

A Quantitative Characterization of Weighted Kripke Structures in Temporal Logic.

Uli Fahrenberg and Kim G. Larsen and Claus Thrane

Dept. Computer Science



QUANTLOG 2009

Related work

- Brown, Clarke, Grömberg: Characterising Kripke Structures in Temporal Logic.
- Chatterjee, Doyern and Henzinger: Quantitative Languages.
- Henzinger, Majumdar, and Prabhu: Quantifying Similarities Between Timed Systems.
- Alfaro, Faella, and Stoelinga: Linear and Branching Metrics for Quantitative Transition Systems.

Strong ties to probabilistic systems and weighted automata.

Related work

- **Brown, Clarke, Grömberg: Characterising Kripke Structures in Temporal Logic.**
- Chatterjee, Doyern and Henzinger: Quantitative Languages.
- Henzinger, Majumdar, and Prabhu: Quantifying Similarities Between Timed Systems.
- Alfaro, Faella, and Stoelinga: Linear and Branching Metrics for Quantitative Transition Systems.

Strong ties to probabilistic systems and weighted automata.

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

or

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

or

- Checking of models,

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

or

- Checking of models,
- Checking of models with quantities;
e.g. checking time-constraints

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

or

- Checking of models,
- Checking of models with quantities;
e.g. checking time-constraints
- Quantitative checking of models;
e.g. measuring the distance between discrete actions.

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

or

- Checking of models,
- Checking of models with quantities;
e.g. checking time-constraints
- Quantitative checking of models;
e.g. measuring the distance between discrete actions.
- Quantitative checking of models with quantities.
e.g. measuring the distance between real weighted transitions.

Quantitative model checking

- Model-checking: μ -calculus, CTL, LTL ...
- **Quantitative Model** Checking: TCTL, WCTL, PCTL ...
- Quantitative **Model Checking**

or

- Checking of models,
- Checking of models with quantities;
e.g. checking time-constraints
- Quantitative checking of models;
e.g. measuring the distance between discrete actions.
- **Quantitative checking of models with quantities.**
e.g. measuring the distance between real weighted transitions.

What is quantitative model-checking?

The quantitative model-checking problem

Given a state s of a structure M , and a logical formulae φ

- Does φ hold at s ? **i.e.** $M, s \models \varphi$?

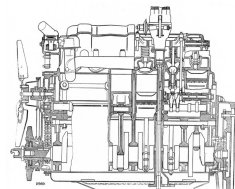
What is quantitative model-checking?

The quantitative model-checking problem

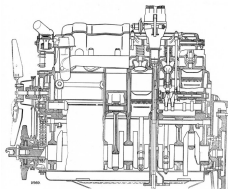
Given a state s of a structure M , and a logical formulae φ

- What is the degree ε with which φ holds at s ?
i.e. compute ε s.t. $\llbracket \varphi \rrbracket_M(s) = \varepsilon$?

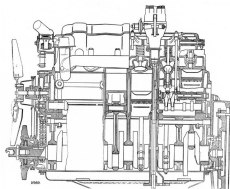
Why deal with quantitative model-checking?



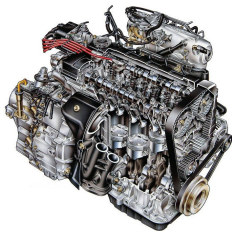
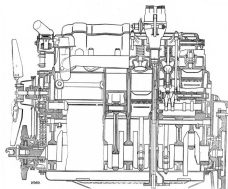
Why deal with quantitative model-checking?



Why deal with quantitative model-checking?



Why deal with quantitative model-checking?



Decidability & Expressiveness

Model-checking

Given φ and $s \in S$: $s \models \varphi$?

Satisfiability

Given φ : $\exists s \in S. s \models \varphi$?

Adequacy

Given $s, t \in S$: $s \sim t$ iff $\forall \varphi. s \models \varphi \iff t \models \varphi$?

Characteristic properties

Given $s, t \in S$: $t \models \varphi_s \iff s \sim t$?

Decidability & Expressiveness

Model-checking

Given φ and $s \in S$: $\llbracket \varphi \rrbracket (s) = \varepsilon$?

Satisfiability

Given φ : $\exists s \in S. \llbracket \varphi \rrbracket (s) = 0$?

Adequacy

Given $s, t \in S$: $s \sim_\varepsilon t$ iff $\forall \varphi. |\llbracket \varphi \rrbracket (s) - \llbracket \varphi \rrbracket (t)| \leq \varepsilon$?

Characteristic properties

Given $s, t \in S$: $\llbracket \varphi_s \rrbracket (t) = \varepsilon \iff t \sim_\varepsilon s$?

Decidability & Expressiveness

Model-checking

Given φ and $s \in S$: $\llbracket \varphi \rrbracket (s) = \varepsilon$?

Satisfiability

Given φ : $\exists s \in S. \llbracket \varphi \rrbracket (s) = 0$?

Adequacy

Given $s, t \in S$: $s \sim_\varepsilon t$ iff $\forall \varphi. |\llbracket \varphi \rrbracket (s) - \llbracket \varphi \rrbracket (t)| \leq \varepsilon$?

Characteristic properties

Given $s, t \in S$: $\llbracket \varphi_s \rrbracket (t) = \varepsilon \iff t \sim_\varepsilon s$?

- 1 Introduction
- 2 Weighted Kripke Structures
- 3 Metrics and Distances
- 4 The WCTL Logic
- 5 Conclusion

Weighted Kripke structure

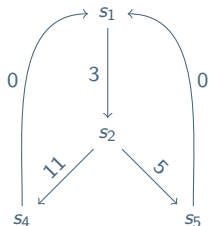
For a finite set \mathcal{AP} of atomic propositions, a **weighted Kripke structure** is a quadruple $M = (S, T, \mathcal{L}, w)$ where

- S is a finite set of states,
- $T \subseteq S \times S$ is a transition relation
- $\mathcal{L} : S \rightarrow 2^{\mathcal{AP}}$ is the proposition labelling, and
- $w : T \rightarrow \mathbb{R}_{\geq 0}$ assigns weights to transitions.

Weighted Kripke structure

For a finite set \mathcal{AP} of atomic propositions, a **weighted Kripke structure** is a quadruple $M = (S, T, \mathcal{L}, w)$ where

- S is a finite set of states,
- $T \subseteq S \times S$ is a transition relation
- $\mathcal{L} : S \rightarrow 2^{\mathcal{AP}}$ is the proposition labelling, and
- $w : T \rightarrow \mathbb{R}_{\geq 0}$ assigns weights to transitions.



Bisimulation for WKS

Let (S, T, \mathcal{L}, w) be a WKS on a set \mathcal{AP} of atomic propositions. A relation $B \subseteq S \times S$ is a c -weighted bisimulation relation, provided that for all $(s, t) \in B$:

- $\mathcal{L}(s) = \mathcal{L}(t)$ and
- if $s \xrightarrow{c} s'$, then also $t \xrightarrow{c} t'$ where $(s', t') \in B$ for some $t' \in S'$,
- if $t \xrightarrow{c} t'$, also also $s \xrightarrow{c} s'$ where $(s', t') \in B$ for some $s' \in S$;

We say that s and t are c -weighted bisimilar, written $s \sim t$ if $(s, t) \in B$ for some unweighted bisimulation B .

Bisimulation for WKS

Let (S, T, \mathcal{L}, w) be a WKS on a set \mathcal{AP} of atomic propositions. A relation $B \subseteq S \times S$ is a **un**weighted bisimulation relation, provided that for all $(s, t) \in B$:

- $\mathcal{L}(s) = \mathcal{L}(t)$ and
- if $s \rightarrow s'$, then also $t \rightarrow t'$ where $(s', t') \in B$ for some $t' \in S'$,
- if $t \rightarrow t'$, also also $s \rightarrow s'$ where $(s', t') \in B$ for some $s' \in S$;

We say that s and t are **un**weighted bisimilar, written $s \overset{u}{\sim} t$ if $(s, t) \in B$ for some unweighted bisimulation B .

Bridging the Boolean gap

By definition membership of a relation, is **true** or **false**; We provide a family of relations st.

$$\sim \supseteq \dots \supseteq \sim_i \supseteq \sim_j \supseteq \dots \supseteq \overset{u}{\sim}$$

For $i, j \in \mathbb{R}_{\geq 0}$ and $i < j$

Equivalence relations

Definition

Given a set X a binary relation $R \subseteq X \times X$ is an equivalence relation, if and only if,
 $\forall x, y, z \in X, R$ is

- $x R x$
- $x R y \iff y R x$
- $x R y$ and $y R z$ then $x R z$

...or

Given a set X a map $R : X \times X \rightarrow \{0, 1\}$ is an equivalence relation, if and only if, $\forall x, y, z \in X, R$ and:

- $R(x, x) = 0$
- $R(x, y) = R(y, x)$
- $R(x, y) \leq R(y, z) + R(y, z)$

Where **tt** = 0 and **ff** = 1

Equivalence relations

Definition

Given a set X a binary relation $R \subseteq X \times X$ is an equivalence relation, if and only if,
 $\forall x, y, z \in X, R$ is

- $x R x$
- $x R y \iff y R x$
- $x R y$ and $y R z$ then $x R z$

...or

Given a set X a map $R : X \times X \rightarrow \{0, 1\}$ is an equivalence relation, if and only if, $\forall x, y, z \in X, R$ and:

- $R(x, x) = 0$
- $R(x, y) = R(y, x)$
- $R(x, y) \leq R(y, z) + R(y, z)$

Where **tt** = 0 and **ff** = 1

Boolean

Metrics

Definition

Given a set X . Then a metric on X is a function $d : X \times X \rightarrow \mathbb{R}_{\geq 0}$ which $\forall x, y, z \in X$ satisfies:

- 1 $d(x, y) = 0$ if and only if $x = y$
- 2 $d(x, y) = d(y, x)$
- 3 $d(x, z) \leq d(x, y) + d(y, z)$

Distances on sequences of real numbers

For sequences $a = (a_i)$, $b = (b_i)$, we may consider the following distances:

$$d_{\cdot}(a, b) = \sup_i \{|a_i - b_i|\}$$

$$d_{+}(a, b) = \sum_i |a_i - b_i|$$

$$d_{\pm}(a, b) = \sup_i \left\{ \left| \sum_{j=0}^i a_j - \sum_{j=0}^i b_j \right| \right\} \quad ([HMP'05])$$

Quantifying bisimulation

We extend bisimulation, with $d_.$, d_+ or d_{\pm} measurements, as well as a **discounting** factor $0 \leq \lambda \leq 1$.

- Point-wise (bi)simulation
- Accumulated (bi)simulation
- Max-lead (bi)simulation
(Henzinger et. al, FORMATS'05 to be poly-time^a decidable for timed automata)

^ain the size of the region graph, which in turn is exponential in the size of clocks

A **bisimulation distance** is a function $d : S \times S \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$ which satisfies the following for all $s_1, s_2, s_3 \in S$:

- $d(s_1, s_1) = 0$,
- $d(s_1, s_2) + d(s_2, s_3) \geq d(s_1, s_3)$,
- $d(s_1, s_2) = d(s_2, s_1)$,
- $s_1 \sim s_2$ implies $d(s_1, s_2) = 0$
- $d(s_1, s_2) \neq \infty$ implies $s_1 \overset{u}{\sim} s_2$

Accumulated bisimulation distance

A family of relations $\mathbf{R} = \{\mathcal{R}_\varepsilon \subseteq S \times S \mid \varepsilon > 0\}$

Accumulated bisimulation distance

A family of relations $\mathbf{R} = \{\mathcal{R}_\varepsilon \subseteq S \times S \mid \varepsilon > 0\}$ on a WKS (S, T, \mathcal{L}, w) is an **accumulating bisimulation family** provided that for all $(s, t) \in \mathcal{R}_\varepsilon \in \mathbf{R}$:

- $\mathcal{L}(s) = \mathcal{L}(t)$ and
- for all $s \xrightarrow{c} s'$, also $t \xrightarrow{d} t'$ with $|c - d| \leq \varepsilon$ for some $d \in \mathbb{R}_{\geq 0}$ and $(s', t') \in \mathcal{R}_{\varepsilon'} \in \mathbf{R}$ with $\varepsilon' \leq \frac{\varepsilon - |c - d|}{\lambda}$,
- for all $t \xrightarrow{c} t'$, also $s \xrightarrow{d} s'$ with $|c - d| \leq \varepsilon$ for some $d \in \mathbb{R}_{\geq 0}$ and $(s', t') \in \mathcal{R}_{\varepsilon'} \in \mathbf{R}$ with $\varepsilon' \leq \frac{\varepsilon - |c - d|}{\lambda}$.

We write $s \overset{+}{\sim}_\varepsilon t$ if $(s, t) \in \mathcal{R}_\varepsilon \in \mathbf{R}$ for an accumulating bisimulation family \mathbf{R} .

Linear and Branching d_+ -distances

(simulation as equations)

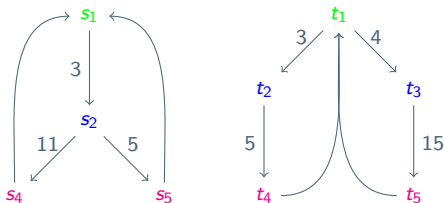
For states $s, t \in S$, the **accumulated branching distances** is the minimal fixed points to the following recursive equation:

$$\lambda s, t \lambda_+ = \sup_{s \xrightarrow{c} s'} \inf_{t \xrightarrow{d} t'} |c - d| + \lambda \cdot \lambda s', t' \lambda_+$$

and the **accumulated linear distance** is:

$$|s, t|_+ = \sup_{\sigma \in P(s)} \inf_{\sigma' \in P(t)} \sum_i \lambda^i |\sigma(i)_w - \sigma'(i)_w|$$

Example

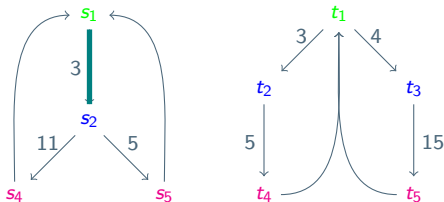


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example

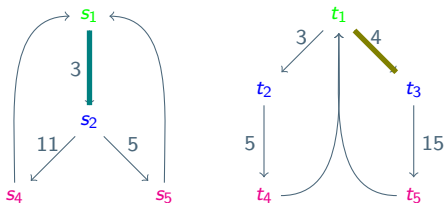


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example

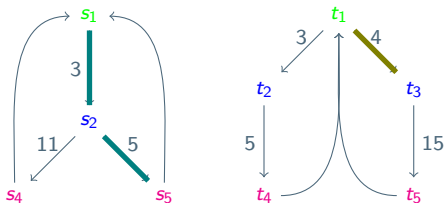


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example

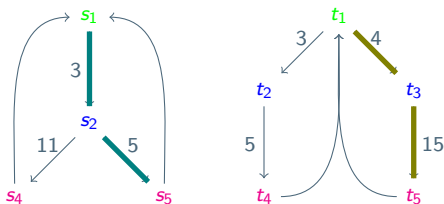


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example

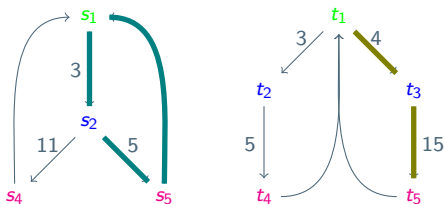


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example

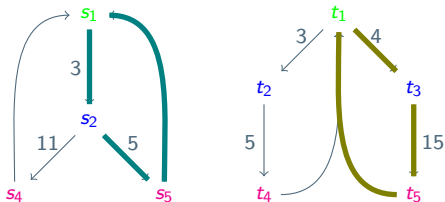


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example

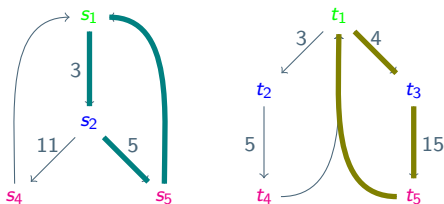


For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

Example



For $\lambda = 0.9$ $\mathcal{AP} = \{\bullet, \bullet, \bullet\}$ Simulation distances:

$$|s_1, t_1|_+ = \sum_i (1 + 4\lambda)\lambda^{3i} \approx 17.0$$

$$\lambda s_1, t_1 \lambda_+ = 1 + 10\lambda + \lambda^3 |s_1, t_1|_+ \approx 36.9$$

And w.r.t their accumulating bisimulation distance; $s_1 \overset{+}{\sim}_{37} t_1$.

Properties

- For all states $s, t \in S$, we have

$$|s, t|_{\cdot} \leq \lambda s, t|_{\cdot} \quad |s, t|_{+} \leq \lambda s, t|_{+} \quad |s, t|_{\pm} \leq \lambda s, t|_{\pm}$$

- The distances $|\cdot, \cdot|_{\cdot}$ and $\lambda \cdot, \cdot|_{\cdot}$ are topologically inequivalent. Similarly, $|\cdot, \cdot|_{+}$ and $\lambda \cdot, \cdot|_{+}$, and also $|\cdot, \cdot|_{\pm}$ and $\lambda \cdot, \cdot|_{\pm}$, are topologically inequivalent.

Computability for Weighted Timed Automata

Theorem

For discounting factor $\lambda < 1$ and $|\cdot, \cdot|$ any of the three **trace distances**, it is undecidable whether $|s, t| = 0$ for weighted timed automata.

Theorem

For discounting factor $\lambda < 1$, **accumulating branching distance** from deterministic to non-deterministic **weighted timed automata** is computable.

Setting the scene for the logic

- A simple general syntactic extension of CTL,
- The semantics of a formulae φ defines a map $[[\cdot]] : S \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$. such that
 - The state semantics are **shared** for d , and d_+ and d_{\pm} .
 - The path semantics are **specific** the respective distances.
- To obtain a correspondence, with the **bisimulation distance**.

Weighted CTL

For any of the metrics d_* and d_+ and d_{\pm} , we define the syntax:

Definition

For $p \in \mathcal{AP}$, Φ generates the set of state formulae, and Ψ , the set of path formulae, annotated by weights $c \in \mathbb{R}_{\geq 0}$, according to the following abstract syntax:

$$\begin{aligned} \Phi &::= p \mid \neg p \mid \Phi_1 \wedge \Phi_2 \mid \Phi_1 \vee \Phi_2 \mid E\Psi \mid A\Psi \\ \Psi &::= X_c\Phi \mid G_c\Phi \mid F_c\Phi \mid [\Phi_1 U_c \Phi_2] \end{aligned}$$

The logic WCTL is the set of state formulae, which we denote $\mathcal{L}_w(\mathcal{AP})$ or simply \mathcal{L}_w .

Semantics for state formulae

Let $\varphi, \varphi_1, \varphi_2$ be state formulae and ψ a path formula. The valuation $\llbracket \cdot \rrbracket : S \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$ is defined inductively.

$$\llbracket p \rrbracket(s) = \begin{cases} 0 & \text{if } p \in \mathcal{L}(s) \\ \infty & \text{otherwise} \end{cases}$$

$$\llbracket \neg p \rrbracket(s) = \begin{cases} 0 & \text{if } p \in \mathcal{AP} \setminus \mathcal{L}(s) \\ \infty & \text{otherwise} \end{cases}$$

$$\llbracket \varphi_1 \vee \varphi_2 \rrbracket(s) = \inf \{ \llbracket \varphi_1 \rrbracket(s), \llbracket \varphi_2 \rrbracket(s) \}$$

$$\llbracket \varphi_1 \wedge \varphi_2 \rrbracket(s) = \sup \{ \llbracket \varphi_1 \rrbracket(s), \llbracket \varphi_2 \rrbracket(s) \}$$

$$\llbracket E\psi \rrbracket(s) = \inf \{ \llbracket \psi \rrbracket(\sigma) \mid \sigma \in P(s) \}$$

$$\llbracket A\psi \rrbracket(s) = \sup \{ \llbracket \psi \rrbracket(\sigma) \mid \sigma \in P(s) \}$$

d_+ Path semantics

$$\llbracket \varphi \rrbracket(\sigma) = \llbracket \varphi \rrbracket(\sigma(0)_s)$$

$$\llbracket X_c \varphi \rrbracket(\sigma) = |c - \sigma(0)_w| + \lambda \llbracket \varphi \rrbracket(\sigma^1)$$

$$\llbracket F_c \varphi \rrbracket(\sigma) = \inf_k \left(\left| \sum_{j=0}^{k-1} \lambda^j \sigma(j)_w - c \right| + \lambda^k \llbracket \varphi \rrbracket(\sigma^k) \right)$$

$$\llbracket G_c \varphi \rrbracket(\sigma) = \sup_k \left(\left| \sum_{j=0}^{k-1} \lambda^j \sigma(j)_w - c \right| + \lambda^k \llbracket \varphi \rrbracket(\sigma^k) \right)$$

$$\llbracket \varphi_1 U_c \varphi_2 \rrbracket(\sigma) = \inf_k \left(\left| \sum_{j=0}^{k-1} \lambda^j \llbracket \varphi_1 \rrbracket(\sigma^j) - c \right| + \lambda^k \llbracket \varphi_2 \rrbracket(\sigma^k) \right)$$

Properties of the logic

Theorem: Adequacy

For states $s, t \in S$, $s \overset{\dagger}{\sim}_{\varepsilon} t$ if and only if $|\llbracket \varphi \rrbracket(s) - \llbracket \varphi \rrbracket(t)| \leq \varepsilon$ for all $\varphi \in \mathcal{L}_W$.

Theorem: Expressivity & Characteristic formulae

For each $s \in S$ and every $\gamma \in \mathbb{R}_+$, there exists a state formula $\varphi_{\gamma}^s \in \mathcal{L}_W$ which characterizes s up to accumulating bisimulation and up to γ , **i.e.** such that for all $s' \in S$, $s \overset{\dagger}{\sim}_{\varepsilon} s'$ if and only if $\llbracket \varphi_{\gamma}^s \rrbracket(s') \in [\varepsilon - \gamma, \varepsilon + \gamma]$ for all γ .

Properties of the logic

Theorem: Adequacy

For states $s, t \in S$, $s \overset{\dagger}{\sim}_{\varepsilon} t$ if and only if $|\llbracket \varphi \rrbracket(s) - \llbracket \varphi \rrbracket(t)| \leq \varepsilon$ for all $\varphi \in \mathcal{L}_W$.

Theorem: Expressivity & Characteristic formulae

For each $s \in S$ and every $\gamma \in \mathbb{R}_+$, there exists a state formula $\varphi_{\gamma}^s \in \mathcal{L}_W$ which characterizes s up to accumulating bisimulation and up to γ , **i.e.** such that for all $s' \in S$, $s \overset{\dagger}{\sim}_{\varepsilon} s'$ if and only if $\llbracket \varphi_{\gamma}^s \rrbracket(s') \in [\varepsilon - \gamma, \varepsilon + \gamma]$ for all γ .

Building Characteristic formulae

HML for infinite state systems [Graf, Sifakis '86]

Use recursive properties

CTL for finite state systems [Brown, Clarke, Grömsberg '87]

Use the characteristic number c

WCTL for accumulating quantitative checking

Building Characteristic formulae

HML for infinite state systems [Graf, Sifakis '86]

Use recursive properties

CTL for finite state systems [Brown, Clarke, Grömsberg '87]

Use the characteristic number c

WCTL for accumulating quantitative checking

For infinite sums, the constant c , is insufficient

d_+ Characteristic formulae

For each $s \in S$ and $n \in \mathbb{N}$, denote $\mathcal{L}(s) = \{p_1, \dots, p_k\}$ and $\mathcal{AP} \setminus \mathcal{L}(s) = \{q_1, \dots, q_\ell\}$ the formula $\varphi(s, n)$ is defined inductively as:

$$\varphi(s, 0) = (p_1 \wedge \dots \wedge p_k) \wedge (\neg q_1 \wedge \dots \wedge \neg q_\ell)$$

$$\varphi(s, n+1) = \bigwedge_{s \xrightarrow{w} s'} EX_w \varphi(s', n) \wedge \bigwedge_{w: s \xrightarrow{w} s'} AX_w \left(\bigvee_{s \xrightarrow{w} s'} \varphi(s', n) \right) \wedge \varphi(s, 0)$$

d_+ Characteristic formulae

For each $s \in S$ and $n \in \mathbb{N}$, denote $\mathcal{L}(s) = \{p_1, \dots, p_k\}$ and $\mathcal{AP} \setminus \mathcal{L}(s) = \{q_1, \dots, q_\ell\}$ the formula $\varphi(s, n)$ is defined inductively as:

$$\varphi(s, 0) = (p_1 \wedge \dots \wedge p_k) \wedge (\neg q_1 \wedge \dots \wedge \neg q_\ell)$$

$$\varphi(s, n+1) = \bigwedge_{s \xrightarrow{w} s'} EX_w \varphi(s', n) \wedge \bigwedge_{w: s \xrightarrow{w} s'} AX_w \left(\bigvee_{s \xrightarrow{w} s'} \varphi(s', n) \right) \wedge \varphi(s, 0)$$

- It is easy to see that $\llbracket \varphi(s, n) \rrbracket(s) = 0$ for all n .

d_+ Characteristic formulae

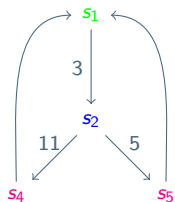
For each $s \in S$ and $n \in \mathbb{N}$, denote $\mathcal{L}(s) = \{p_1, \dots, p_k\}$ and $\mathcal{AP} \setminus \mathcal{L}(s) = \{q_1, \dots, q_\ell\}$ the formula $\varphi(s, n)$ is defined inductively as:

$$\varphi(s, 0) = (p_1 \wedge \dots \wedge p_k) \wedge (\neg q_1 \wedge \dots \wedge \neg q_\ell)$$

$$\varphi(s, n+1) = \bigwedge_{s \xrightarrow{w} s'} EX_w \varphi(s', n) \wedge \bigwedge_{w: s \xrightarrow{w} s'} AX_w \left(\bigvee_{s \xrightarrow{w} s'} \varphi(s', n) \right) \wedge \varphi(s, 0)$$

- It is easy to see that $\llbracket \varphi(s, n) \rrbracket(s) = 0$ for all n .
- Observe that for each $\gamma > 0$, there is $n(\gamma) \in \mathbb{N}$ such that $\varphi(s, n(\gamma))$ can play the role of φ_γ^s in the theorem.

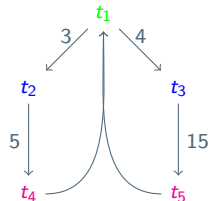
Example



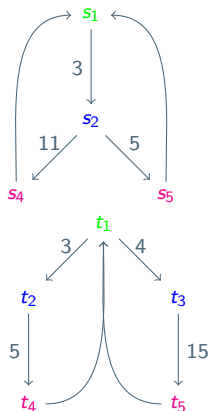
$$\varphi(t_1, n) = (\bullet \wedge \neg \bullet \wedge \neg \bullet)$$

$$\wedge EX_3\varphi(t_2, n-1) \wedge EX_4\varphi(t_3, n-1)$$

$$\wedge AX_3\varphi(t_2, n-1) \wedge AX_4\varphi(t_3, n-1)$$



Example



$$\varphi(t_1, n) = (\bullet \wedge \neg \bullet \wedge \neg \bullet)$$

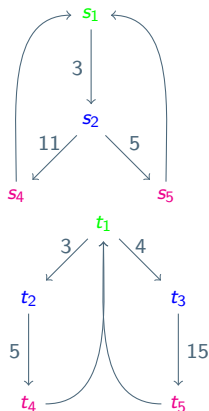
$$\wedge EX_3\varphi(t_2, n-1) \wedge EX_4\varphi(t_3, n-1)$$

$$\wedge AX_3\varphi(t_2, n-1) \wedge AX_4\varphi(t_3, n-1)$$

$$\varphi(t_2, n-1) = (\neg \bullet \wedge \bullet \wedge \neg \bullet)$$

$$\wedge EX_5\varphi(t_4, n-2) \wedge AX_5\varphi(t_4, n-2)$$

Example



$$\varphi(t_1, n) = (\bullet \wedge \neg \bullet \wedge \neg \bullet)$$

$$\wedge EX_3\varphi(t_2, n-1) \wedge EX_4\varphi(t_3, n-1)$$

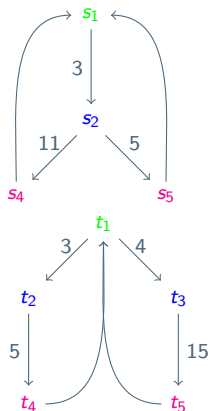
$$\wedge AX_3\varphi(t_2, n-1) \wedge AX_4\varphi(t_3, n-1)$$

$$\varphi(t_2, n-1) = (\neg \bullet \wedge \bullet \wedge \neg \bullet)$$

$$\wedge EX_5\varphi(t_4, n-2) \wedge AX_5\varphi(t_4, n-2)$$

$$\varphi(t_3, n-1) = \dots$$

Example



$$\varphi(t_1, n) = (\bullet \wedge \neg \bullet \wedge \neg \bullet)$$

$$\wedge EX_3\varphi(t_2, n-1) \wedge EX_4\varphi(t_3, n-1)$$

$$\wedge AX_3\varphi(t_2, n-1) \wedge AX_4\varphi(t_3, n-1)$$

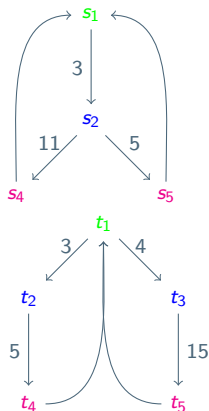
$$\varphi(t_2, n-1) = (\neg \bullet \wedge \bullet \wedge \neg \bullet)$$

$$\wedge EX_5\varphi(t_4, n-2) \wedge AX_5\varphi(t_4, n-2)$$

$$\varphi(t_3, n-1) = \dots$$

$$\varphi(t_4, n-2) = \dots$$

Example



$$\varphi(t_1, n) = (\bullet \wedge \neg \bullet \wedge \neg \bullet)$$

$$\wedge EX_3\varphi(t_2, n-1) \wedge EX_4\varphi(t_3, n-1)$$

$$\wedge AX_3\varphi(t_2, n-1) \wedge AX_4\varphi(t_3, n-1)$$

$$\varphi(t_2, n-1) = (\neg \bullet \wedge \bullet \wedge \neg \bullet)$$

$$\wedge EX_5\varphi(t_4, n-2) \wedge AX_5\varphi(t_4, n-2)$$

$$\varphi(t_3, n-1) = \dots$$

$$\varphi(t_4, n-2) = \dots$$

$$\varphi(t_5, n-2) = \dots$$

Example (cont.)

For $\llbracket \varphi_{0.1}^{t_1} \rrbracket(s_1)$ we compute:

$$\llbracket \varphi(t_1, 56) \rrbracket(s_1) = \max \left\{ \begin{array}{l} \max \left\{ \begin{array}{l} \inf\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \inf\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max \left\{ \begin{array}{l} \sup\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \sup\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max\{\llbracket \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1)\} \end{array} \right.$$

Example (cont.)

For $\llbracket \varphi_{0.1}^{t_1} \rrbracket(s_1)$ we compute:

$$\llbracket \varphi(t_1, 56) \rrbracket(s_1) = \max \begin{cases} \max \left\{ \begin{array}{l} \inf \{ |3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1) \} \\ \inf \{ |4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1) \} \end{array} \right. \\ \max \left\{ \begin{array}{l} \sup \{ |3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1) \} \\ \sup \{ |4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1) \} \end{array} \right. \\ \max \{ \llbracket \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1) \} \end{cases}$$

$$\llbracket \varphi(t_3, 55) \rrbracket(\sigma = s_1 \xrightarrow{3} s_2 \xrightarrow{5} s_5 \xrightarrow{0} s_1 \dots) =$$

Example (cont.)

For $\llbracket \varphi_{0.1}^{t_1} \rrbracket(s_1)$ we compute:

$$\llbracket \varphi(t_1, 56) \rrbracket(s_1) = \max \left\{ \begin{array}{l} \max \left\{ \begin{array}{l} \inf\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \inf\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max \left\{ \begin{array}{l} \sup\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \sup\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max\{\llbracket \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1)\} \end{array} \right.$$

$$\llbracket \varphi(t_3, 55) \rrbracket(\sigma = s_1 \xrightarrow{3} s_2 \xrightarrow{5} s_5 \xrightarrow{0} s_1 \cdots) =$$

$$\llbracket \varphi(t_3, 55) \rrbracket(s_2) = \max \left\{ \begin{array}{l} \inf\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \sup\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 54) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \max\{\llbracket \neg \bullet \rrbracket(s_2), \llbracket \bullet \rrbracket(s_2), \llbracket \neg \bullet \rrbracket(s_2)\} \end{array} \right.$$

Example (cont.)

For $\llbracket \varphi_{0.1}^{t_1} \rrbracket(s_1)$ we compute:

$$\llbracket \varphi(t_1, 56) \rrbracket(s_1) = \max \left\{ \begin{array}{l} \max \left\{ \begin{array}{l} \inf\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \inf\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max \left\{ \begin{array}{l} \sup\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \sup\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max\{\llbracket \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1)\} \end{array} \right.$$

$$\llbracket \varphi(t_3, 55) \rrbracket(\sigma = s_1 \xrightarrow{3} s_2 \xrightarrow{5} s_5 \xrightarrow{0} s_1 \cdots) =$$

$$\llbracket \varphi(t_3, 55) \rrbracket(s_2) = \max \left\{ \begin{array}{l} \inf\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \sup\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 54) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \max\{\llbracket \neg \bullet \rrbracket(s_2), \llbracket \bullet \rrbracket(s_2), \llbracket \neg \bullet \rrbracket(s_2)\} \end{array} \right.$$

...

Example (cont.)

For $\llbracket \varphi_{0.1}^{t_1} \rrbracket(s_1)$ we compute:

$$\llbracket \varphi(t_1, 56) \rrbracket(s_1) = \max \left\{ \begin{array}{l} \max \left\{ \begin{array}{l} \inf\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \inf\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max \left\{ \begin{array}{l} \sup\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \sup\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max\{\llbracket \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1)\} \end{array} \right.$$

$$\llbracket \varphi(t_3, 55) \rrbracket(\sigma = s_1 \xrightarrow{3} s_2 \xrightarrow{5} s_5 \xrightarrow{0} s_1 \dots) =$$

$$\llbracket \varphi(t_3, 55) \rrbracket(s_2) = \max \left\{ \begin{array}{l} \inf\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \sup\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 54) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \max\{\llbracket \neg \bullet \rrbracket(s_2), \llbracket \bullet \rrbracket(s_2), \llbracket \neg \bullet \rrbracket(s_2)\} \end{array} \right.$$

...

$$\llbracket \varphi(t_5, 54) \rrbracket(s_4) = \max \dots$$

Example (cont.)

For $\llbracket \varphi_{0.1}^{t_1} \rrbracket(s_1)$ we compute: ≈ 36.809

$$\llbracket \varphi(t_1, 56) \rrbracket(s_1) = \max \left\{ \begin{array}{l} \max \left\{ \begin{array}{l} \inf\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \inf\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max \left\{ \begin{array}{l} \sup\{|3 - \sigma(0)_w| + \lambda \llbracket \varphi(t_2, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \\ \sup\{|4 - \sigma(0)_w| + \lambda \llbracket \varphi(t_3, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_1)\} \end{array} \right. \\ \max\{\llbracket \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1), \llbracket \neg \bullet \rrbracket(s_1)\} \end{array} \right.$$

$$\llbracket \varphi(t_3, 55) \rrbracket(\sigma = s_1 \xrightarrow{3} s_2 \xrightarrow{5} s_5 \xrightarrow{0} s_1 \dots) =$$

$$\llbracket \varphi(t_3, 55) \rrbracket(s_2) = \max \left\{ \begin{array}{l} \inf\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 55) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \sup\{|15 - \sigma(0)_w| + \lambda \llbracket \varphi(t_5, 54) \rrbracket(\sigma^1) \mid \sigma \in P(s_2)\} \\ \max\{\llbracket \neg \bullet \rrbracket(s_2), \llbracket \bullet \rrbracket(s_2), \llbracket \neg \bullet \rrbracket(s_2)\} \end{array} \right.$$

...

$$\llbracket \varphi(t_5, 54) \rrbracket(s_4) = \max \dots$$

Summary / Status

- Outlined a general frame-work for quantitative analysis
- including:
 - trace inclusion / equality
 - (Bi)simulation
- Results on mutually, topologically inequivalence, and
- A generalized result on the relationship of (bi)simulation and language inclusion/equality.
- (Decidability of accumulating simulation distance for WTA).
- A characterising WCTL logic, with d_+ trace semantics. (adequate and expressive).

Future work

- Duality of operators.
- A d_1 wCTL path semantics.
- A d_{\pm} wCTL path semantics.
- Maybe $d_?$ wCTL path semantics.
- Computability and Complexity (especially lower bounds).
- Continuity w.r.t composition.
- Game characterizations.
- ...and more.