Learning Broadcast Protocols (with LeoParDS)

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Automata Learning







Parametrized protocols

Parametrized protocols

Given a protocol $P, n \in \mathbb{N}$.

A parallel running of protocol P for n processes: $P^n = P \parallel P \parallel \cdots \parallel P$

The "language" of
$$P^n$$
: $\mathcal{L}(P^n) = P \parallel P \parallel \cdots \parallel P$
n
The "language" of a protocol P : $\mathcal{L}(P) = \bigcup_{n \in \mathbb{N}} \mathcal{L}(P^n)$









[1999] On the Verification of Broadcast Protocols









state vector













state vector





state vector





state vector





Note that $\mathcal{L}(P^1) \subseteq \mathcal{L}(P^2) \subseteq \mathcal{L}(P^3) \subseteq \cdots$

Are these inclusions strict, or does there exist an *n* s.t. adding more processes does not change the language?

Cutoff

If $\exists n \in \mathbb{N}$ s.t. $\forall m > n \mathcal{L}(P^n) = \mathcal{L}(P^m)$

If such an *n* exists, then the system has a **cutoff**, *n*. Otherwise, we say there is no cutoff.

Cutoff

$\exists n \in \mathbb{N} \text{ s.t. } \forall m > n \mathcal{L}(P^n) = \mathcal{L}(P^m)$

Fine BPs

A BP that:

1. Has no hidden states

2. A cutoff exists

Learning paradigms





Protocols Inference



Consistent sample

Inference

We provide an inference algorithm for BPs, given a sample of words that are consistent with a BP, infers a correct BP.



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Consistency

Let C be the class of fine BPs,

Given sample S and $k \in \mathbb{N}$, determine whether there

exists a BP $B \in C$ consistent with S with at most k states.

Consistency

We show that consistency is NP-hard for the class of fine BPs.



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Consistency



Figure 3: Reduction of DFA-consistency to BP-consistency.



Polynomial data

Is there an inference-algorithm \mathcal{A} s.t. for all BP $B \in C$, one can associate a polynomial-sized sample \mathcal{S}_B so that \mathcal{A} correctly infers $\mathcal{L}(B)$ from any sample subsuming \mathcal{S}_B .

Recall: We mark the class of fine BPs as C.

Polynomial data

We show that there exist fine BPs for which there is no characteristic set of polynomial size.

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Polynomial data





Polynomial Predictability

Can a learner correctly classify an unknown word with high probability after asking polynomially many membership queries.

Polynomial Predictability

We show that under plausible cryptography assumptions, fine BPs (thus BPs in general) are not polynomially predictable.

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Polynomial Predictability



Figure 4: A BP simulating intersection of k DFAs.





