# Decidability of Parameterized Probabilistic Information Flow 

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- And somebody observes a part of its behavior
- We fix some property of the system
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## Can the observer recover that property?

Our result: there is an algorithm answering the above question given any system and property

## Outline

(1) Information Flow: Definitions

- System
- Observation
- System Properties
- Definition of Information Flow


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(1) Information Flow: Definitions

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(2) Decidability Results


## Part I

## What is a system?

What is a partial observation of its behavior?

## What is a property of the system?

When does a system have information flow?

## System is a Probability Distribution

- A system is a probability distribution over traces
- A trace is a finite or infinite sequence of alphabet characters
- Today: $\Sigma=L \cup H$ (low-level and high-level events)
- The distribution is described by Finite Markov Automaton


## Finite Markov Automaton

- Finite number of states
- Edges are labelled by alphabet characters
- Every edge has a probability
- For every edge the sum of probabilities over all outgoing edges is equal to 1



## Observation model

For any trace $\alpha$ observation is a projection to low-level events $\left.\alpha\right|_{L}$

Projection is just deleting all characters from $H$ from the sequence

## Defining a System Property

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Today we restrict ourselves to properties recognized by Muller automaton

## Muller Automaton

- Finite number of states
- Initial state, family $\mathcal{F}$ of "accepting" sets of states
- Every edge is labelled by alphabet character
- The automaton is complete and deterministic: for every pair $(v, a)$ there exist a unique outgoing edge from the vertex $v$ with that label a
- Muller automaton accepts trace if during "reading" it the set of states visited infinitely many times belongs to $\mathcal{F}$



## Property-Specific Information Flow

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System $S$ has no information flow for property $P$ if

$$
\forall u \quad \operatorname{Pr}_{S}(P \mid u)=\operatorname{Pr}_{r_{S}}(P)
$$

## General Information Flow

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A system is without sequential information flow iff it has no flow for every sequential property $P$

## Information Flow: Example



The Markov chain above has no sequential information flow

## Information Flow: Example



Now the Markov chain above has sequential information flow

## Part II

## Given system/property can we determine the existence of information flow?

## Deciding Property-Specific Information Flow

## Theorem

There is an algorithm deciding property-specific information flow for every pair of
system/property (i.e. for pair of Muller automaton and Markov chain)

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Reduction to linear algebra:
(1) Compute a composition of Markov automaton and Buc̈hi automaton
(2) Simplify it by the rule " $H^{*} I \rightarrow I$ "

## Algorithm Inside (1/2)

We reduce property-specific information flow to the following mathematical problem:

Input: vectors $a, c$, matrices $M_{1}, \ldots, M_{n}$
Question: does there exist a finite sequence of indices such that $a M_{i_{1}} \ldots M_{i_{k}} c \neq 0$ ?

## Algorithm Inside (2/2)

For every $k$ we will compute basis for linear hull of $V_{k}=\left\{a M_{i_{1}} \ldots M_{i_{k}}\right\} \cup V_{k-1}$

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(3) Stopping condition: $\operatorname{dim}\left(V_{k+1}\right)=\operatorname{dim}\left(V_{k}\right)$
(9) Check whether $V_{k} \perp c$

## Deciding General Information Flow

Theorem
There is an algorithm deciding general information flow for every system described by a Markov chain

Theorem
There is an algorithm deciding sequential information flow for every system described by a Markov chain

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## Future work

- More general models for systems and properties
- Quantitative measure for information flow


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Thanks for your attention! Questions?

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## Some related work：

曷
D．Beauquier，M．Duflot，Y．Lifshits
Decidability of Parameterized Probabilistic Information Flow．CSR＇07．
http：／／yury．name／papers／beauquier2007decidability．pdf
茙
D．Beauquier，M．Duflot，M．Minea
A Probabilistic Property－Specific Approach to Information Flow．MMM－ACNS＇05． http：／／www．univ－paris12．fr／lacl／Rapports／publications／TR－2005－02．pdf

輷
A．Slissenko
Complexity problems in the analysis of information systems security．MMM＝ACNS＇03． http：／／www．springerlink．com／index／WKDENHGBAFE28KNC．pdf

