

# Domain Theory and Task-structured Probabilistic Input/Output Automata

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Joint Work With  
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- Summary and Future work

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$$\mathcal{P} \leq_s \mathcal{F} \text{ iff } (\forall \mathcal{A}dv)(\exists \mathcal{I})(\forall \mathcal{E}) \quad \mathcal{P} \parallel \mathcal{A}dv \parallel \mathcal{E} \simeq \mathcal{F} \parallel \mathcal{I} \parallel \mathcal{E}$$

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- $\leq_s$ : Trace distributions of observable events from  $(\mathcal{P} \parallel \mathcal{A}dv \parallel \mathcal{E})$  all appear in the trace distributions of  $(\mathcal{F} \parallel \mathcal{I} \parallel \mathcal{E})$ .

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- *Focus of talk:* **Task-structured Probabilistic Input/Output Automata**

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- $S$  - countable set of states,  $s_0$  - start state
- Act ::=  $I \cup O \cup H$  - countable set of actions
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**Transition determinism:**  $(s, a, \mu), (s, a, \nu) \in D \Rightarrow \mu = \nu$

**Input enabling:**  $(\forall s \in S)(\forall a \in I)(\exists \mu) (s, a, \mu) \in D$

## How to compose PIOAs

$\mathcal{A}_i = (S_i, s_{\mathcal{A}_i}, I_i, O_i, H_i, D_i)$ ,  $i = 1, 2$  are *compatible* if

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$\mathcal{A}_1 \parallel \mathcal{A}_2 = (S_{\mathcal{A}_1, \mathcal{A}_2}, s_{\mathcal{A}_1, \mathcal{A}_2}, I_{\mathcal{A}_1, \mathcal{A}_2}, O_{\mathcal{A}_1, \mathcal{A}_2}, H_{\mathcal{A}_1, \mathcal{A}_2}, D_{\mathcal{A}_1, \mathcal{A}_2})$  where:

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$\mathcal{A}$  is *closed* if  $I_{\mathcal{A}} = \emptyset$

## Tasks

■ **Tasks:** Equivalence relation  $\mathcal{R} \subseteq (O \cup H) \times (O \cup H)$ .

*Task:* Any equivalence class of  $\mathcal{R}$ .

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- **Application:** Via *task schedules* -  $\rho = T_1 T_2 \dots$

## Example

- *Player*: Flips coin and announces result
- *Opponent*: Announces *Heads* or *Tails*
- If they match, *Player* wins, otherwise *Opponent* wins.

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States:  $\perp$ , flipped, announced, initially  $\perp$   
Actions: Flip; Output: *Heads*, *Tails*  
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  - *Task Schedules*:  
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## PIOA Semantics

$$\mathcal{A} = (S, s_0, I, O, H, D)$$

$$\text{Frags}^*(\mathcal{A}) = \bigcup_n \{\alpha \in (S \times \text{Act})^n \times S \mid \alpha \text{ finite execution fragment}\}$$

Execution fragment:

$$\alpha = s_1 a_1 s_2 a_2 \cdots \text{ with } s_{i+1} \in \text{supp}(\mu_i) \text{ & } (s_i, a_i, \mu_i) \in D$$

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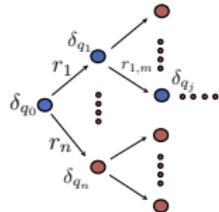
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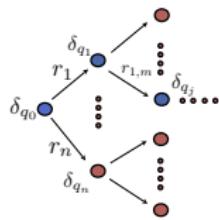
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Apply task schedule  $\rho = T_1 T_2 \cdots$

$$\delta_{q_0} \xrightarrow{T_1} \sum_q \mu_{q_0, a_{T_1}}(q) \delta_{q_0 a_{T_1} q} \xrightarrow{T_2} \cdots$$

## Defining Apply

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For  $\rho = T_1 \cdots T_n$

$$\text{Apply}(\mu, \rho) = \text{Apply}(\text{Apply}(\mu, T_1), T_2 \cdots T_n))$$

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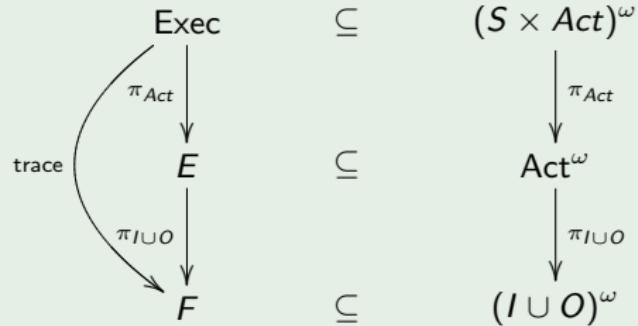
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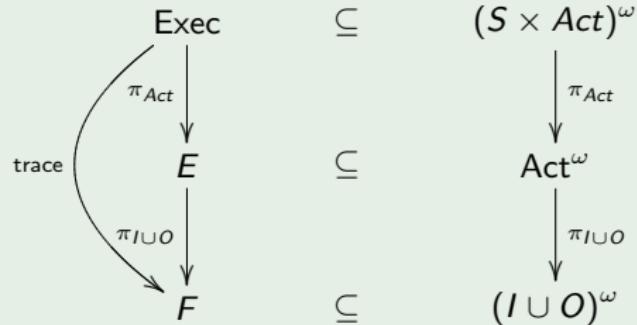
Hence  $\text{Apply}(\mu, \rho) = \sup_n \text{Apply}(\mu, T_1 \dots T_n)$  is well-defined.

## Semantics of Observable Events



where  $E = \{\alpha|_{Act^\omega} \mid \alpha \in \text{Exec}\}$  and  $F = \{\alpha|_{(I \cup O)^\omega} \mid \alpha \in \text{Exec}\}$

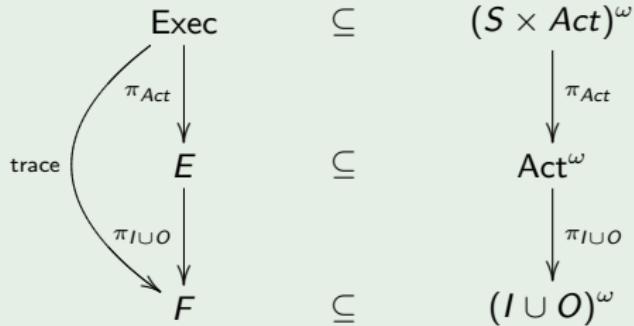
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Given  $\mathcal{A}$  and  $\rho = T_1 \dots$ , we'd like to have a *scheduler* – determined in advance – that would represent applying  $\rho$  to any  $\mu \in \text{Prob}(\text{Frags}^*(\mathcal{A}))$ .

## Task Schedulers

A *task scheduler* is a map

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## Measures from Schedulers - Original Definition

If  $\sigma: \text{Frags}^*(\mathcal{A}) \rightarrow \mathbb{V}(\text{Act})$  is a scheduler and  $\alpha \in \text{Frags}^*(\mathcal{A})$  then define  
 $\epsilon_\sigma: \text{Prob}(\text{Frags}^*(\mathcal{A})) \rightarrow \text{Prob}(\text{Frags}^*(\mathcal{A}))$  by

$$\epsilon_{\sigma, \alpha}(\uparrow \alpha') = \begin{cases} 0 & \text{if } \alpha' \not\leq \alpha \not\leq \alpha' \\ 1 & \text{if } \alpha' \leq \alpha \\ \epsilon_{\sigma, \alpha}(\uparrow \alpha'') \sigma(\alpha'')(a) \mu_{\alpha'', a}(s) & \text{if } \alpha \leq \alpha' = \alpha'' as, \end{cases}$$

where  $\mu_{\alpha'', a}(s)$  is the probability of landing in state  $s$  starting from  $\text{ls}(\alpha)$  after executing action  $a$ .

## Measures from Schedulers - Our Definition

Let  $\mathcal{A}$  be a task PIOA and let  $\sigma: \text{Frags}^*(\mathcal{A}) \rightarrow \mathbb{V}(\text{Act})$  be a task scheduler. If  $\alpha \in \text{Frags}^*(\mathcal{A})$ , we define

$$\epsilon'_{\sigma, \alpha} = (1 - ||\sigma(\alpha)||) \delta_\alpha + \sum_{a \in \text{Act}} \sigma(\alpha)(a) \left( \sum_s \mu_{\alpha, a}(s) \epsilon'_{\sigma, \alpha s} \right)$$

Then,  $\epsilon'_{\sigma, \alpha} = \epsilon_{\sigma, \alpha}$  for all schedulers  $\sigma$  and  $\alpha \in \text{Frags}^*(\mathcal{A})$ .

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Neat fact:

$$\epsilon'_{\sigma, \alpha, 0} = \delta_\alpha$$

$$\epsilon'_{\sigma, \alpha, n+1} = (1 - ||\sigma(\alpha)||) \delta_\alpha + \sum_{a \in \text{Act}} \sigma(\alpha)(a) \left( \sum_s \mu_{\alpha, a}(s) \epsilon'_{\sigma, \alpha a s, n} \right)$$

implies  $\epsilon'_{\sigma, \alpha} = \sup_n \epsilon'_{\sigma, \alpha, n}$

## Schedulers vs. Task Schedules

### Theorem

Let  $\mu \in \text{Prob}(\text{Frags}^*(\mathcal{A}))$  have support consisting of incomparable fragments, and let  $\rho$  be a task schedule. Then there is a scheduler  $\sigma_\rho: \text{Frags}^*(\mathcal{A}) \rightarrow \mathbb{V}(\text{Act})$  such that  $\text{Apply}(\mu, \rho) = \epsilon_{\sigma_\rho, \mu}$ .

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In fact, for  $\mu = \sum_\alpha \mu(\alpha) \delta_\alpha$  and  $\rho = \rho' T$ , the scheduler  $\sigma_\rho$  is deterministic:

$$\sigma_\rho(\alpha) = \begin{cases} \delta_{a_{ls(\alpha), T}} & \text{if } \alpha \in A_T \cap \text{supp } \mu, \\ \sigma_{\rho'}(\alpha) & \text{if } \sigma_{\rho'}(\alpha) \neq 0, \\ 0 & \text{otherwise.} \end{cases}$$

$\rho = T_1 \dots$  infinite implies  $\sigma_\rho = \cup_n \sigma_{T_1 \dots T_n}$ .

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**Corollary** Let  $\mathcal{A}$  be a Task PIOA.

- For each task schedule  $\rho$ , there is a deterministic task scheduler  $\sigma_\rho$  satisfying

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- $\text{tdist}(\mathcal{A}) \subseteq \{\text{trace}(\epsilon_{\sigma, \delta_{s_0}}) \mid \sigma \text{ deterministic scheduler}\}$ .

## Simulations

Recall

$$(\mathcal{P} \parallel \mathcal{A}dv \parallel \mathcal{E}) \simeq (\mathcal{F} \parallel \mathcal{I} \parallel \mathcal{E})$$

For us,

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Let  $(\mathcal{A}_i, \mathcal{R}_i)$  be two task PIOAs and let  $f: \mathcal{R}_1^* \times \mathcal{R}_1 \rightarrow \mathcal{R}_2^*$  be a function. We define  $\text{full}(f): \mathcal{R}_1^* \rightarrow \mathcal{R}_2^*$  by

$$\begin{aligned}\text{full}(f)(\langle \rangle) &= \langle \rangle \\ \text{full}(f)(\rho T) &= \text{full}(f)(\rho) \wedge f(\rho, T)\end{aligned}$$

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$R \subseteq \text{Prob}(\text{Exec}(\mathcal{A}_1)) \times \text{Prob}(\text{Exec}(\mathcal{A}_2))$  is a *simulation* if

- $(\mu_1, \mu_2) \in R \Rightarrow \text{tdist}(\mu_1) \subseteq \text{tdist}(\mu_2)$
- $(\delta_{s_0,1}, \delta_{s_0,2}) \in R$
- $(\exists f: \mathcal{R}_1^* \times \mathcal{R}_1 \rightarrow \mathcal{R}_2^*)(\forall \rho \in \mathcal{R}_1^*)(\forall T \in \mathcal{R}_1)$

$$\begin{aligned}
 & (\mu_1, \mu_2) \in R \wedge \text{supp}(\mu_1) \subseteq \rho \wedge \text{supp}(\mu_2) \subseteq \text{full}(f)(\rho) \\
 & \Rightarrow (\text{Apply}(\mu_1, T), \text{Apply}(\mu_2, \text{full}(f)(\rho), T)) \in \mathcal{E}(R)
 \end{aligned}$$

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**Theorem** (Canetti, et al)

Let  $\mathcal{A}_1$  and  $\mathcal{A}_2$  be comparable task-PIOAs that are closed and action-deterministic. If there exists a simulation relation from  $\mathcal{A}_1$  to  $\mathcal{A}_2$ , then  $\text{tdist}(\mathcal{A}_1) \subseteq \text{tdist}(\mathcal{A}_2)$ .

## Expansions and Monads

$(X, \Sigma_X), (Y, \Sigma_Y)$  measure spaces,  $R \subseteq X \times Y$ . The *lift* of  $R$  is  $\widehat{R} \subseteq \mathbb{V}X \times \mathbb{V}Y$  defined by

$$(\sum_x r_x \delta_x, \sum_y s_y \delta_y) \in \widehat{R} \Leftrightarrow \exists t: X \times Y \rightarrow [0, 1] \text{ with}$$

- $r_x = \sum_y t(x, y)$  ( $\forall x$ )
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$\mathbb{V}: \text{Meas} \rightarrow \text{Meas}$  is a *monad*, and  $R \mapsto \mathcal{E}(R)$  utilizes the *lifting* and *multiplication* of the monad.

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## Future Work

- More about the use of the monad  $\mathbb{V}$
- Application to Dining Cryptographers
- Application to other protocols, combining the UC of oblivious transfer due to Canetti, et al.