



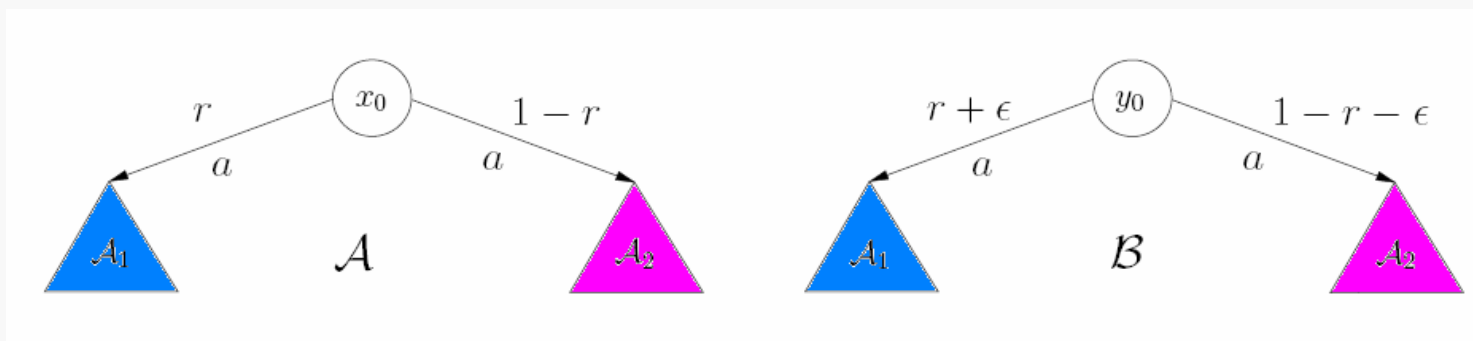
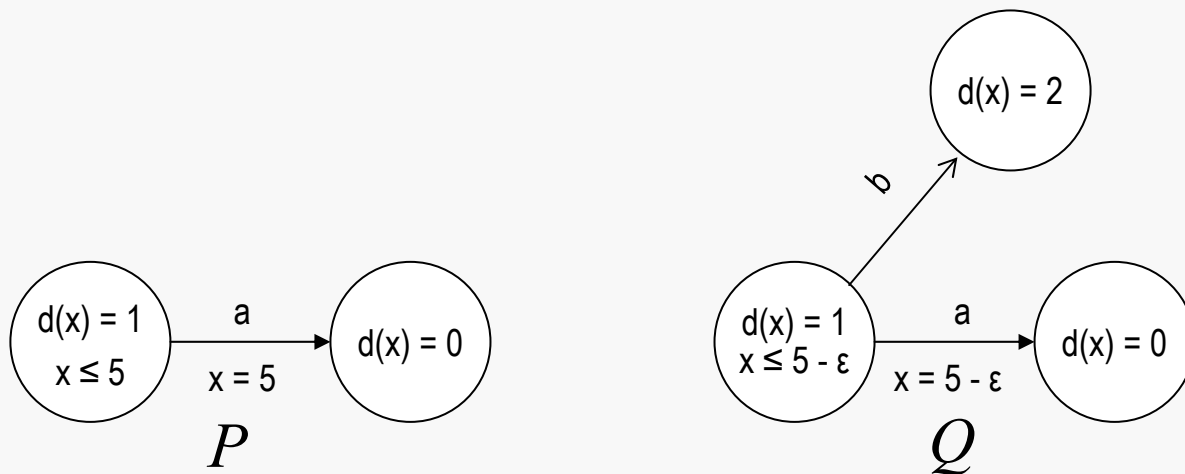
Approximate Simulations for Task-Structured Probabilistic I/O Automata

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Implementation

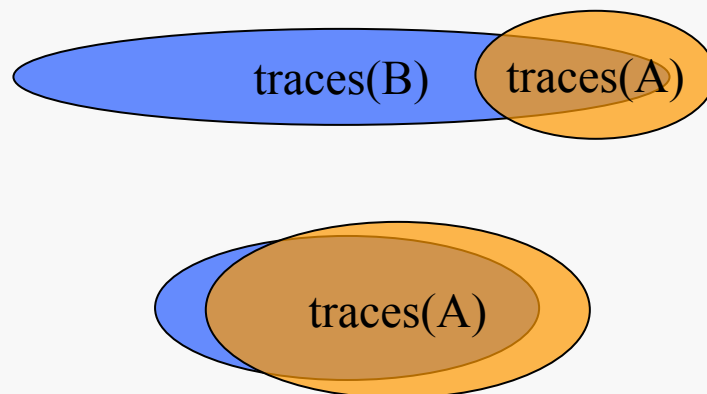
- Implementation: a fundamental notion in concurrency theory
- “traces” or observable behavior, e.g. sequence of events, timing of events, probabilities
- \mathbf{A} implements \mathbf{B} if $traces(\mathbf{A}) \subseteq traces(\mathbf{B})$
- \mathbf{A} is equivalent to \mathbf{B} if they implement each other, i.e., $traces(\mathbf{A}) = traces(\mathbf{B})$

Fragility of equality



Unequal, but similar

- A metric \mathbf{d} on the space \mathbf{T} of traces of \mathbf{A} (and \mathbf{B})
- (\mathbf{T}, \mathbf{d}) is a metric space
- \mathbf{A} *approximately implements* \mathbf{B} if the one-sided Hausdorff distance from $\text{traces}(\mathbf{A})$ to $\text{traces}(\mathbf{B})$ is small.



Previously

- Metric-based approximate simulations and bisimulations
 - PIOA [**Jou and Smolka '90**]
 - Labelled Markov Processes [**Desharnais, et. al. '04**] [**Breugel, Mislove '03**]
 - Hybrid Systems [**Girard, Julius, Pappas '05**]
 - GSMP [**Gupta, Jagadeesan, Panagaden '04**]
 - Linear stochastic hybrid automata [**Julius '06**]

- This talk: approximate implementations & simulations for task-PIOAs
 - Probabilistic and nondeterministic choices
 - Simulation relation on distributions of executions (not states)
 - Metric on trace distributions
state space / space of external actions need not be metric spaces.

Outline

- Background
- Task PIOA vocabulary
- Definitions: metrics and simulations
- Soundness (sketch)
- Discussions
 - Generalization
 - Applications
 - Future directions

Task PIOA

$A = (Q, \nu, A, D, R)$ [Segala'96] [Canetti, et. al. 2006]

- Countable set of states Q
- Initial distribution on states ν
- Countable set of actions $A = I \cup O \cup H$
 - If $I = \emptyset$ then A is *closed*
 - $O \cup H$ set of *locally controlled* actions
- Set of (q, a, μ) transitions D
- An equivalence R relation on locally controlled actions
 - Each equivalence class of R is a *task*

- Input enabled
- For any action there is at most one transition
- For any task T , there is at most one enabled action in T

Nondeterministic choice over tasks.

Task PIOA Vocabulary

- Execution fragment $\alpha = q_0 a_1 q_1 a_2 \dots$
- α is an execution if q_0 in $\text{supp}(v)$
- $\text{trace}(\alpha)$ is obtained by deleting all q 's and the a 's in H .
 - trace is a measurable function

- Task scheduler σ is a sequence of tasks $T_1 T_2 T_3 \dots$
- **apply** (μ, σ) = probability distribution over fragments

- $\text{tdist}(\mu)$ is the image measure of μ under the trace function
- $\text{tdists}(A) = \{\text{tdist}(\text{apply}(v, \sigma)) : \sigma \text{ is a task scheduler for } A\}$

Previously in PIOA: Exact implementations

In [Canetti, et. al. 2006] the $tdists(A) \subseteq tdists(B)$ exact implementation relation used for verifying an Oblivious Transfer protocol.

- $R \subseteq Disc(Execs^*(A)) \times Disc(Execs^*(B))$ is a simulation relation if:
 - $\mu_1 R \mu_2$ implies $tdist(\mu_1) = tdist(\mu_2)$
 - $v_1 R v_2$
 - If $\mu_1 R \mu_2$, then $\mu_1' \mathcal{E}(R) \mu_2'$.

$\mathcal{E}(R)$ is defined using lifting and flattening

Approximate implementations

- Uniform metric on trace distributions

$$\mathbf{d}_u(\mu_1, \mu_2) = \sup_{C \in F_{Traces}} |\mu_1(C) - \mu_2(C)|$$

- A δ -implements B if for every μ_1 there is a μ_2 with $\mathbf{d}_u(\mu_1, \mu_2) \leq \delta$

$$\phi(Disc(Exec^*(A)) \times Disc(Exec^*(B))) \rightarrow \mathfrak{R}_{\geq 0} \cup \{\infty\}$$

is (ε, δ) -approximate simulation function if:

$$\text{Start} : \phi(v_1, v_2) \leq \varepsilon$$

$$\text{Step} : \phi(\mu_1, \mu_2) \leq \varepsilon \text{ implies } \widehat{\phi}(\mu_1', \mu_2') \leq \varepsilon$$

$$\text{Trace} : \phi(\mu_1, \mu_2) \leq \varepsilon \text{ implies } \mathbf{d}_u(\text{tdist}(\mu_1), \text{tdist}(\mu_2)) \leq \delta$$

Phi and Phi Hat

Given $\phi(X, Y) \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$

$$\hat{\phi}(x_1, y_1) = \min_{\substack{\Gamma \in D(X \times Y) \\ x_1 = \mathbf{E}[\Gamma_x] \\ y_1 = \mathbf{E}[\Gamma_y]}} \left[\max_{(x, y) \in \text{supp}(\Gamma)} \phi(x, y) \right]$$

$\hat{\phi}(x_1, y_1) \leq \varepsilon \Leftrightarrow \exists$ a witnessing joint distribution $\psi \in D(X \times Y)$

such that $\max_{x, y \in \text{supp}(\psi)} \phi(x, y) \leq \varepsilon$

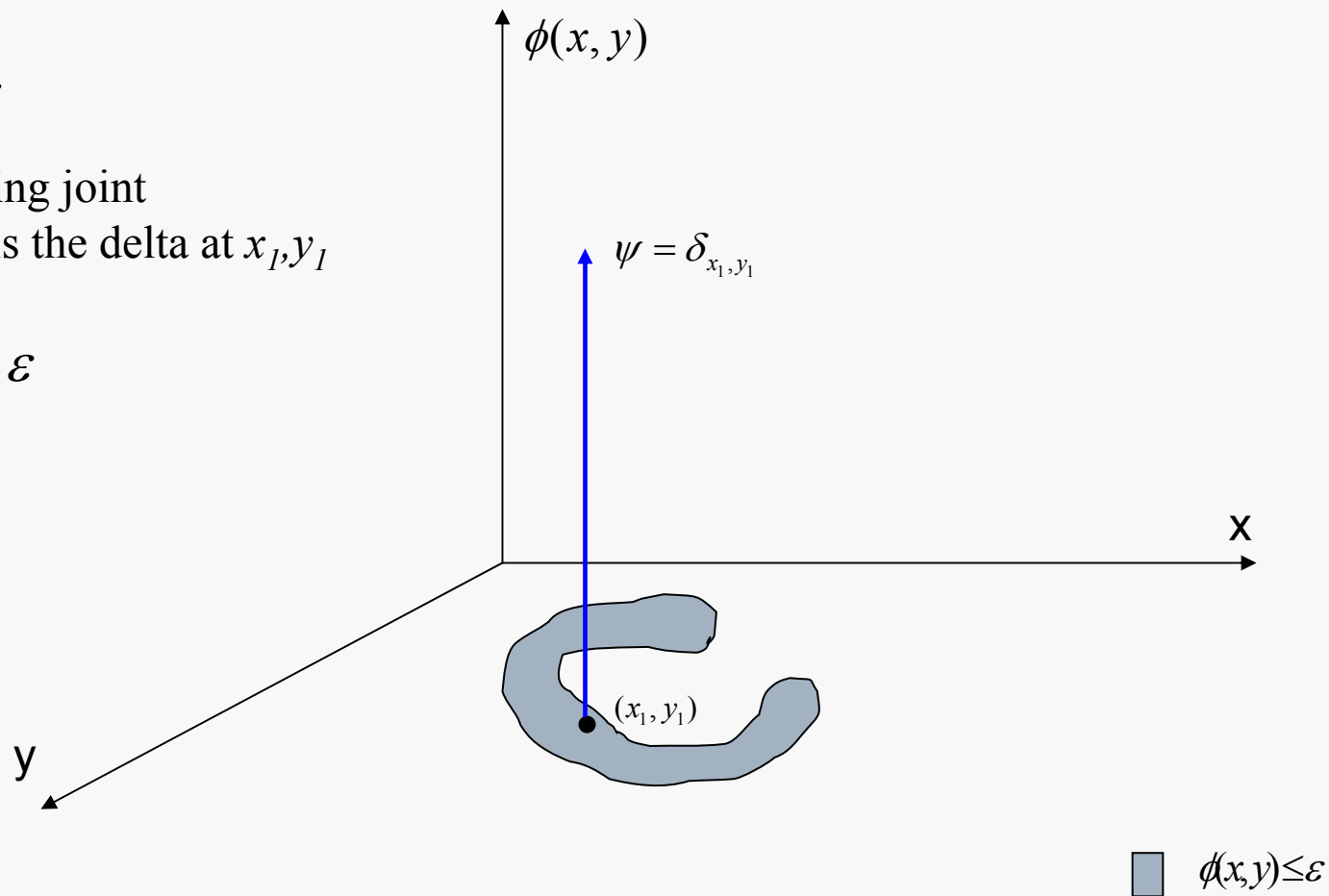
$$x_1 = \sum_{x, y} \psi(x, y) x \quad \text{and} \quad y_1 = \sum_{x, y} \psi(x, y) y$$

Expansion

$$\phi(x_1, y_1) \leq \varepsilon$$

One witnessing joint distribution is the delta at x_1, y_1

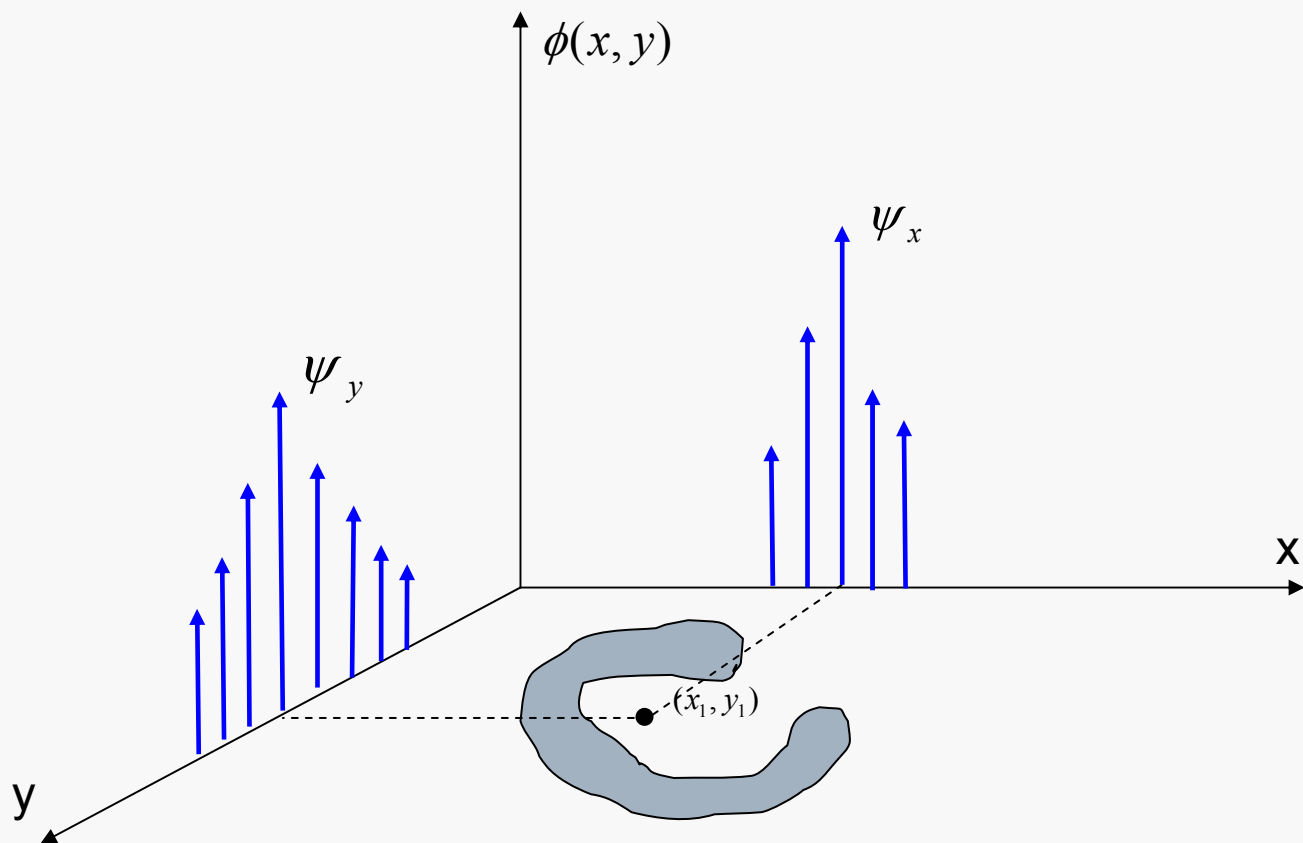
$$\hat{\phi}(x_1, y_1) \leq \varepsilon$$



Expansion

$$\phi(x, y) > \varepsilon$$

$$\widehat{\phi}(x, y) \leq \varepsilon$$



$$\blacksquare \quad \phi(x, y) \leq \varepsilon$$

Approximate simulation

ϕ is an (ε, δ) -approximate simulation function from A to B if:

1. $\phi(v_1, v_2) \leq \varepsilon$

2. There exists a function $c: R_1^* \times R_1 \rightarrow R_2^*$ such that for any task T of A and any schedule σ of A if μ_1 is consistent with σ and μ_2 is consistent with $full(c)(\sigma)$ then $\phi(\mu_1, \mu_2) \leq \varepsilon$ implies $\widehat{\phi}(\text{apply}(\mu_1, T), \text{apply}(\mu_2, c(\sigma, T))) \leq \varepsilon$

3. $\phi(\mu_1, \mu_2) \leq \varepsilon$ implies $\mathbf{d}_u(\text{tdist}(\mu_1), \text{tdist}(\mu_2)) \leq \delta$

Lemma 1: ϕ $(\varepsilon, 0)$ -approximate simulation then $R := \{(\mu_1, \mu_2) \mid \phi(\mu_1, \mu_2) \leq \varepsilon\}$ is an exact simulation relation.

Key Lemmas

□ **Lemma 2:** $\hat{\phi}(\mu_1, \mu_2) \leq \varepsilon$ with witness ψ .

To prove that $\hat{\phi}(f_1(\mu_1), f_2(\mu_2)) \leq \varepsilon$ it suffices to show that

$\forall \rho_1, \rho_2 \in \text{supp}(\psi), \hat{\phi}(f_1(\rho_1), f_2(\rho_2)) \leq \varepsilon$.

Key Lemmas

- **Lemma 3:** $\widehat{\phi}(\mu_1, \mu_2) \leq \varepsilon$ implies $\mathbf{d}_u(\text{tdist}(\mu_1), \text{tdist}(\mu_2)) \leq \delta$
- **Lemma 4:** If $\mu = \text{Lt}_{n \rightarrow \infty} \mu_n$ then $\text{tdist}(\mu) = \text{Lt}_{n \rightarrow \infty} \text{tdist}(\mu_n)$.
- **Lemma 5:** If $\mu_{1i} \rightarrow \mu_1$ and $\mu_{2i} \rightarrow \mu_2$ then $\mathbf{d}_u(\mu_{1i}, \mu_{2i}) \rightarrow \mathbf{d}_u(\mu_1, \mu_2)$.

Soundness

- **Theorem:** If there exists an (ϵ, δ) -approximate simulation function from A to B , then A δ -implements B .
- Construct chain of distributions for A applying one task at a time. Construct the corresponding chain for B .
- Induction on the length of the chain
 - Base case from start condition
 - Induction step from Lemma 2
- Use Lemmas 4 & 5 for $n \rightarrow \infty$

Probabilistic Safety

- X be a random variable on $(\mathbf{T}, F_{\mathbf{T}})$.
- If A_I is δ -equivalent to B and for every trace distribution μ_2 of B , $\mu_2[X=x] = p$ then $\mu_1[X=x] \leq p + \delta$.

- $X_u: \mathbf{T} \rightarrow \{0,1\}$ defined as $X_u(\beta) := 1$ if some unsafe action U occurs in β .
- If B is safe with probability p then A is safe with probability at least $p + \delta$.

Task-PIOAs

- An *environment* E for A is a task-PIOA such that $E||A$ is closed
- *External behavior* of A is a function mapping each environment E of A to the set of trace distributions of $E||A$
- A_1 δ -implements B if for every environment E , for every trace distribution μ_1 in $extbeh_A(E)$ there is a trace distribution μ_2 in $extbeh_A(E)$.
- Suppose for every environment E , there exists a (ϵ_E, δ) -approximate simulation function from $A_1||E$ to $B||E$, then A_1 δ -implements B .

Future directions

- Applications: randomized consensus protocols,
- Approximate implementations and simulation relations for task-PIOAs with continuous state spaces.
- Simulations as functions of distributions over states (as opposed to distributions over fragments).
- Explore the possibility of automating simulation proofs by solving optimization problems.
(See my thesis)

Thank you!

Key Lemmas

□ **Lemma 2:** $\hat{\phi}(\mu_1, \mu_2) \leq \varepsilon$ with witness ψ .

$f_i : \text{Disc}(X_i) \rightarrow \text{Disc}(X_i)$ are distributive functions.

If $\forall \rho_1, \rho_2 \in \text{supp}(\psi), \hat{\phi}(f_1(\rho_1), f_2(\rho_2)) \leq \varepsilon$ then $\hat{\phi}(f_1(\mu_1), f_2(\mu_2)) \leq \varepsilon$.

For each $\rho_1, \rho_2 \in \text{supp}(\psi)$, let ψ_{ρ_1, ρ_2} be the witnessing joint for $\hat{\phi}(f_1(\rho_1), f_2(\rho_2)) \leq \varepsilon$.

Define $\psi' := \sum_{\rho_1, \rho_2 \in \text{supp}(\psi)} \psi(\rho_1, \rho_2) \psi_{\rho_1, \rho_2}$

Show : $f_i(\mu_i) = \sum_{\eta_1, \eta_2} \psi'(\eta_1, \eta_2) \eta_i$ and $\eta_1, \eta_2 \in \text{supp}(\psi')$ implies $\phi(\eta_1, \eta_2) \leq \varepsilon$