}\wedge\mathsf{f}'})\}$

where $\mathcal{M}^{e}_{f \wedge f'}$ is the finite model of $\sim (f \wedge f')$ of parameter e > 0.

Lemma: For arbitrary $f, f' \in \mathcal{L}$

 $\delta(\mathbf{f},\mathbf{f}') \leq \delta^*(\mathbf{f},\mathbf{f}') + 2/e$

Theorem [Weak Robustness]:

For arbitrary f,f' $\in \mathcal{L}$, and arbitrary (m, \mathcal{M}) $\in \mathcal{D}$,

 $\mathsf{d}((\mathsf{m},\mathcal{M}),\mathsf{f}') \leq \mathsf{d}((\mathsf{m},\mathcal{M}),\mathsf{f}) + \delta^*(\mathsf{f},\mathsf{f}') + 2/\mathsf{e}$

Towards a metric semantics

Working hypothesis:

- Let (\$\vec{p},D\$) be a pseudometrizable space of Markovian systems such that D converges to bisimulation;
- Let L be the continuous Markovian logic (that characterizes the bisimulation and is completely axiomatized for Ø)

 $\begin{aligned} \mathcal{L} & f := T \mid \neg f \mid f \land f \mid L^{a}_{r}f \mid M^{a}_{r}f \\ \mathcal{L}(+) & g := T \mid g \land g \mid L^{a}_{r}f \mid M^{a}_{r}f \\ \mathcal{L}(-) &= \mathcal{L} - \mathcal{L}(+) \end{aligned}$

Theorem:	If $\vdash f \leftrightarrow g$, then $\delta(f,g)=0$.
Theorem:	If $\delta(f,g)=0$ and $f\in \mathscr{L}(+)$, then $\vdash g \to f$.
Theorem:	If $\delta(f,g)=0$ and $f,g\in \mathscr{L}(+)$, then $\vdash f \leftrightarrow g$.

In this context, δ is a pseudometric that measure the syntactical equivalence on $\mathscr{L}(+)$.

Future work: some dualities

Working hypothesis:

- Let (\$\vec{p},D\$) be a pseudometrizable space of Markovian systems such that D converges to bisimulation;
- Let L be the continuous Markovian logic (that characterizes the bisimulation and is completely axiomatized for (2)
- \mathcal{L} has a *canonical model* $\Omega = (\Omega, 2^{\Omega}, \theta)$, where each $F \in \Omega$ is a maximally consistent set of formulas: for each CMP (\mathcal{M}, m) there exists a unique $F \in \Omega$ such that $(m, \mathcal{M}) \sim (F, \Omega)$.

In fact, $F = \{f \in \mathscr{L}, (m, \mathcal{M}) \models f\}.$

If for an arbitrary distance D we use D_H to denote the Hausdorff distance associated to D, then the complete axiomatization suggest the following conjectures.

Conjecture1:	$(D_H)_H = D$
Conjecture2:	$(\delta_{H})_{H} = \delta$