

Quantitative Verification

The Good, The Bad and The Ugly

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Nice People

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Model Checking

model

specification

Mod

\models

Spec

Quantitative Model Checking

quantitative model

quantitative specification

Mod

\models

Spec

Quantitative Model Checking

quantitative model

quantitative specification

Mod

\models

Spec

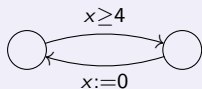
↑
not sufficient

replace by

\models_{ϵ}

Claus T: Quantitative Quantitative Quantitative Analysis

Quantitative Models



Quantitative Logics

$$\Pr_{\leq .1}(\diamond error)$$

Quantitative Verification

$$\llbracket \phi \rrbracket (s) = 3.14$$

$$d(s, t) = 42$$

Boolean world

Trace equivalence \equiv

Bisimilarity \sim

$s \sim t$ implies $s \equiv t$

$s \models \phi$ or $s \not\models \phi$

$s \sim t$ iff $\forall \phi : s \models \phi \Leftrightarrow t \models \phi$

“Quantification”

Linear distances d_L

Branching distances d_B

$d_L(s, t) \leq d_B(s, t)$

$\llbracket \phi \rrbracket (s)$ is a quantity

$d_B(s, t) = \sup_{\phi} d(\llbracket \phi \rrbracket (s), \llbracket \phi \rrbracket (t))$

Compositional Verification

model

specification

Mod

\models

Spec

- $\text{Mod} \models \text{Spec}_1 \ \& \ \text{Spec}_1 \leq \text{Spec}_2 \implies \text{Mod} \models \text{Spec}_2$
- $\text{Mod} \models \text{Spec}_1 \ \& \ \text{Mod} \models \text{Spec}_2 \implies \text{Mod} \models \text{Spec}_1 \wedge \text{Spec}_2$
- $\text{Mod}_1 \models \text{Spec}_1 \ \& \ \text{Mod}_2 \models \text{Spec}_2 \implies \text{Mod}_1 \parallel \text{Mod}_2 \models \text{Spec}_1 \parallel \text{Spec}_2$
- $\text{Mod}_1 \models \text{Spec}_1 \ \& \ \text{Mod}_2 \models \text{Spec} / \text{Spec}_1 \implies \text{Mod}_1 \parallel \text{Mod}_2 \models \text{Spec}$
- bottom-up **and** top-down

Quantitative Compositional Verification?

quantitative model

quantitative specification

Mod

\models_ε

Spec

- $\text{Mod} \models_\varepsilon \text{Spec}_1 \ \& \ \text{Spec}_1 \leq_\varepsilon \text{Spec}_2 \implies \text{Mod} \models_\varepsilon \text{Spec}_2$
- $\text{Mod} \models_\varepsilon \text{Spec}_1 \ \& \ \text{Mod} \models_\varepsilon \text{Spec}_2 \implies \text{Mod} \models_\varepsilon \text{Spec}_1 \wedge \text{Spec}_2$
- $\text{Mod}_1 \models_\varepsilon \text{Spec}_1 \ \& \ \text{Mod}_2 \models_\varepsilon \text{Spec}_2 \implies \text{Mod}_1 \parallel \text{Mod}_2 \models_\varepsilon \text{Spec}_1 \parallel \text{Spec}_2$
- $\text{Mod}_1 \models_\varepsilon \text{Spec}_1 \ \& \ \text{Mod}_2 \models_\varepsilon \text{Spec}/\text{Spec}_1 \implies \text{Mod}_1 \parallel \text{Mod}_2 \models_\varepsilon \text{Spec}$
- surely **not the same** ε everywhere!?

User Stories

“In your quantitative verification, what type of distances do you use?”

- point-wise
- accumulating
- limit-average
- discounted
- maximum-lead
- Cantor
- discrete

$$D(\sigma, \tau) = \sup_i |\sigma_i - \tau_i|$$

$$D(\sigma, \tau) = \sum_i |\sigma_i - \tau_i|$$

$$D(\sigma, \tau) = \limsup_N \frac{1}{N} \sum_{i=0}^N |\sigma_i - \tau_i|$$

$$D(\sigma, \tau) = \sum_i \lambda^i |\sigma_i - \tau_i|$$

$$D(\sigma, \tau) = \sup_N \left| \sum_{i=0}^N (\sigma_i - \tau_i) \right|$$

$$D(\sigma, \tau) = 1 / (1 + \inf \{j \mid \sigma_j \neq \tau_j\})$$

$$D(\sigma, \tau) = 0 \text{ if } \sigma = \tau; \infty \text{ otherwise}$$

Asarin-Basset-Degorre 2018

$$D(\sigma, \tau) = \max \left\{ \begin{array}{l} \sup_i \inf_j \{|t_i - s_j| \mid a_i = b_j\} \\ \sup_j \inf_i \{|t_i - s_j| \mid a_i = b_j\} \end{array} \right.$$

Challenge (ca. 2009)

- In quantitative verification, lots of different distances
- Develop theory to cover all/most of them
 - **idea:** use bisimulation games

⇒ The Quantitative Linear-Time–Branching-Time Spectrum

- QAPL 2011, FSTTCS 2011, TCS 2014

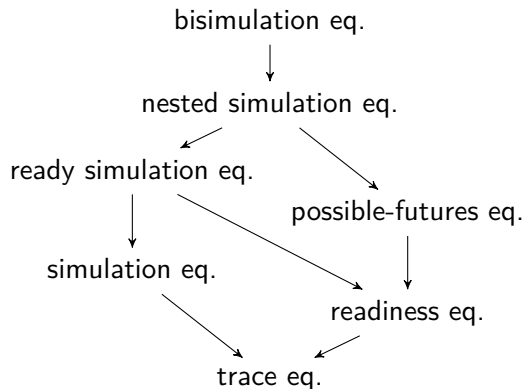
Challenge (ca. 2012):

- How to make this compositional?
- Still not satisfied!

- 1 Introduction
- 2 The Quantitative Linear-Time–Branching-Time Spectrum
- 3 Compositional Verification
- 4 Conclusion

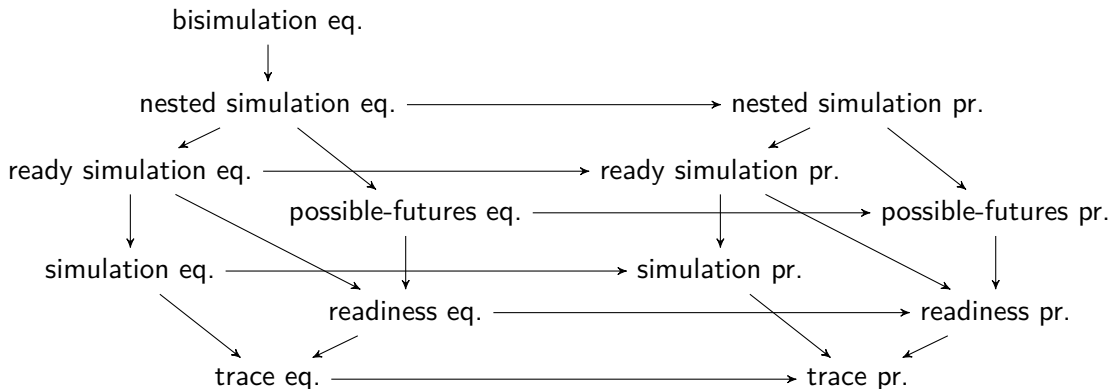
The Linear-Time–Branching-Time Spectrum

van Glabbeek 1990 (excerpt):



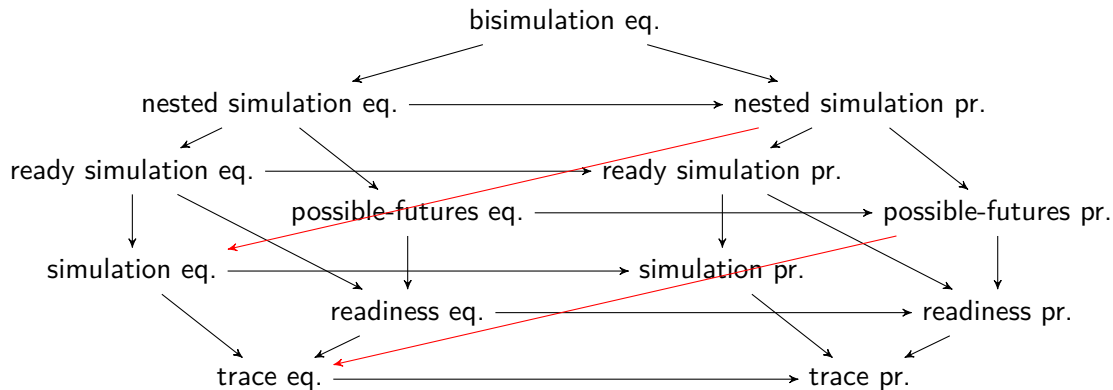
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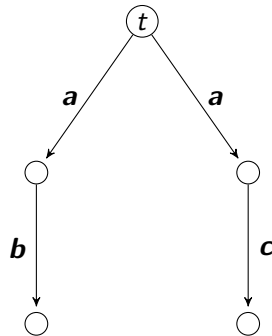
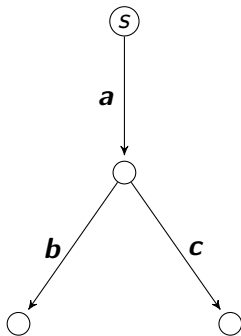


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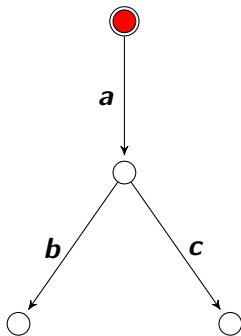


The Simulation Game

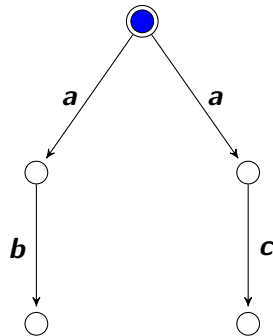


The Simulation Game

Spoiler

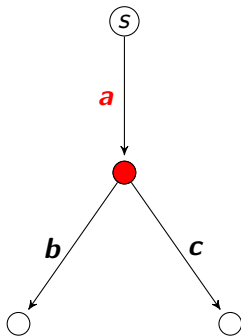


Duplicator

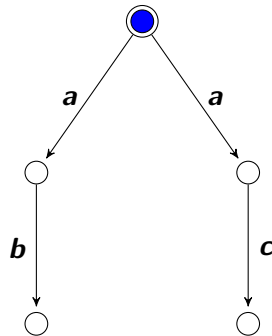


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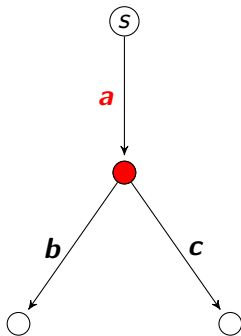


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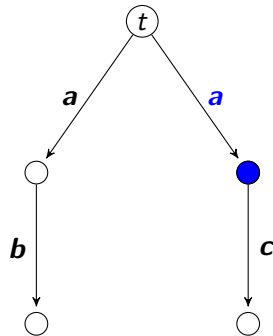


The Simulation Game

Spoiler

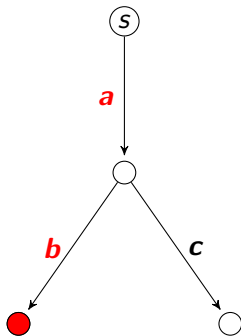


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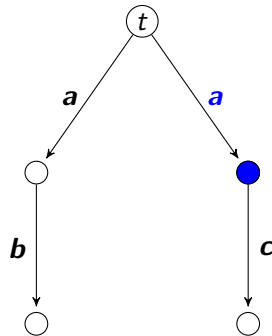


The Simulation Game

Spoiler

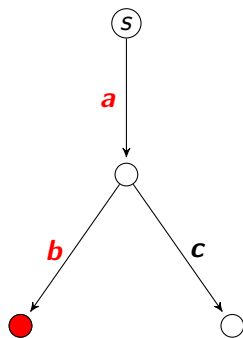


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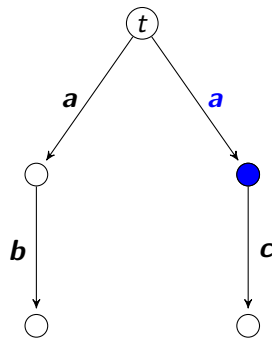


The Simulation Game

Spoiler

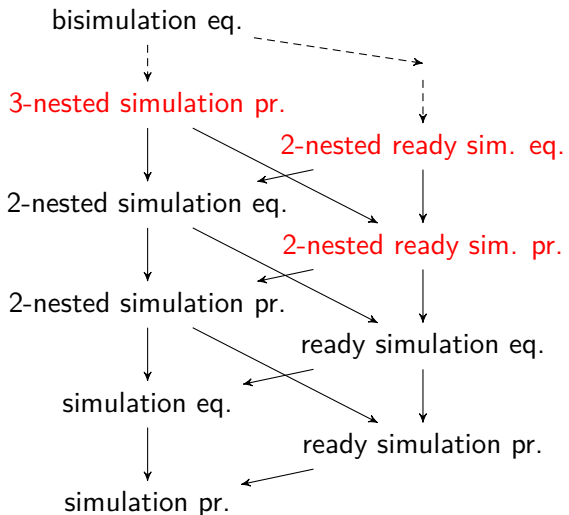


Duplicator

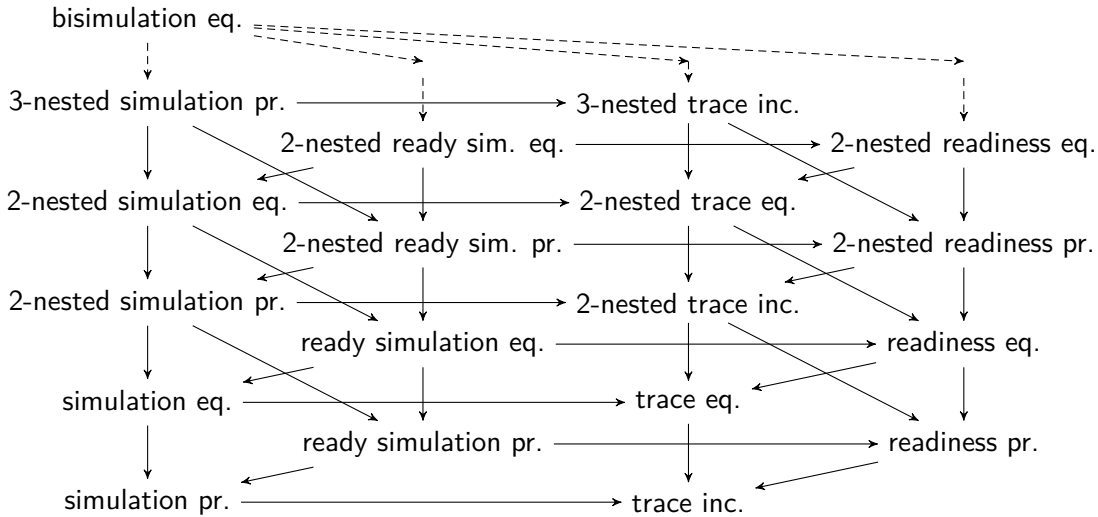


Spoiler wins

The LTBT Spectrum, Game Version



The LTBT Spectrum, Game Version



The Simulation Game, Revisited

1. Player 1 chooses edge from s (leading to s')
 2. Player 2 chooses matching edge from t (leading to t')
 3. Game continues from configuration s', t'
- ω . If Player 2 can always answer: YES, t simulates s .
Otherwise: NO

Or, as an Ehrenfeucht-Fraïssé game (“delayed evaluation”):

1. Player 1 chooses edge from s (leading to s')
 2. Player 2 chooses edge from t (leading to t')
 3. Game continues from new configuration s', t'
- ω . At the end (maybe after infinitely many rounds!), **compare the chosen traces**:
If the trace chosen by t matches the one chosen by s : YES
Otherwise: NO

Quantitative Ehrenfeucht-Fraïssé Games

The quantitative setting:

- Assume we have a way, possibly application-determined, to **measure distances** of (finite or infinite) traces
- a hemimetric $D : (\sigma, \tau) \mapsto D(\sigma, \tau) \in \mathbb{R}_{\geq 0} \cup \{\infty\}$

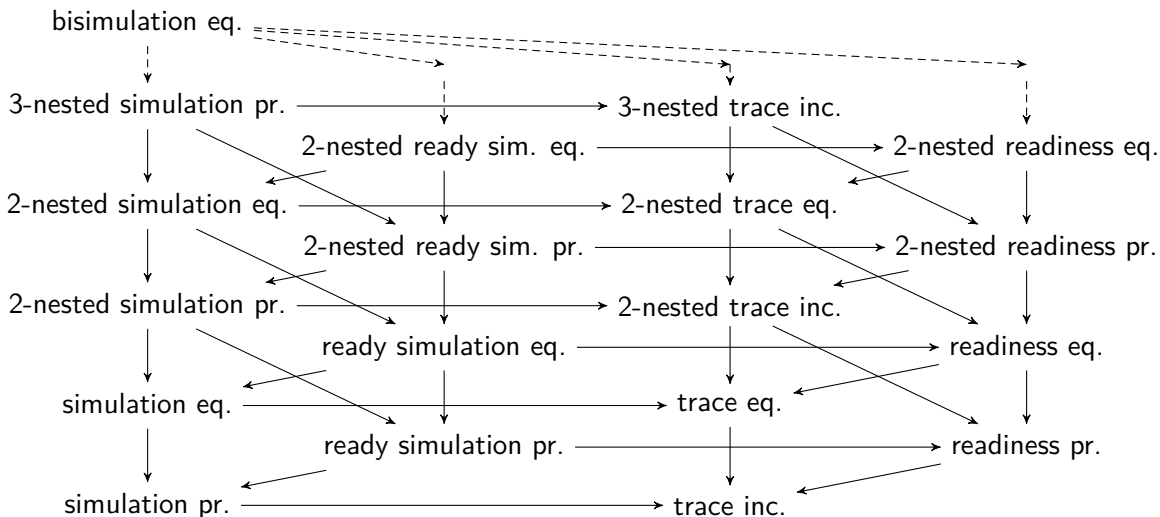
The quantitative Ehrenfeucht-Fraïssé game:

1. Player 1 chooses edge from s (leading to s')
 2. Player 2 chooses edge from t (leading to t')
 3. Game continues from new configuration s', t'
- ω . At the end, compare the chosen traces σ, τ :
- The **simulation distance** from s to t is defined to be $D(\sigma, \tau)$
- Player 1 plays to **maximize** $D(\sigma, \tau)$; Player 2 plays to **minimize**

This can be generalized to **all** the games in the LTBT spectrum.

The Quantitative Linear-Time–Branching-Time Spectrum

For any trace distance $D : (\sigma, \tau) \mapsto D(\sigma, \tau) \in \mathbb{R}_{\geq 0} \cup \{\infty\}$:



Quantitative EF Games: Some Details

- **Configuration** of the game: (π, ρ) : π the Player-1 choices up to now; ρ the Player-2 choices
- **Strategy**: mapping from configurations to next moves
 - Θ_i : set of Player- i strategies
- **Simulation** strategy: Player-1 moves allowed from **end of π**
- **Bisimulation** strategy: Player-1 moves allowed from end of π **or end of ρ**
 - (hence π and ρ are generally not paths – “**mingled paths**”)
- Pair of strategies \implies (possibly infinite) sequence of configurations
- Take the limit; unmingle \implies pair of (possibly infinite) traces (σ, τ)
- **Bisimulation distance**: $\sup_{\theta_1 \in \Theta_1} \inf_{\theta_2 \in \Theta_2} d_T(\sigma, \tau)$
- **Simulation distance**: $\sup_{\theta_1 \in \Theta_1^0} \inf_{\theta_2 \in \Theta_2} d_T(\sigma, \tau)$ (restricting Player 1's capabilities)

Quantitative EF Games: Some Details – II

- **Blind Player-1 strategies**: depend only on the **end** of ρ
 - (“cannot see Player-2 moves”)
 - $\check{\Theta}_1$: set of blind Player-1 strategies
- **Trace inclusion distance**: $\sup_{\theta_1 \in \check{\Theta}_1^0} \inf_{\theta_2 \in \Theta_2} d_T(\sigma, \tau)$
- For **nesting**: count the number of times Player 1 **switches** between end of π and end of ρ
 - Θ_1^k : k switches allowed
- **Nested simulation distance**: $\sup_{\theta_1 \in \Theta_1^1} \inf_{\theta_2 \in \Theta_2} d_T(\sigma, \tau)$
- **Nested trace inclusion distance**: $\sup_{\theta_1 \in \check{\Theta}_1^1} \inf_{\theta_2 \in \Theta_2} d_T(\sigma, \tau)$ (!)
- For **ready**: allow extra “I’ll see you” Player-1 transition from end of ρ

Transfer Theorem

Theorem

*If two equivalences or preorders are **inequivalent** in the **qualitative** setting, and the trace distance D is **separating**, then the corresponding QLTBT distances are **topologically inequivalent**.*

Recursive Characterization

Theorem

If the trace distance $D : (\sigma, \tau) \mapsto d(\sigma, \tau)$ has a decomposition $d = g \circ f : \text{Tr} \times \text{Tr} \rightarrow L \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$ through a complete lattice L , and f has a **recursive characterization**, i.e., such that $f(a.\sigma, b.\tau) = F(a, b, f(\sigma, \tau))$ for some $F : \Sigma \times \Sigma \times L \rightarrow L$ which is **monotone** in the third coordinate, then all distances in the corresponding QLTBT spectrum are given as **least fixed points** of some functionals using F .

All trace distances I know can be expressed recursively like this.

- Example: simulation distance:

$$d_{\text{sim}}(s, t) = \sup_{s \xrightarrow{a} s'} \inf_{t \xrightarrow{b} t'} F(a, b, d_{\text{sim}}(s', t')) \quad (\text{l.f.p.})$$

- L is “memory”
- also gives **relation family** characterization

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Specification Theories

Let **Mod** be a set of models with an equivalence \sim .

Definition

A **complete specification theory** for (Mod, \sim) is $(\text{Spec}, \leq, \parallel, \chi)$ such that

- \leq is a **refinement** preorder on Spec
- $\chi : \text{Mod} \rightarrow \text{Spec}$ picks out **characteristic specifications**
 - i.e., $\forall \mathcal{M}_1, \mathcal{M}_2 \in \text{Mod} : \mathcal{M}_1 \sim \mathcal{M}_2 \iff \chi(\mathcal{M}_1) \leq \chi(\mathcal{M}_2)$
- $(\text{Spec}, \leq, \parallel)$ forms a **bounded commutative distributive residuated lattice up to $\leq \cap \geq$**

$\Rightarrow \vee$ and \wedge on Spec; double distributivity; $\perp, \top \in \text{Spec}$

- everything **up to modal equivalence** $\equiv \equiv \leq \cap \geq$

$\Rightarrow \parallel$ distributes over \vee , has unit \top , has residual $/$ (up to \equiv)

- $\mathcal{S}_1 \parallel \mathcal{S}_2 \leq \mathcal{S}_3 \iff \mathcal{S}_2 \leq \mathcal{S}_3 / \mathcal{S}_1$

Examples

- Disjunctive modal transition systems
- Acceptance automata
- Hennessy-Milner logic with maximal fixed points
- CONCUR 2013, ICTAC 2014, I&C 2020 (all with **bisimulation** as model equivalence \sim)

Acceptance Automata

Let Σ be a finite alphabet.

Definition

A (nondeterministic) **acceptance automaton** (AA) is a structure $\mathcal{A} = (S, S^0, \text{Tran})$, with $S \supseteq S^0$ finite sets of states and initial states and $\text{Tran} : S \rightarrow 2^{2^{\Sigma \times S}}$ an assignment of *transition constraints*.

- standard labeled transition system (**LTS**): $\text{Tran} : S \rightarrow 2^{\Sigma \times S}$ (**coalgebraic** view)
- (for AA:) $\text{Tran}(s) = \{M_1, M_2, \dots\}$: **provide M_1 or M_2 or ...**
- a **disjunctive** choice of **conjunctive** constraints
- [J.-B. Raclet 2008](#) (but deterministic); see also [H. H. Hansen 2003](#)
- note multiple initial states

Refinement

Definition

Let $\mathcal{A}_1 = (S_1, S_1^0, \text{Tran}_1)$ and $\mathcal{A}_2 = (S_2, S_2^0, \text{Tran}_2)$ be AA.

A relation $R \subseteq S_1 \times S_2$ is a **modal refinement** if:

- ① $\forall s_1^0 \in S_1^0 : \exists s_2^0 \in S_2^0 : (s_1^0, s_2^0) \in R$ (init)
- ② $\forall (s_1, s_2) \in R : \forall M_1 \in \text{Tran}_1(s_1) : \exists M_2 \in \text{Tran}_2(s_2) :$ (tran)
 - ① $\forall (a, t_1) \in M_1 : \exists (a, t_2) \in M_2 : (t_1, t_2) \in R$
 - ② $\forall (a, t_2) \in M_2 : \exists (a, t_1) \in M_1 : (t_1, t_2) \in R$

Write $\mathcal{A}_1 \leq \mathcal{A}_2$ if there exists such a modal refinement.

- for any **constraint choice** M_1 there is a **bisimilar** choice M_2
- \mathcal{A}_1 has **fewer choices** than \mathcal{A}_2
- no more choices $\hat{=}$ only one $M \in \text{Tran}(s) \hat{=}$ LTS
- formally: an **embedding** $\chi : \text{LTS} \hookrightarrow \text{AA}$
such that $\chi(\mathcal{L}_1) \leq \chi(\mathcal{L}_2)$ iff \mathcal{L}_1 and \mathcal{L}_2 are **bisimilar**

Logical Operations

Let $\mathcal{A}_1 = (S_1, S_1^0, \text{Tran}_1)$ and $\mathcal{A}_2 = (S_2, S_2^0, \text{Tran}_2)$ be AA.

Disjunction: $\mathcal{A}_1 \vee \mathcal{A}_2 = (S_1 \dot{\cup} S_2, S_1^0 \dot{\cup} S_2^0, \text{Tran}_1 \dot{\cup} \text{Tran}_2)$

Conjunction: define $\pi_i : 2^{\Sigma \times S_1 \times S_2} \rightarrow 2^{\Sigma \times S_i}$ by

$$\pi_1(M) = \{(a, s_1) \mid \exists s_2 \in S_2 : (a, s_1, s_2) \in M\}$$

$$\pi_2(M) = \{(a, s_2) \mid \exists s_1 \in S_1 : (a, s_1, s_2) \in M\}$$

Let $\mathcal{A}_1 \wedge \mathcal{A}_2 = (S_1 \times S_2, S_1^0 \times S_2^0, \text{Tran})$ with

$$\text{Tran}((s_1, s_2)) = \{M \subseteq \Sigma \times S_1 \times S_2 \mid \pi_1(M) \in \text{Tran}_1(s_1), \pi_2(M) \in \text{Tran}_2(s_2)\}$$

Theorem

For all LTS \mathcal{L} and AA $\mathcal{A}_1, \mathcal{A}_2$:

$$\mathcal{L} \models \mathcal{A}_1 \vee \mathcal{A}_2 \iff \mathcal{L} \models \mathcal{A}_1 \text{ or } \mathcal{L} \models \mathcal{A}_2$$

$$\mathcal{L} \models \mathcal{A}_1 \wedge \mathcal{A}_2 \iff \mathcal{L} \models \mathcal{A}_1 \ \& \ \mathcal{L} \models \mathcal{A}_2$$

Structural Operations: Composition

Let $\mathcal{A}_1 = (S_1, S_1^0, \text{Tran}_1)$ and $\mathcal{A}_2 = (S_2, S_2^0, \text{Tran}_2)$ be AA.

For $M_1 \subseteq \Sigma \times S_1$ and $M_2 \subseteq \Sigma \times S_2$, define

$$M_1 \parallel M_2 = \{(a, (t_1, t_2)) \mid (a, t_1) \in M_1, (a, t_2) \in M_2\}$$

(assumes CSP synchronization, but can be generalized)

Let $\mathcal{A}_1 \parallel \mathcal{A}_2 = (S_1 \times S_2, S_1^0 \times S_2^0, \text{Tran})$ with

$$\text{Tran}((s_1, s_2)) = \{M_1 \parallel M_2 \mid M_1 \in \text{Tran}_1(s_1), M_2 \in \text{Tran}_2(s_2)\}$$

Theorem (independent implementability)

For all AA $\mathcal{A}_1, \mathcal{A}_2, \mathcal{A}_3, \mathcal{A}_4$:

$$\mathcal{A}_1 \leq \mathcal{A}_3 \ \& \ \mathcal{A}_2 \leq \mathcal{A}_4 \implies \mathcal{A}_1 \parallel \mathcal{A}_2 \leq \mathcal{A}_3 \parallel \mathcal{A}_4$$

Structural Operations: Quotient

Let $\mathcal{A}_1 = (S_1, S_1^0, \text{Tran}_1)$ and $\mathcal{A}_2 = (S_2, S_2^0, \text{Tran}_2)$ be AA.

Define $\mathcal{A}_1/\mathcal{A}_2 = (S, S^0, \text{Tran})$:

- $S = 2^{S_1 \times S_2}$
- write $S_2^0 = \{s_2^{0,1}, \dots, s_2^{0,p}\}$ and let $S^0 = \{(s_1^{0,q}, s_2^{0,q}) \mid q \in \{1, \dots, p\}\} \mid \forall q : s_1^{0,q} \in S_1^0\}$
- $\text{Tran} =$

Structural Operations: Quotient

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- $\text{Tran} =$



Structural Operations: Quotient

Let $\mathcal{A}_1 = (S_1, S_1^0, \text{Tran}_1)$ and $\mathcal{A}_2 = (S_2, S_2^0, \text{Tran}_2)$ be AA.

Define $\mathcal{A}_1 / \mathcal{A}_2 = (S, S^0, \text{Tran})$:

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- write $S_2^0 = \{s_2^{0,1}, \dots, s_2^{0,p}\}$ and let $S^0 = \{\{(s_1^{0,q}, s_2^{0,q}) \mid q \in \{1, \dots, p\}\} \mid \forall q : s_1^{0,q} \in S_1^0\}$
- $\text{Tran} = \dots$

Theorem

For all AA $\mathcal{A}_1, \mathcal{A}_2, \mathcal{A}_3$:

$$\mathcal{A}_1 \parallel \mathcal{A}_2 \leq \mathcal{A}_3 \iff \mathcal{A}_2 \leq \mathcal{A}_3 / \mathcal{A}_1$$

- up to \equiv , $/$ is the **adjoint** (or **residual**) of \parallel

Quantitative Specification Theories?

Definition (recall)

A **complete specification theory** for (Mod, \sim) is $(\text{Spec}, \leq, \parallel, \chi)$ such that

- \leq is a **refinement** preorder on Spec
 - $\mathcal{M}_1 \sim \mathcal{M}_2 \iff \chi(\mathcal{M}_1) \leq \chi(\mathcal{M}_2)$
 - $(\text{Spec}, \leq, \parallel)$ forms a **b.c.d. residuated lattice up to \equiv**
-
- generalize \sim by **pseudometric** d_{Mod}
 - $d_{\text{Mod}}(\mathcal{M}_1, \mathcal{M}_2) = 0$ iff $\mathcal{M}_1 \sim \mathcal{M}_2$
 - generalize \leq by **hemimetric** d
 - $d_{\text{Mod}}(\mathcal{M}_1, \mathcal{M}_2) = d(\chi(\mathcal{M}_1), \chi(\mathcal{M}_2))$
 - $d(\mathcal{M}, \mathcal{S}) = d(\chi(\mathcal{M}), \mathcal{S})$
 - still want $(\text{Spec}, \leq, \parallel)$ to be a b.c.d. residuated lattice up to \equiv

Acceptance Automata

For DMTS/AA/HML_{max}:

- d_{Mod} : any bisimulation distance
- d : corresponding modal refinement distance
- transitivity \rightsquigarrow triangle ineq.: $d(\mathcal{S}_1, \mathcal{S}_2) + d(\mathcal{S}_2, \mathcal{S}_3) \geq d(\mathcal{S}_1, \mathcal{S}_3)$
- $d(\mathcal{S}, \mathcal{S}_1 \wedge \mathcal{S}_2) = \max(d(\mathcal{S}, \mathcal{S}_1), d(\mathcal{S}, \mathcal{S}_2))$ or ∞
- $d(\mathcal{S}_1 \vee \mathcal{S}_2, \mathcal{S}) = \max(d(\mathcal{S}_1, \mathcal{S}), d(\mathcal{S}_2, \mathcal{S}))$ or ∞
- quotient is quantitative residual: $d(\mathcal{S}_1 \parallel \mathcal{S}_2, \mathcal{S}_3) = d(\mathcal{S}_2, \mathcal{S}_3 / \mathcal{S}_1)$
- for \parallel itself, **uniform continuity**: a function $P : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ such that $d(\mathcal{S}_1 \parallel \mathcal{S}_2, \mathcal{S}_3 \parallel \mathcal{S}_4) \leq P(d(\mathcal{S}_1, \mathcal{S}_3), d(\mathcal{S}_2, \mathcal{S}_4))$

**The Bad
and/or
Ugly**

Silent Moves in QLTBT?

- Any serious spectrographer needs to think about **silent moves**
- (van Glabbeek 1993: LTBT II)
- Bisping, Jansen 2023: Energy games for the weak spectrum
 - but uses **power set** for linear part (recall: we use **blindness** instead)
 - difficult to reconcile power set with quantitative setting
- otherwise, some **coalgebra** approaches:
 - Sprunger, Katsumata, Dubut, Hasuo 2021: Fibrational bisimulations and quantitative reasoning
 - Ford, Milius, Schröder, Beohar, König 2022: Graded monads and behavioural equivalence games
 - Beohar, Gurke, König, Messing 2023: Hennessy-Milner theorems via Galois connections
 - again, **power set** seems very popular . . .
- status: IT'S COMPLICATED

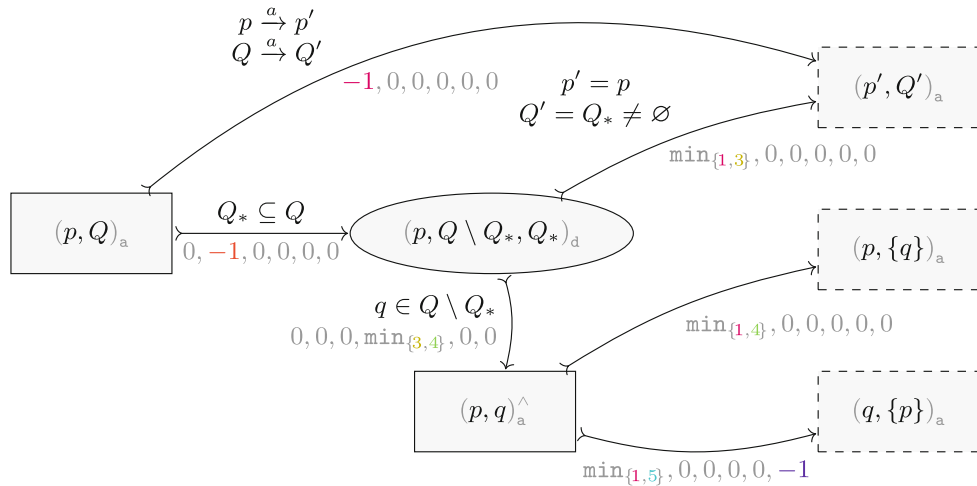


Fig. 7. Schematic spectroscopy game \mathcal{G}_Δ of Definition 10.

Asarin-Basset-Degorre Distance

Recall:

$$D(\sigma, \tau) = \max \left\{ \begin{array}{l} \sup_i \inf_j \{|t_i - s_j| \mid a_i = b_j\} \\ \sup_j \inf_i \{|t_i - s_j| \mid a_i = b_j\} \end{array} \right.$$

Asarin-Basset-Degorre Distance

On the practical side, if we observed timed words with some finite precision (say 0.01s), then it would be difficult to distinguish the order of close events, e.g. detect the difference between

$$w_1 = (a, 1), (b, 2), (c, 2.001) \text{ and } w_2 = (a, 1.001), (c, 1.999), (b, 2.001).$$

Moreover, it is even difficult to count the number of events that happen in a short lapse of time, e.g. the words w_1, w_2 look very similar to

$$w_3 = (a, 1), (c, 1.999), (c, 2), (b, 2.001), (c, 2.0002).$$

A slow observer, when receiving timed words w_1, w_2, w_3 will just sense an a at the date ≈ 1 and b and c at the date ≈ 2 .

As the main contribution of this paper, we introduce a metric on timed words (with non-fixed number of events) for which w_1, w_2, w_3 are very close to each

Asarin-Basset-Degorre Distance

Recall:

$$D(\sigma, \tau) = \max \left\{ \begin{array}{l} \sup_i \inf_j \{|t_i - s_j| \mid a_i = b_j\} \\ \sup_j \inf_i \{|t_i - s_j| \mid a_i = b_j\} \end{array} \right.$$

- takes into account permutations of symbols which are close in timing
- but in a way which may lose symbols
- relation to **timed pomsets**? [Amrane, Bazille, Clement, UF 2024: Languages of HDTA](#)
- status: HOPEFUL

Robustness

A quantitative system is **robust** if

$$\left. \begin{array}{l} \text{small changes in inputs} \\ \text{small environment perturbations} \\ \text{small measuring errors} \end{array} \right\} \implies \text{small changes in behavior}$$

Formulate using **uniform continuity**: there is a constant K such that

$$d_{\text{behavior}}(S, S') \leq K d_{\text{syntax}}(S, S')$$

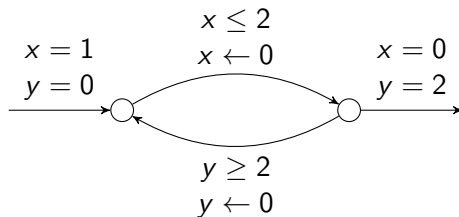
for all perturbations S' of S .

- Standard formulation in control theory
- Generally want systems to be robust

Robustness, lack of

Our quantitative models are **not robust**:

(Merci, Nico)



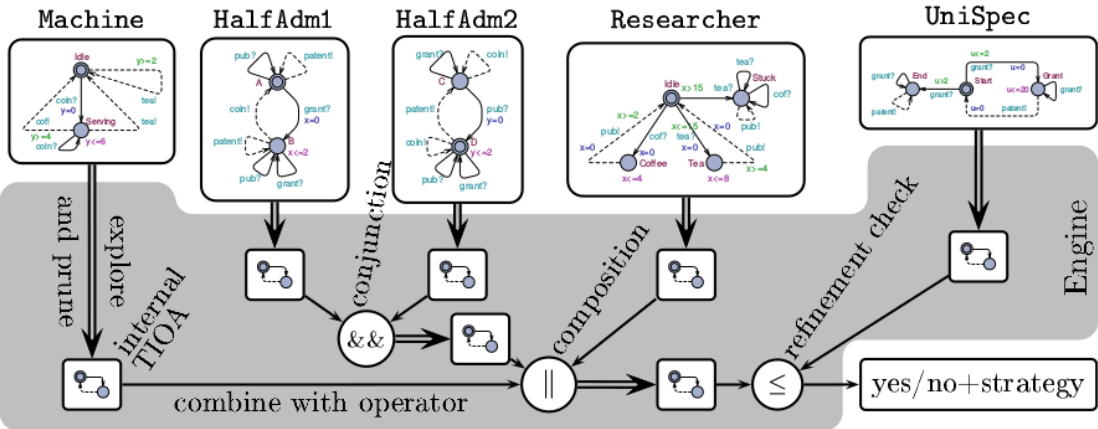
- restrict to robust models?
- quantify robustness?
- use different models?
- ...

Compositionality?

Timed input-output automata:

- David, Larsen, Legay, Nyman, Traonouez, Wąsowski 2015: Real-time specifications
- Goorden, Larsen, Legay, Lorber, Nyman, Wąsowski 2023: Timed I/O Automata: It is never too late to complete your timed specification theory
- complete, with quotient, but without disjunction
- only **deterministic** specifications
- tool support: **ECDAR** / **Uppaal TiGa** (Aalborg)
- some work on **robustness** and **implementability**: Larsen, Legay, Traonouez, Wąsowski 2014: Robust synthesis for real-time systems

Timed Input-Output Automata



Specification Theories for Real-Time Systems, contd.

Modal **event-clock** specifications:

- Bertrand, Legay, Pinchinat, Raclet 2012: Modal event-clock specifications for timed component-based design
- complete, with quotient, but without disjunction
- only **deterministic** specifications
- some work on **robustness**: UF, Legay 2012: A robust specification theory for modal event-clock automata

Synchronous time-triggered interface theories:

- Delahaye, UF, Henzinger, Legay, Ničković 2012: Synchronous interface theories and time triggered scheduling
- no quotient, dubious conjunction, no implementation
- relation to **BIP**

Specification Theories for Hybrid Systems

Specification Theories for Hybrid Systems

- Quesel, Fränzle, Damm 2011: Crossing the bridge between similar games

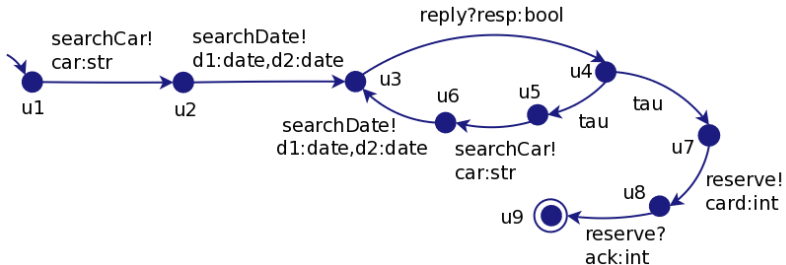
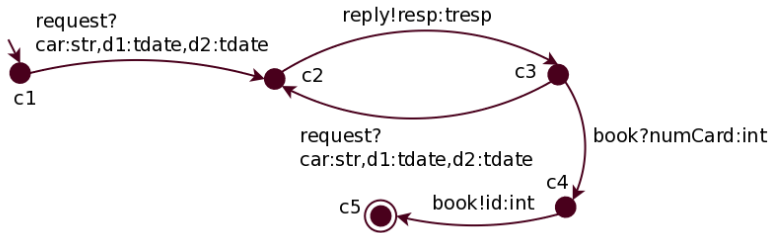
Conclusion

- general theory of quantitative verification
- general theory of **compositional** quantitative verification
 - algebraic properties
 - quantitative algebraic properties
 - silent moves
- for real-time systems
 - robustness
 - compositionality
 - robust compositionality
- for hybrid systems

✓
 $\neg \backslash_{(\ominus)} _ / \neg$
 ✓
 ✗
 ✗
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 ✗

5 Applications

Interface Compatibility

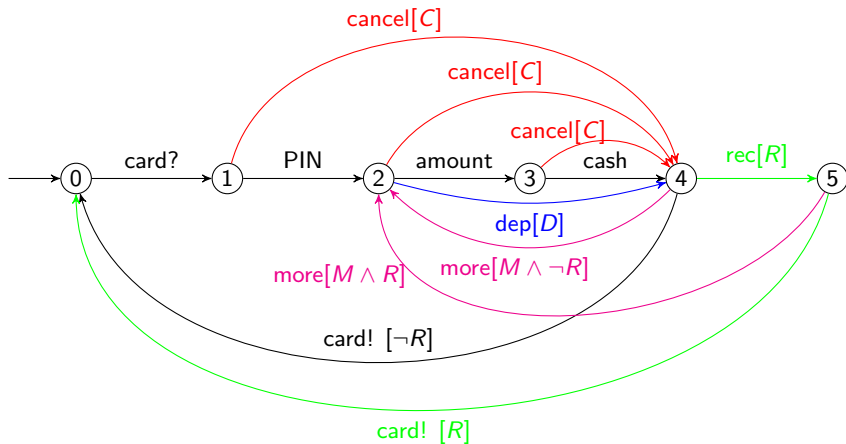


Interface Compatibility

- Use discounted (bi)simulation distances for measuring interface compatibility
- With A. Legay, M. Ouederni, G. Salaün
 - bisimulation d. for symmetric compatibility
 - ready simulation d. for asymmetric compatibility
- With tool support:

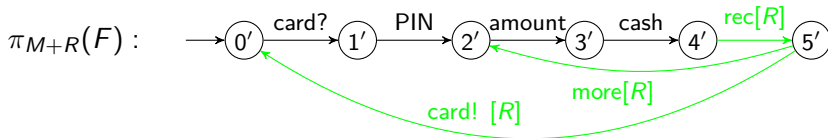
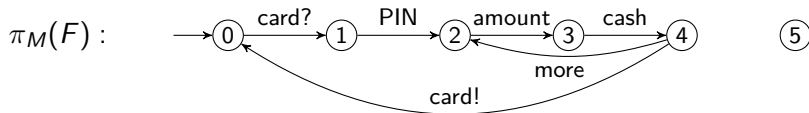
Results of the Compatibility Measure					
Example	medicalSystem				
STSs	MedServer.xml Client.xml				
Compatibility Notion	Unspecified Receptions				
Compatibility Matrix		c1	c2	c3	c4
	s1	1.0	0.06	0.01	0.01
	s2	0.05	1.0	0.35	0.01
	s3	0.01	0.26	1.0	0.01
	s4	0.01	0.01	0.01	1.0
	s5	0.01	0.26	0.64	0.01
Global Compatibility	1				
Mismatches	Download				

Behaviour Interactions in Product Lines



Behaviour Interactions in Product Lines

- Use a variant of Cantor bisimulation distance for counting the number of behaviour interactions in feature transition systems
- With J. Atlee, S. Beidu, A. Legay
- Use projections to products and compute Cantor bisimulation distance without repetitions:



Inter-Textual Distances in Statistical NLP

- Use discounted bisimulation distance to measure differences between texts
- With F. Biondi, S. Kongshøj, A. Legay
 - texts are very simple transition systems!
- Implementation
- Works better for some cases than standard distances used in statistical natural language processing
- New collaboration with NLP people in Grenoble

Inter-Textual Distances in Statistical NLP

- Let $A = (a_1, a_2, \dots, a_{N_A})$ and $B = (b_1, b_2, \dots, b_{N_B})$ be texts
- Write $\delta_{i,j} = [\text{if } a_i = b_j \text{ then } 0 \text{ else } 1]$ (**word match indicator**)
- **position match** (λ : discounting factor, $0 \leq \lambda < 1$):

$$d_{\text{pm}}(i, j) = \delta_{i,j} + \lambda\delta_{i+1,j+1} + \lambda^2\delta_{i+2,j+2} + \lambda^3\delta_{i+3,j+3} + \dots$$

- “try to match n -grams for n as high as possible, but don't be too sad if very long phrases don't match”
- **global distance**:

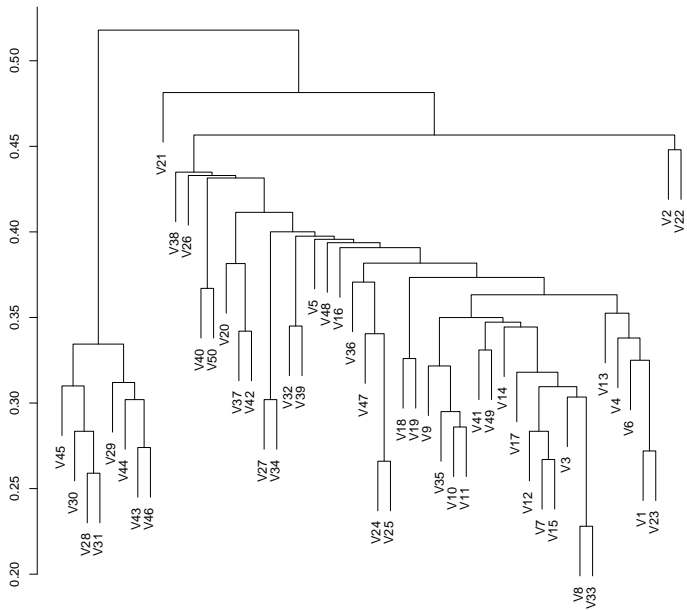
$$d_3(A, B) = \frac{1 - \lambda}{N_A} \sum_{i=1}^{N_A} \min_{j=1, \dots, N_B} d_{\text{pm}}(i, j)$$

- find best possible match for **each position** in A , average, and scale
- and symmetrize: $d_4(A, B) = \max(d_3(A, B), d_3(B, A))$

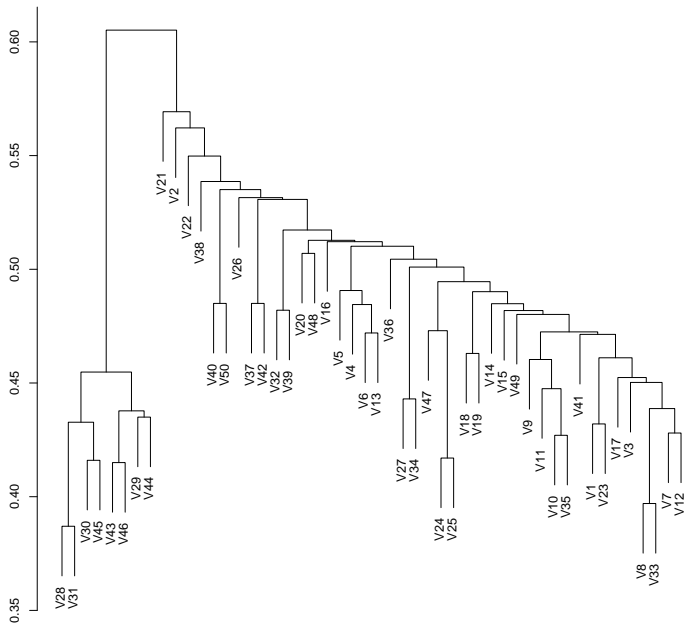
Inter-Textual Distances in Statistical NLP

Experiment:

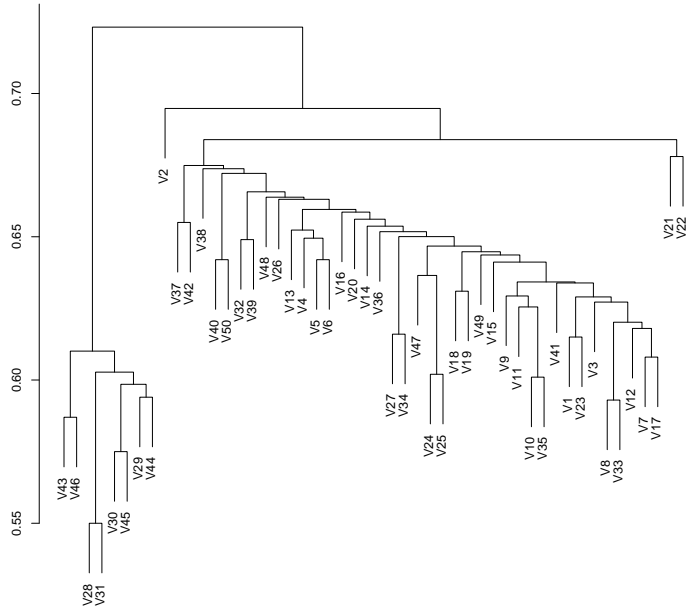
- 42 genuine scientific papers, all about modeling and verification
- 8 automatically generated “fake” papers (“SciGen”)



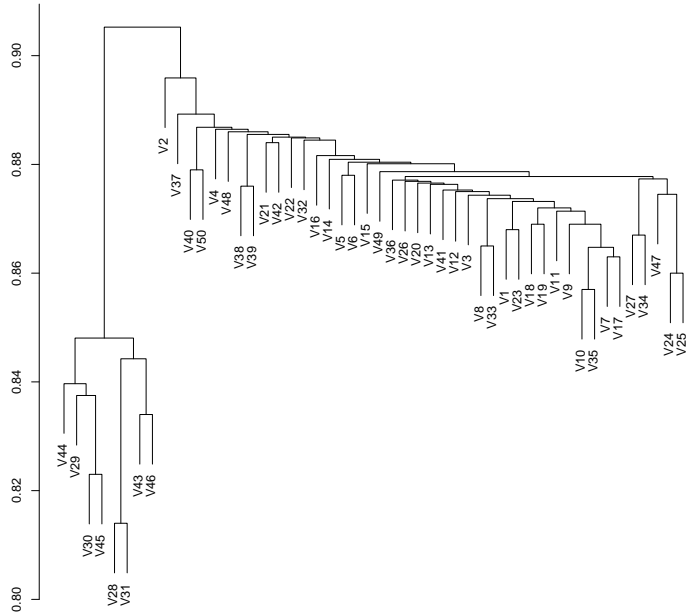
$$\lambda = 0$$



$$\lambda = .4$$



$\lambda = .7$

 $\lambda = .95$