

# Learning and Guessing Winning Policies in LTL Synthesis via Semantics

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joint work with

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IARCS Verification Seminar

October 15, 2024



Alonzo Church, 1957

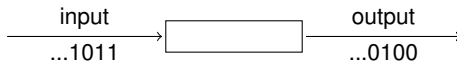
“Given a requirement which a circuit is to satisfy, we may suppose the requirement expressed in some suitable logistic system which is an extension of restricted recursive arithmetic. The **synthesis problem** is then to find recursion equivalences representing a circuit that satisfies the given requirement (or alternatively, to determine that there is no such circuit).”



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Given a requirement on a bit stream transformation



fill the box by a machine with output, satisfying the requirement (or state that the requirement is not satisfiable).



Given a **specification**  $\varphi$  and atomic propositions partitioned into  $I$ (nput) and  $O$ (utput), synthesize a (finite circuit)  $f$ :

$$\forall \text{ input stream } I : \quad I \parallel O \models \varphi$$

where



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$$\forall \text{ input stream } I : \quad I \parallel O \models \varphi$$

where

$$I = i_1 i_2 i_3 \dots \quad \text{with } i_n \subseteq I$$

$$O = o_1 o_2 o_3 \dots = f(i_1) f(i_1 i_2) f(i_1 i_2 i_3) \dots \quad \text{with } o_n \subseteq O$$

$$I \parallel O = i_1 \cup o_1 \quad i_2 \cup o_2 \quad i_3 \cup o_3 \dots$$

$$\varphi ::= a \mid \neg a \mid \varphi \wedge \varphi \mid \varphi \vee \varphi \mid \mathbf{F}\varphi \mid \mathbf{G}\varphi \mid \varphi \mathbf{U}\varphi$$

Example:  $I = \{a\}, O = \{b\}, \varphi = \mathbf{G}(a \text{ xor } b)$

	$\ell_1$	$\ell_2$	$\dots$
$I$	$\{a\}$	$\emptyset$	
$O$	$\emptyset$	$\{b\}$	
$I \parallel O$	$\{a\}$	$\{b\}$	



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Example:  $I = \{r\}, O = \{g\}, \varphi = \mathbf{G}(r \Rightarrow \mathbf{F}g)$

	$\ell_1$	$\ell_2$	$\dots$
$I$	$\{r\}$	$\{r\}$	
$O$	$\emptyset$	$\{g\}$	
$I \parallel O$	$\{r\}$	$\{r, g\}$	



Given a **specification**  $\varphi$  and atomic propositions partitioned into **I**(nput) and **O**(utput), synthesize a (finite circuit)  $f$ :

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Example:

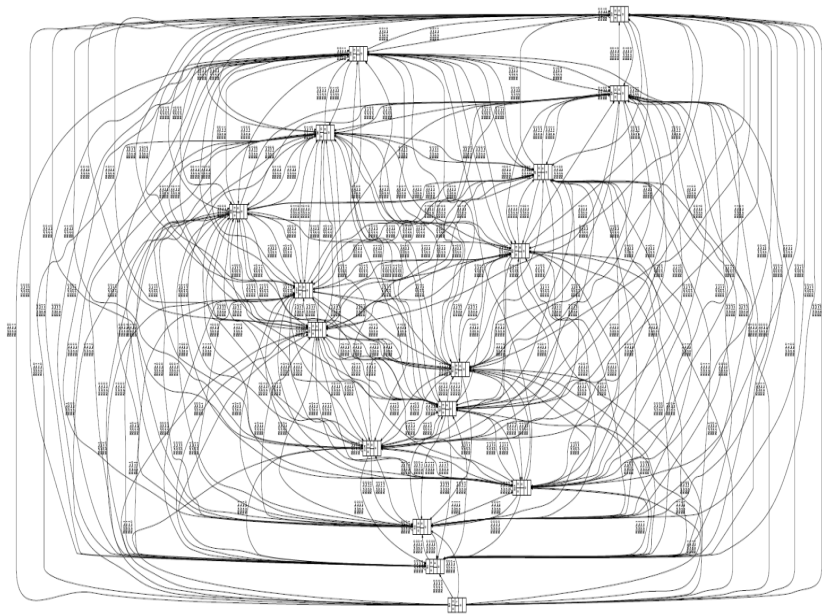
	$\ell_1$	$\ell_2$	$\dots$
$I$			
$O$			
$I \parallel O$			







MEM-OUT



## Formal methods: precise synthesis

- + optimality of results
- scalability issues
- + precise outputs (guaranteed safety)
- precise inputs (known models)
- + specification-oriented
- ad hoc algorithms

## Learning: insightful guesses

- weaker guarantees
- + scalable
- imprecise outputs
- + imprecise inputs
- simple objectives only
- + problem-independent



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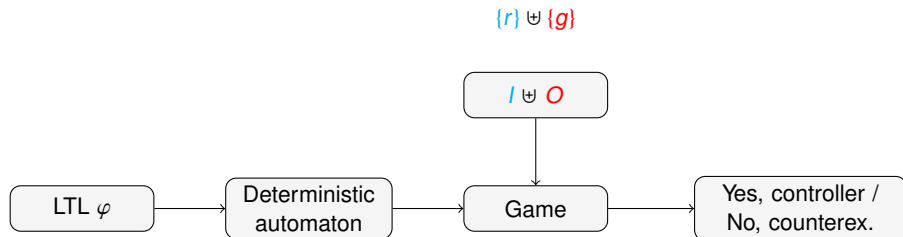
*precise computation*



*focus on important stuff*

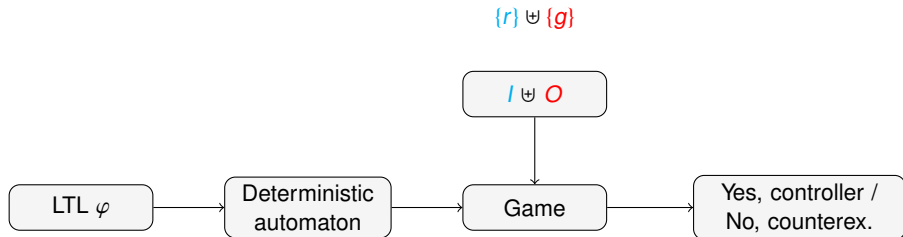
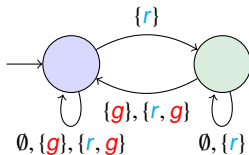


LTL synthesis:

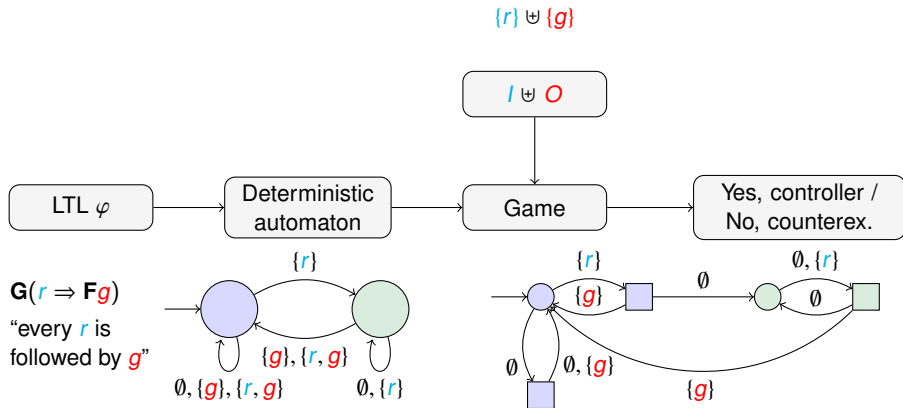
 $\mathbf{G}(r \Rightarrow \mathbf{F}g)$ 

"every  $r$  is  
followed by  $g$ "

LTL synthesis:

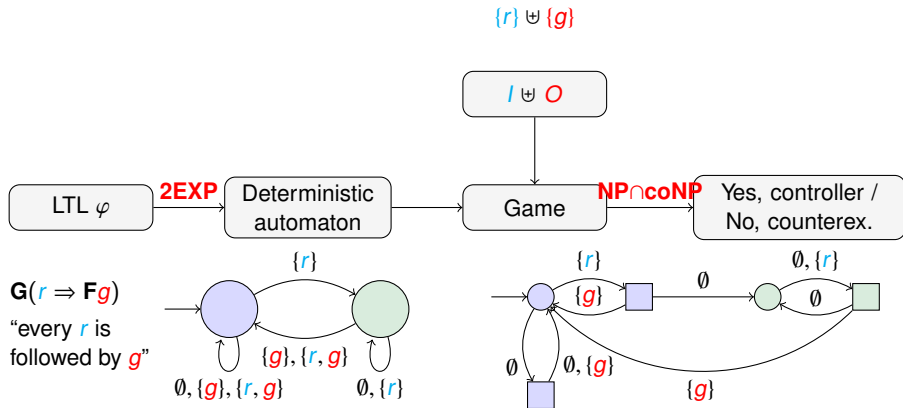
 $\mathbf{G}(r \Rightarrow \mathbf{F}g)$ "every  $r$  is followed by  $g$ "

LTL synthesis:





LTL synthesis:



### *Fragments:*

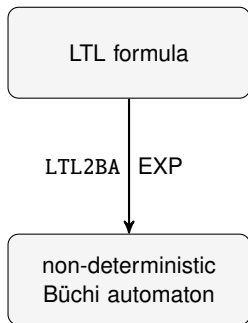
- ▶ R. Alur, S. La Torre: Deterministic generators and games for LTL fragments. ACM ToCL 2004
- ▶ N. Piterman, A. Pnueli, Y. Sa'ar: Synthesis of Reactive(1) Designs. VMCAI 2006

### *Safraless:*

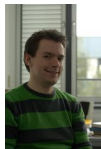
- ▶ O. Kupferman, N. Piterman, M. Y. Vardi: Safraless Compositional Synthesis. CAV 2006
- ▶ B. Jobstmann, R. Bloem: Optimizations for LTL Synthesis. FMCAD 2006

### *Bounded synthesis:*

- ▶ S. Schewe, B. Finkbeiner: Bounded Synthesis. ATVA 2007
- ▶ R. Ehlers: Unbeast: Symbolic Bounded Synthesis. TACAS 2011
- ▶ A. Bohy, V. Bruyère, E. Filiot, N. Jin, J.-F. Raskin: Acacia+, a Tool for LTL Synthesis. CAV 2012
- ▶ P. Faymonville, B. Finkbeiner, L. Tentrup: BoSy: An Experimentation Framework for Bounded Synthesis. CAV 2017



$\bigwedge_{i \in \{1, \dots, n\}} \mathbf{GF} a_i \Rightarrow \mathbf{GF} b_i$	NBA
	LTL2BA
$n = 1$	4
$n = 2$	14
$n = 3$	40



A. Gaiser

LTL formula

LTL2BA EXP

non-deterministic  
Büchi automaton

ltl2dstar EXP

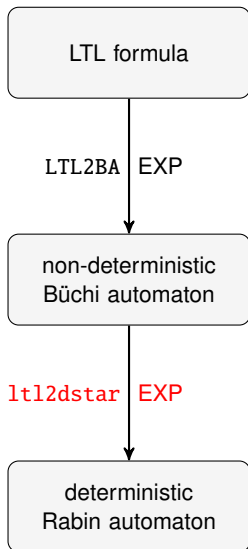
deterministic  
Rabin automaton

$\bigwedge_{i \in \{1, \dots, n\}} \mathbf{GF}a_i \Rightarrow \mathbf{GF}b_i$	NBA
	LTL2BA
$n = 1$	4
$n = 2$	14
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S. Safra

determinization [Safra'88, ...]



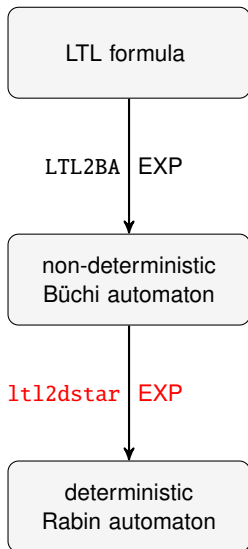
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$n = 1$	4	4
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- ▶ inefficient



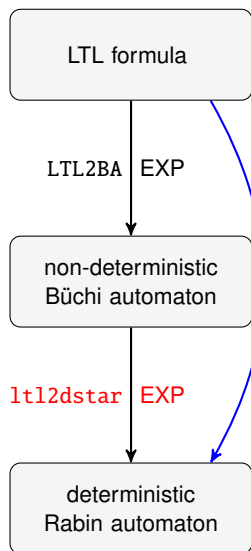
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O. Kupferman

determinization [Safra'88, ...]

- ▶ inefficient
- ▶ "messy" [Kupferman'12]

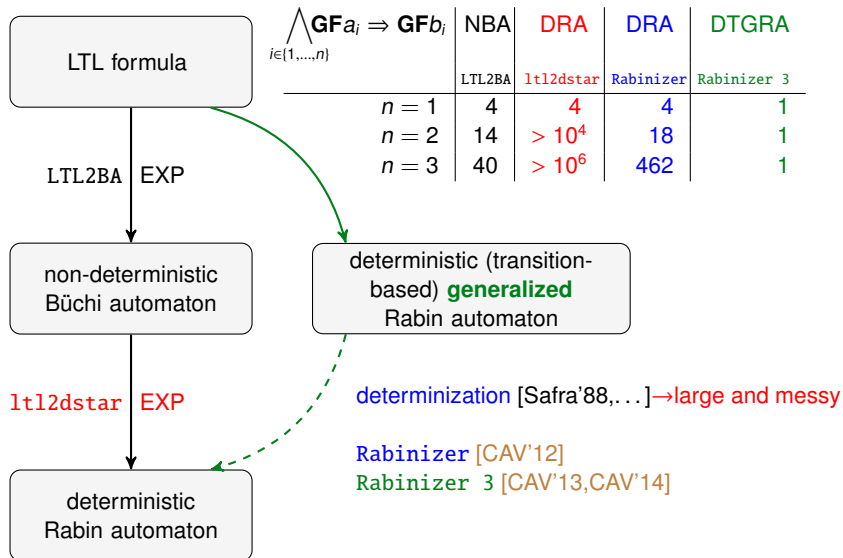


$\bigwedge_{i \in \{1, \dots, n\}} \text{GF}a_i \Rightarrow \text{GF}b_i$	NBA	DRA	DRA
	LTL2BA	ltl2dstar	Rabinizer
$n = 1$	4	4	4
$n = 2$	14	$> 10^4$	18
$n = 3$	40	$> 10^6$	462

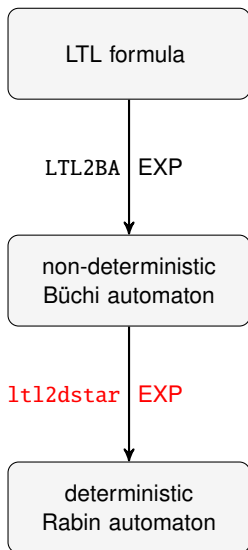
determinization [Safra'88, ...]  $\rightarrow$  large and messy

Rabinizer [CAV'12]

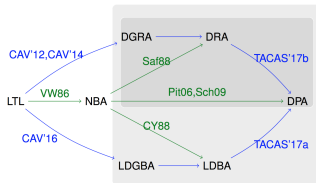
- ▶ smaller automata
- ▶ logical structure of states







$\bigwedge_{i \in \{1, \dots, n\}} \mathbf{GF}a_i \Rightarrow \mathbf{GF}b_i$	NBA	DRA	DRA	DTGRA
	LTL2BA	ltl2dstar	Rabinizer	Rabinizer 3
$n = 1$	4	4	4	1
$n = 2$	14	$> 10^4$	18	1
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determinization [Safr88, ...]  $\rightarrow$  large and messy

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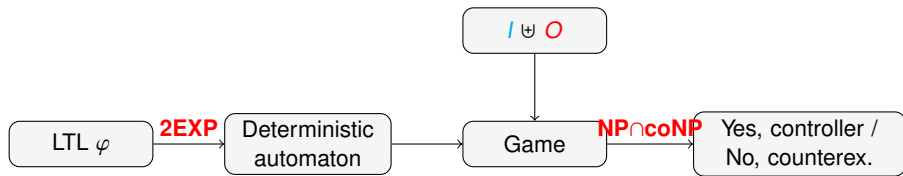
Rabinizer 3 [CAV'13, CAV'14]

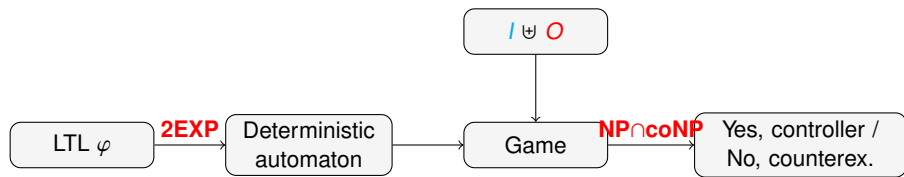
... [CAV'16, TACAS'17a, TACAS'17b, CAV'18]

... [LICS'18, Journal of ACM'20]

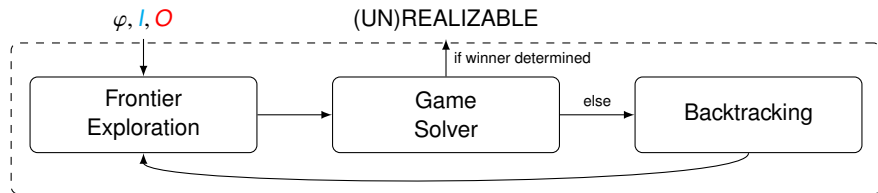
$\rightarrow$  **small and with logical structure**



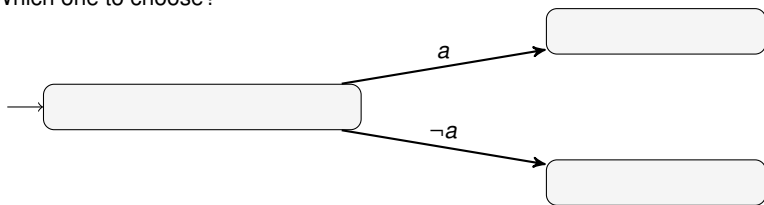




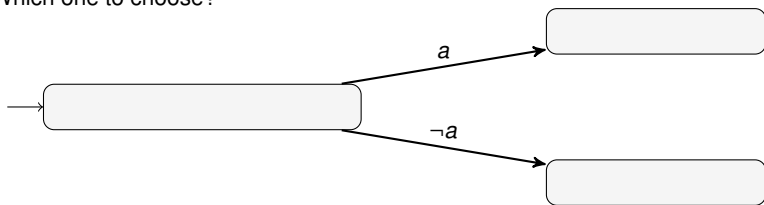
On-the-fly approaches:



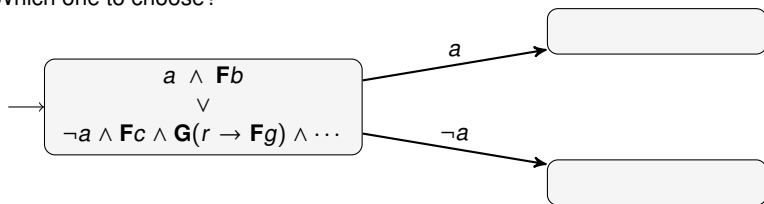
Which one to choose?



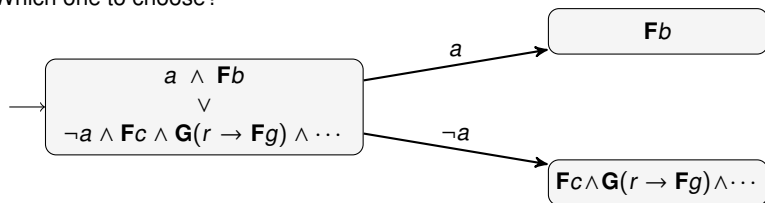
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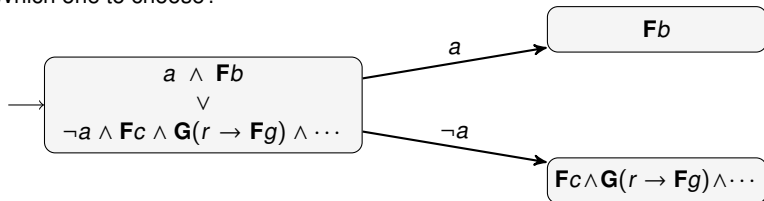


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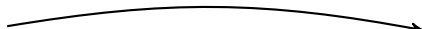




Which one to choose?



train



exploration hint



## Semantic labelling

- ▶ similar to NBA
  - ▶ obscured by Safra
  - ▶ recovered by direct approaches
- ⇒ learning what to do

Which formula is easier to satisfy?

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Naive approximation:

1. see the formula as Boolean, e.g.  $a \wedge \mathbf{GF}a$  as  $A \wedge B$
2. compute the probability of satisfaction by a random assignment

$$\frac{\# \text{ satisfying assignments}}{\# \text{ all assignments}}$$

(easy for BDDs)

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- ▶ modal structure
- ▶ subgoals



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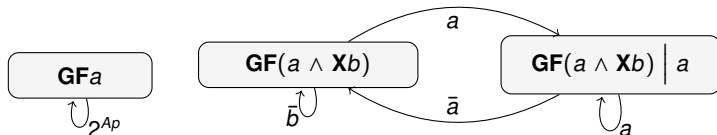
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Immediately solved games by initialization<sup>1</sup>

	Random	Trueness
(Co-)Safety	32%	65%
Near (Co-)Safety	11%	67%
Parity	10%	56%

Strix [MSL18]<sup>2</sup>

- ▶ winning LTL tracks in SyntComp 2018–2023
- ▶ by new translations + on-the-fly strategy iteration
- ▶ trueness-aided exploration + classic algorithm for parity games

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<sup>1</sup>J. K., A. Manta, T. Meggendorfer: Semantic Labelling and Learning for Parity Game Solving in LTL Synthesis. ATVA 2019

<sup>2</sup>P. Meyer, S. Sickert, M. Luttenberger: Strix: Explicit Reactive Synthesis Strikes Back! CAV 2018

Involve **machine learning**<sup>34</sup>

1. Consider further features
2. Learn on winning/losing transitions from previously solved games

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<sup>3</sup>J. K., A. Manta, T. Meggendorfer: Semantic Labelling and Learning for Parity Game Solving in LTL Synthesis. ATVA 2019

<sup>4</sup>J. K., T. Meggendorfer, M. Prokop, S. Rieder: Guessing Winning Policies in LTL Synthesis by Semantic Learning. CAV 2023



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Some aspects:

- ▶ Models: SVM, DT, NN, GNN,...

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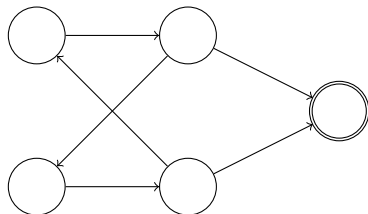
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Involve **machine learning**<sup>34</sup>

1. Consider further features
2. Learn on winning/losing transitions from previously solved games

Some aspects:

- ▶ Models: SVM, DT, NN, GNN,...
- ▶ Ground truth: Beyond safety no maximally permissive strategies!



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<sup>3</sup>J. K., A. Manta, T. Meggendorfer: Semantic Labelling and Learning for Parity Game Solving in LTL Synthesis. ATVA 2019

<sup>4</sup>J. K., T. Meggendorfer, M. Prokop, S. Rieder: Guessing Winning Policies in LTL Synthesis by Semantic Learning. CAV 2023

Games where the a-priori learnt strategy is already winning [ATVA'19]:

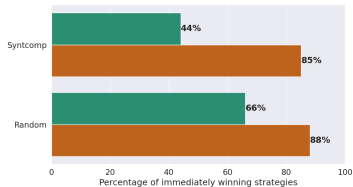
category	random	trueness	svm
small	20.00%	60.00%	<b>93.33%</b>
small safety	16.67%	<b>100.00%</b>	<b>100.00%</b>
small p-safety	16.67%	75.00%	<b>100.00%</b>
small co-safety	41.67%	<b>100.00%</b>	<b>100.00%</b>
small p-co-safety	33.33%	83.33%	<b>100.00%</b>
large	6.67%	53.33%	<b>93.33%</b>
large safety	0.00%	<b>100.00%</b>	<b>100.00%</b>
large co-safety	23.08%	<b>100.00%</b>	<b>100.00%</b>
lily	11.11%	44.44%	<b>55.56%</b>
ltl2dba	0.00%	0.00%	<b>62.50%</b>
ltl2dpa	0.00%	0.00%	<b>54.55%</b>
<b>total</b>	15.91%	68.18%	<b>89.39%</b>

# SVM in action

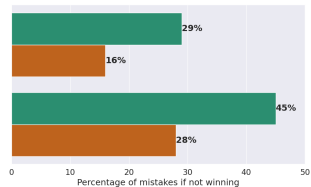
**Trueness** – How easy is it to satisfy the formula (in one step)?

Train **SVM** with SYNTCOMP data and manual features

Predicted strategy winning?



How many changes by SI, if not winning?

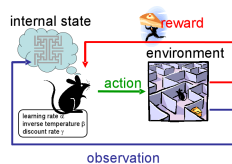


Translation LTL→automata:

- ▶ 2EXP (via Safra's determinization)
- ▶ new, direct approaches
  - ▶ practically more efficient
  - ▶ preserving semantic information

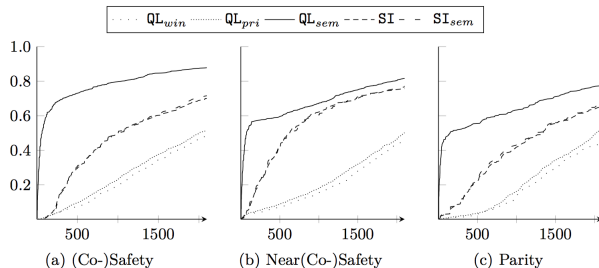
Solving parity games

- ▶  $NP \cap coNP$
- ▶ heuristics
  - ▶ initialization
  - ▶ on the fly

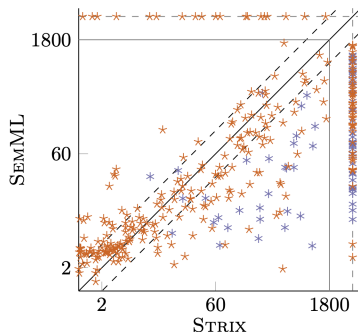


Solved games by number of evaluation steps

- ▶  $QL_{win}$ : Q-learning with only win/loss as reward signal
- ▶  $QL_{pri}$ : Q-learning with priority-based rewards
- ▶  $QL_{sem}$ : Q-learning with semantic rewards



SYNTCOMP		SEMML	
		solved	unsolved
STRIX	solved	951	27
	unsolved	49	84



Speed-ups:

SYNTCOMP	0	5	30	300	Synthetic	0	5	30	300
ratio	0.09	1.37	2.08	3.58	ratio	8.56	8.56	9.48	13.44
count	951	148	89	30	count	30	30	28	14

- ▶ semantic information  $\implies$  apply learning
- ▶ SEMML winning SyntComp
- ▶ not easy to learn well (complex discrete structure)
- ▶ lots of future work on (i) learning and (ii) theory



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Thank you!



Proximity and relationship between formulae **beyond logical implication**  
⇒ **embedding of formulae into spaces** where learning can be done well<sup>56</sup>

Further applications:

**“Learning” model checking** Determine (a probability of) satisfaction of a formula based on satisfaction of other (unrelated) formulae.

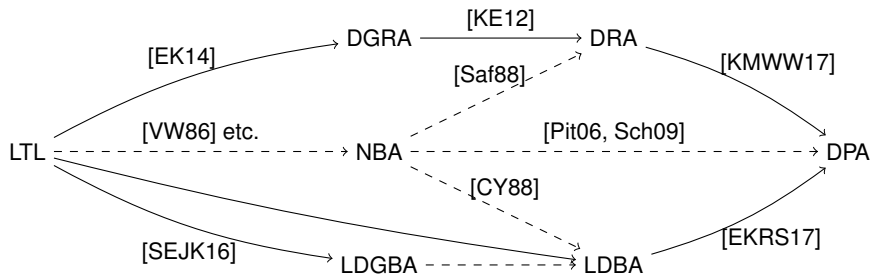
**Translating, sanitizing and simplifying specifications** Find the closest simple formula to the inadequate translation from English to logic.

**Requirement mining** Lifting the search problem from the discrete combinatorial space of syntactic structures of formulae to a continuous space in which distances preserve semantic similarity.

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<sup>5</sup>L. Bortolussi, G. M. Gallo, J.K., L. Nenzi: Learning Model Checking and the Kernel Trick for Signal Temporal Logic on Stochastic Processes. TACAS 2022

<sup>6</sup>G. Saveri, L. Nenzi, L. Bortolussi, J.K.: stl2vec: Semantic and Interpretable Vector Representation of Temporal Logic. ECAI 2024



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Costas Courcoubetis and Mihalis Yannakakis.

Verifying temporal properties of finite-state probabilistic programs.

In *FOCS*, pages 338–345, 1988.



Javier Esparza and Jan Křetínský.

From LTL to deterministic automata: A Safrales compositional approach.

In *CAV*, pages 192–208, 2014.



Javier Esparza, Jan Křetínský, Jean-Francois Raskin, and Salomon Sickert.

From LTL and limit-deterministic Büchi automata to deterministic parity automata.

In *TACAS*, pages 426–442, 2017.



Jan Křetínský and Javier Esparza.

Deterministic automata for the (F,G)-fragment of LTL.

In *CAV*, volume 7358 of *LNCS*, pages 7–22, 2012.



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Games where the initial strategy is winning

category	games	random		trueness		svm	
small	15	3	20.00%	9	60.00%	14	<b>93.33%</b>
small safety	12	2	16.67%	12	<b>100.00%</b>	12	<b>100.00%</b>
small p-safety	12	2	16.67%	9	75.00%	12	<b>100.00%</b>
small co-safety	12	5	41.67%	12	<b>100.00%</b>	12	<b>100.00%</b>
small p-co-safety	12	4	33.33%	10	83.33%	12	<b>100.00%</b>
large	15	1	6.67%	8	53.33%	14	<b>93.33%</b>
large safety	13	0	0.00%	13	<b>100.00%</b>	13	<b>100.00%</b>
large co-safety	13	3	23.08%	13	<b>100.00%</b>	13	<b>100.00%</b>
lily	9	1	11.11%	4	44.44%	5	<b>55.56%</b>
ltl2dba	8	0	0.00%	0	0.00%	5	<b>62.50%</b>
ltl2dpa	11	0	0.00%	0	0.00%	6	<b>54.55%</b>
<b>total</b>	<b>132</b>	<b>21</b>	<b>15.91%</b>	<b>90</b>	<b>68.18%</b>	<b>118</b>	<b>89.39%</b>