

Descriptive Complexity: Survey and Recent Progress

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- ▶ Descriptive Complexity

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- ▶ Dichotomy

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- ▶ Dichotomy
- ▶ Dynamic Complexity

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- ▶ Dynamic Complexity
- ▶ SAT Solvers

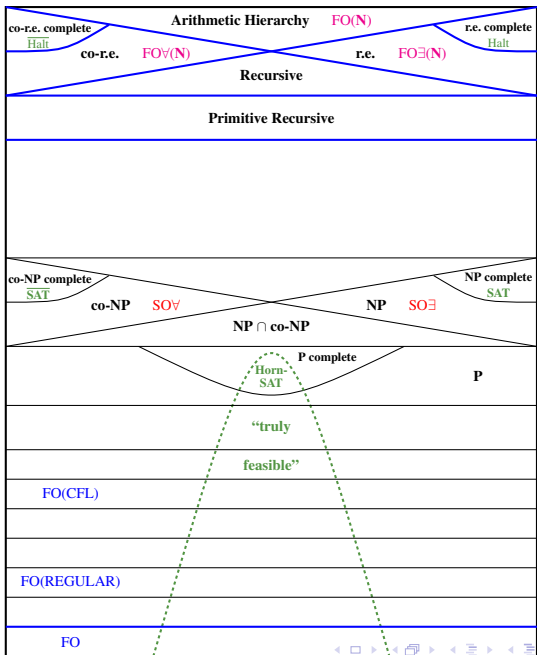
- ▶ Descriptive Complexity
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- ▶ SAT Solvers
- ▶ Computer Software: Crisis and Opportunity

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Personal perspective

$$P = \bigcup_{k=1}^{\infty} \text{DTIME}[n^k]$$

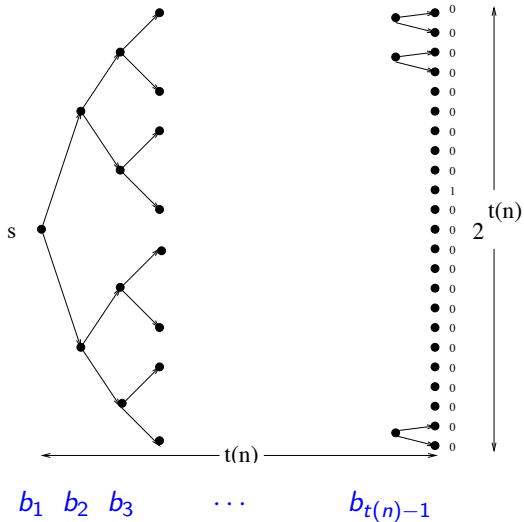
P is a good mathematical wrapper for “truly feasible”.



NTIME[$t(n)$]: a mathematical fiction

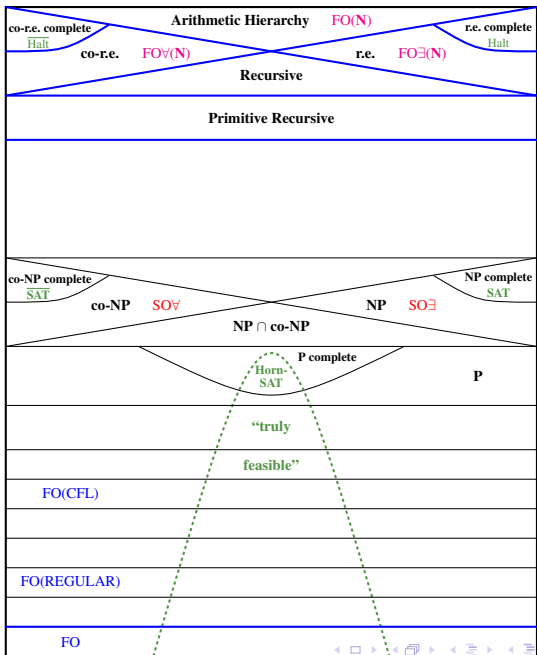
input w

$$|w| = n$$

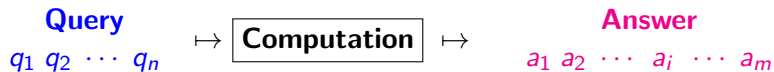


$$NP = \bigcup_{k=1}^{\infty} NTIME[n^k]$$

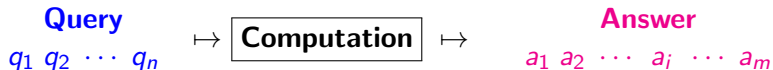
Many optimization problems we want to solve are NP complete.



Descriptive Complexity



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Restrict attention to the complexity of computing individual bits of the output, i.e., **decision problems**.

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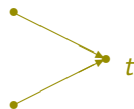
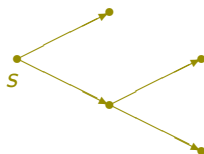
How rich a language do we need to **express** property S ?

There is a **constructive isomorphism** between these two approaches.

Interpret Input as Finite Logical Structure

Graph

$$G = (\{v_1, \dots, v_n\}, E, s, t)$$



Binary

$$\mathcal{A}_w = (\{p_1, \dots, p_8\}, S^{\mathcal{A}_w} = \{p_2, p_5, p_7, p_8\})$$

String

$$w = 01001011$$

Relational Database

$$D = (U, R_1^D, \dots, R_k^D)$$

$$\text{Vocabularies: } \tau_g = (E^2, s, t), \quad \tau_s = (S^1), \quad \tau_d = (R_1^{a_1}, \dots, R_k^{a_k})$$

First-Order Logic

input symbols: from τ

variables: x, y, z, \dots

boolean connectives: \wedge, \vee, \neg

quantifiers: \forall, \exists

numeric symbols: $=, \leq, +, \times, \min, \max$

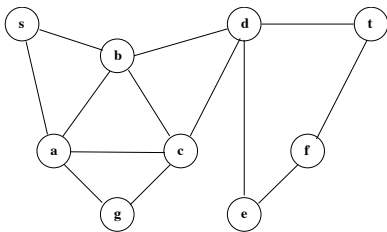
$$\alpha \equiv \forall x \exists y (E(x, y)) \quad \in \mathcal{L}(\tau_g)$$

$$\beta \equiv \exists x \forall y (x \leq y \wedge S(x)) \quad \in \mathcal{L}(\tau_s)$$

$$\beta \equiv S(\min) \quad \in \mathcal{L}(\tau_s)$$

Second-Order Logic

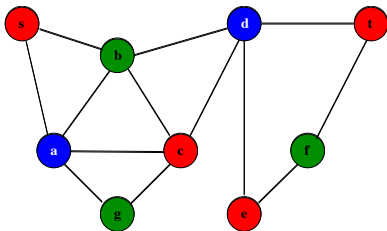
$$\begin{aligned}\Phi_{3\text{-color}} \equiv & \exists R^1 G^1 B^1 \forall x y ((R(x) \vee G(x) \vee B(x)) \wedge \\ & (E(x, y) \rightarrow (\neg(R(x) \wedge R(y)) \wedge \neg(G(x) \wedge G(y)) \\ & \wedge \neg(B(x) \wedge B(y))))))\end{aligned}$$



Second-Order Logic

Fagin's Theorem: NP = SO \exists

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Addition is First-Order

$$Q_+ : \text{STRUC}[\tau_{AB}] \rightarrow \text{STRUC}[\tau_s]$$

$$\begin{array}{rcccccc} A & & a_1 & a_2 & \dots & a_{n-1} & a_n \\ B & + & b_1 & b_2 & \dots & b_{n-1} & b_n \\ \hline S & & s_1 & s_2 & \dots & s_{n-1} & s_n \end{array}$$

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$$C(i) \equiv (\exists j > i) \left(A(j) \wedge B(j) \wedge \right. \\ \left. (\forall k. j > k > i) (A(k) \vee B(k)) \right)$$

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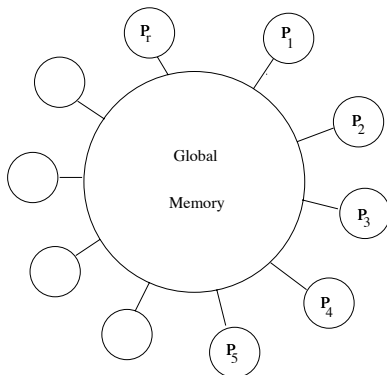
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$$C(i) \equiv (\exists j > i) \left(A(j) \wedge B(j) \wedge \right. \\ \left. (\forall k. j > k > i) (A(k) \vee B(k)) \right)$$

$$Q_+(i) \equiv A(i) \oplus B(i) \oplus C(i)$$

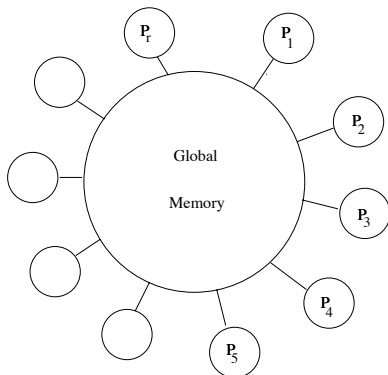
Parallel Machines:

$$\text{CRAM}[t(n)] = \text{CRCW-PRAM-TIME}[t(n)] - \text{HARD}[n^{O(1)}]$$



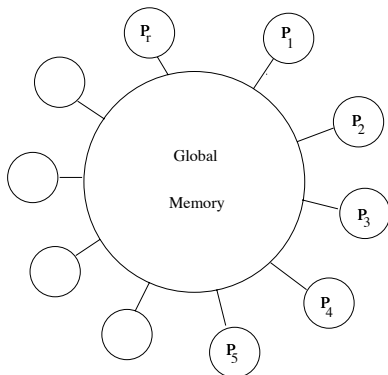
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Assume array $A[x] : x = 1, \dots, r$ in memory.

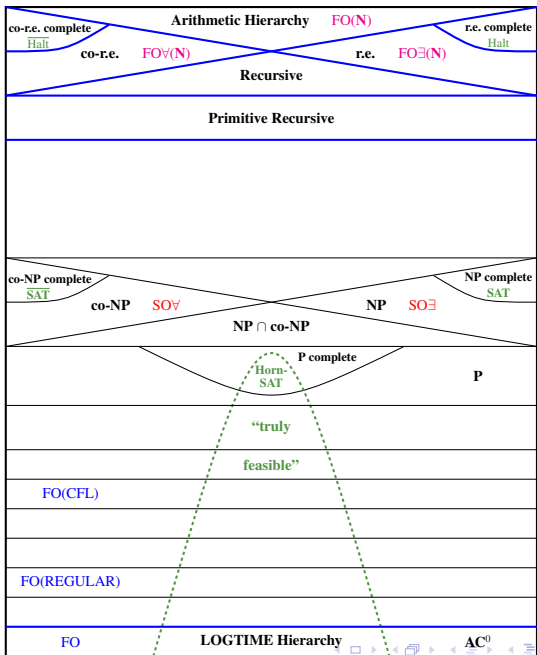


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Assume array $A[x] : x = 1, \dots, r$ in memory.

$$\forall x(A(x)) \equiv \text{write}(1); \text{proc } p_i : \text{if } (A[i] = 0) \text{ then write}(0)$$


FO
 =
 CRAM[1]
 =
 AC⁰
 =
 Logarithmic-
 Time
 Hierarchy



CRAM[$t(n)$] = concurrent parallel random access machine;
polynomial hardware, parallel time $O(t(n))$

IND[$t(n)$] = first-order, depth $t(n)$ inductive definitions

FO[$t(n)$] = $t(n)$ repetitions of a block of restricted quantifiers:

QB = $[(Q_1 x_1 . M_1) \cdots (Q_k x_k . M_k)]$; M_i quantifier-free

$\varphi_n = \underbrace{[\text{QB}][\text{QB}] \cdots [\text{QB}]}_{t(n)} M_0$

Thm: For all constructible, polynomially bounded $t(n)$,

$$\text{CRAM}[t(n)] = \text{IND}[t(n)] = \text{FO}[t(n)]$$

Thm: For all $t(n)$, even beyond polynomial,

$$\text{CRAM}[t(n)] = \text{FO}[t(n)]$$

For $t(n)$ poly
bdd,

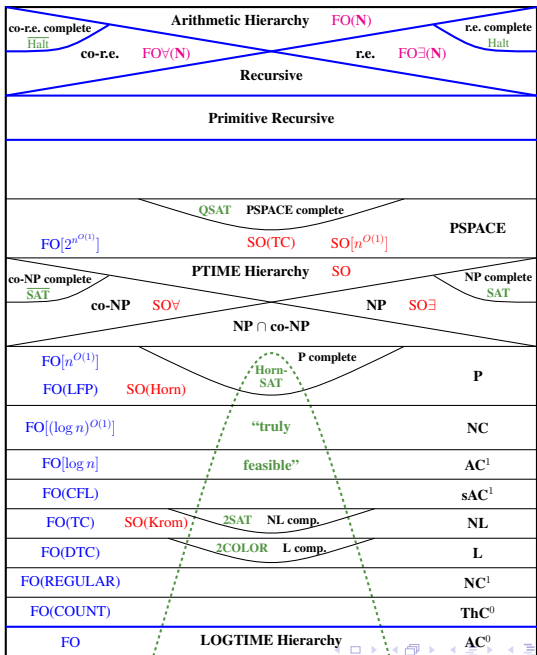
CRAM[$t(n)$]

=

IND[$t(n)$]

=

FO[$t(n)$]



Recent Breakthroughs in Descriptive Complexity

Theorem [Ben Rossman] Any first-order formula with any numeric relations ($\leq, +, \times, \dots$) that means “I have a clique of size k ” must have at least $k/4$ variables.

- ▶ Creative new proof idea using Håstad’s Switching Lemma gives the essentially optimal bound.
- ▶ First lower bound of its kind for number of variables with ordering.
- ▶ This lower bound is for a fixed formula, if it were for a sequence of polynomially-sized formulas, it would show that $\text{CLIQUE} \notin \text{P}$ and thus $\text{P} \neq \text{NP}$.

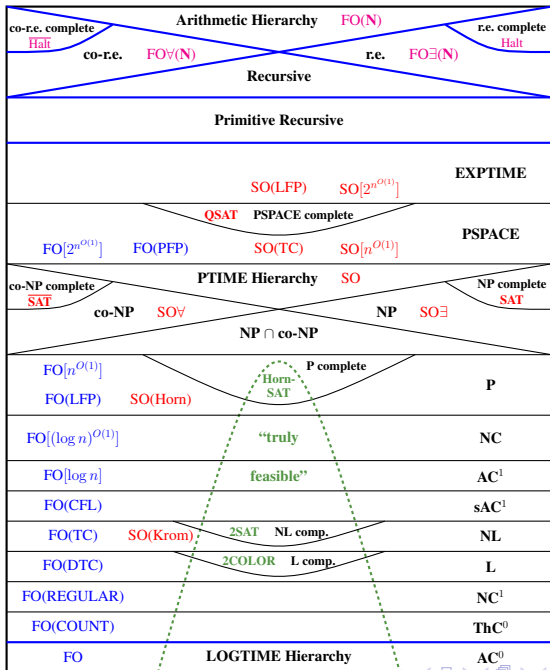
Recent Breakthroughs in Descriptive Complexity

Theorem [Martin Grohe] Fixed-Point Logic with Counting captures Polynomial Time on all classes of graphs with excluded minors.

Grohe proves that for every class of graphs with excluded minors, there is a constant k such that two graphs of the class are isomorphic iff they agree on all k -variable formulas in fixed-point logic with counting.

Thus every class of graphs with excluded minors admits the same general polynomial time canonization algorithm: we're isomorphic iff we agree on all formulas in C_k and in particular, you are isomorphic to me iff your C_k canonical description is equal to mine.

See: “**The Nature and Power of Fixed-Point Logic with Counting**” by **Anuj Dawar** in *SigLog Newsletter*.



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- ▶ Schaefer; Feder-Vardi: CSP Dichotomy Conjecture
- ▶ Tremendous progress using Universal Algebra. (Solved for domains of size 2 and 3, and for undirected graphs.)
See: **“Constraint Satisfaction Problem and Universal Algebra”** by **Libor Barto** in *SigLog Newsletter*.

Static

1. Read entire input
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Dynamic

1. Long series of Inserts, Deletes, Changes, and, Queries
2. On **query**, very quickly compute $Q(\text{current database})$
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4. What **additional information** should we maintain? —
auxiliary data structure

Dynamic (Incremental) Applications

- ▶ Databases
- ▶ LaTeXing a file
- ▶ Performing a calculation
- ▶ Processing a visual scene
- ▶ Understanding a natural language
- ▶ Verifying a circuit
- ▶ Verifying and compiling a program
- ▶ Surviving in the wild

Parity

Current Database: S	Request	Auxiliary Data: b
0000000		0

Parity

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0000000		0
	ins (3,S)	

Parity

Current Database: S	Request	Auxiliary Data: b
0000000		0
0010000	ins (3, S)	1

Parity

Current Database: S	Request	Auxiliary Data: b
0000000		0
0010000	ins (3, S)	1
	ins (7, S)	

Parity

Current Database: S	Request	Auxiliary Data: b
0000000		0
0010000	ins (3, S)	1
0010001	ins (7, S)	0

Parity

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0000000		0
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0010001	ins (7, S)	0
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ins(a,S)

$$S'(x) \equiv S(x) \vee x = a$$

$$b' \equiv (b \wedge S(a)) \vee (\neg b \wedge \neg S(a))$$

del(a,S)

$$S'(x) \equiv S(x) \wedge x \neq a$$

$$b' \equiv (b \wedge \neg S(a)) \vee (\neg b \wedge S(a))$$

Parity

- ▶ Does binary string w have an odd number of 1's?
- ▶ **Static:** $\text{TIME}[n]$, $\text{FO}[\Omega(\log n / \log \log n)]$
- ▶ **Dynamic:** $\text{Dyn-TIME}[1]$, Dyn-FO

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REACH_u

- ▶ Is t reachable from s in undirected graph G ?
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REACH_u

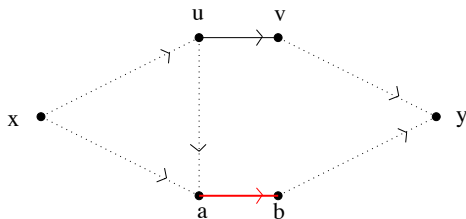
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Minimum Spanning Trees, k -edge connectivity, ...

Fact: [Dong & Su] $\text{REACH}(\text{acyclic}) \in \text{DynFO}$

ins(a, b, E) : $P'(x, y) \equiv P(x, y) \vee (P(x, a) \wedge P(b, y))$

del(a, b, E):



$$\begin{aligned} P'(x, y) \equiv & P(x, y) \wedge \left[\neg(P(x, a) \wedge P(b, y)) \right. \\ & \vee (\exists uv)(P(x, u) \wedge E(u, v) \wedge P(v, y) \\ & \left. \wedge P(u, a) \wedge \neg P(v, a) \wedge (a \neq u \vee b \neq v)) \right] \end{aligned}$$

REACHABILITY Problems

$$\text{REACH} = \{G \mid G \text{ directed, } s \xrightarrow[G]{*} t\} \quad \text{NL}$$

$$\text{REACH}_d = \{G \mid G \text{ directed, outdegree} \leq 1 \text{ } s \xrightarrow[G]{*} t\} \quad \text{L}$$

$$\text{REACH}_u = \{G \mid G \text{ undirected, } s \xrightarrow[G]{*} t\} \quad \text{L}$$

$$\text{REACH}_a = \{G \mid G \text{ alternating, } s \xrightarrow[G]{*} t\} \quad \text{P}$$

Facts about dynamic REACHABILITY Problems:

$\text{Dyn-REACH}(\text{acyclic}) \in \text{Dyn-FO}$ [DS]

$\text{Dyn-REACH}_d \in \text{Dyn-QF}$ [H]

$\text{Dyn-REACH}_u \in \text{Dyn-FO}$ [PI]

$\text{Dyn-REACH} \in \text{Dyn-FO}(\text{COUNT})$ [H]

$\text{Dyn-PAD}(\text{REACH}_a) \in \text{Dyn-FO}$ [PI]

Reachability is in DynFO

by Samir Datta, Raghav Kulkarni, Anish Mukherjee, Thomas Schwentick and Thomas Zeume

<http://arxiv.org/abs/1502.07467>

They show that Matrix Rank is in DynFO and REACH reduces to Matrix Rank.

Thm. 1 [Hesse] Reachability of functional DAG is in DynQF.

proof: Maintain E, E^*, D (outdegree = 1).

Insert $E(i, j)$: (ignore if adding edge violates outdegree or acyclicity)

$$E'(x, y) \equiv E(x, y) \vee (x = i \wedge y = j)$$

$$D'(x) \equiv D(x) \vee x = i$$

$$E^{*'}(x, y) \equiv E^*(x, y) \vee (E^*(x, i) \wedge E^*(j, y))$$

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Delete $E(i, j)$:

$$E'(x, y) \equiv E(x, y) \wedge (x \neq i \vee y \neq j)$$

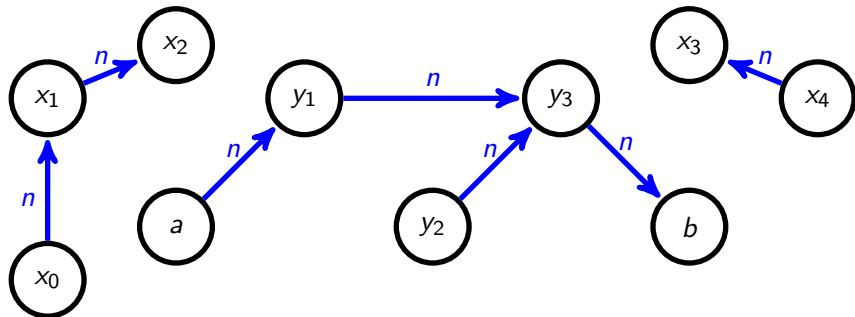
$$D'(x) \equiv D(x) \wedge (x \neq i \vee \neg E(i, j))$$

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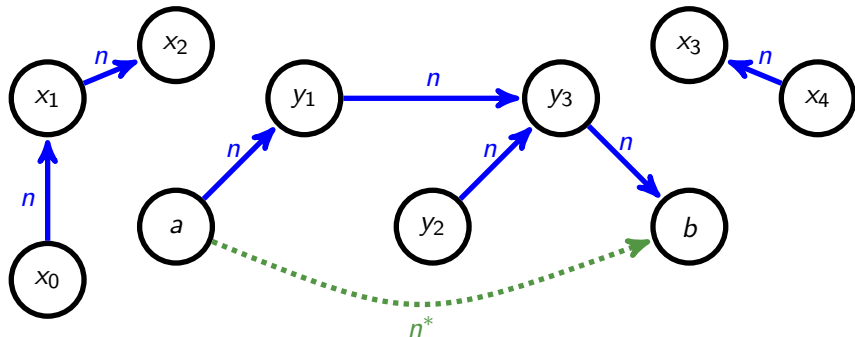
Dynamic Reasoning

Reasoning About reachability – can we get to b from a by following a sequence of pointers – is **crucial for proving that programs meet their specifications**.



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However reasoning about reachability in general is **undecidable**.

Ideas:

- ▶ Can express tilings and thus runs of Turing Machines.
- ▶ Even worse, can express **finite path** and thus **finite** and thus **standard natural numbers**. Thus $\text{FO}(\text{TC})$ is as hard as the Arithmetic Hierarchy [Avron].

For the time being, let's restrict ourselves to acyclic fields which thus also generate a linear ordering of all points reachable from a given point.

$$\text{acyclic} \equiv \forall xy (n^*(x, y) \wedge n^*(y, x) \rightarrow x = y)$$

$$\text{transitive} \equiv \forall xyz (n^*(x, y) \wedge n^*(y, z) \rightarrow n^*(x, z))$$

$$\text{linear} \equiv \forall xyz (n^*(x, y) \wedge n^*(x, z) \rightarrow n^*(y, z) \vee n^*(z, y))$$

Effectively-Propositional Reasoning about Reachability in Linked Data Structures

- ▶ Automatically transform a program manipulating linked lists to an $\forall\exists$ correctness condition.
- ▶ Using Hesse's dynQF algorithm for $REACH_d$, is that these $\forall\exists$ formulas are closed under weakest precondition.
- ▶ Using acyclic, transitive and linear axioms, the negation of the correctness condition is equi-satisfiable with a propositional formula.
- ▶ use a SAT solver to automatically prove correctness or find counter-example runs, typically in under 3 seconds per program.

Thm. 2 [Hesse] Reachability of functional graphs is in DynQF.

proof idea: If adding an edge, e , would create a cycle, then we maintain relation P – the path relation without the edge completing the cycle – as well as E^* , E and D .

Surprisingly this can all be maintained via quantifier-free formulas, **without remembering which edges we are leaving out** in computing P . □

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Surprisingly this can all be maintained via quantifier-free formulas, **without remembering which edges we are leaving out** in computing P . □

Using Thm. 2, the above methodology has been extended to cyclic deterministic graphs.

- ▶ Itzhaky, Banerjee, Immerman, Aleks Nanevski, Sagiv, “Effectively-Propositional Reasoning About Reachability in Linked Data Structures” CAV 2013.
- ▶ Itzhaky, Banerjee, Immerman, Lahav, Nanevski, Sagiv, “Modular Reasoning about Heap Paths via Effectively Propositional Formulas”, POPL 2014

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- ▶ They provably aren't good on all instances, but they do extremely well in practice.
- ▶ Thus we have a general purpose problem solver.
- ▶ Very useful for checking the correctness of programs, automatically finding counter-example runs, and for synthesizing good code from specifications.

- ▶ Software is everywhere, controlling more and more of our lives.

Software Crisis

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- ▶ Thank you!

co-r.e. complete Halt	Arithmetic Hierarchy		FO(N)	r.e. complete Halt
co-r.e.	FO\forall(N)		r.e.	FO\exists(N)
Recursive				
Primitive Recursive				
				EXPTIME
		SO(LFP)	SO$[2^{n^{O(1)}}]$	
		QSAT	PSPACE complete	
FO$[2^{n^{O(1)}}]$	FO(PFP)	SO(TC)	SO$[n^{O(1)}]$	PSPACE
co-NP complete SAT	PTIME Hierarchy		SO	NP complete SAT
co-NP	SO\forall		NP	SO\exists
NP \cap co-NP				
FO$[n^{O(1)}]$			P complete	P
FO(LFP)	SO(Horn)	Horn-SAT		
FO$[(\log n)^{O(1)}]$			"truly feasible"	NC
FO$[\log n]$				AC¹
FO(CFL)				sAC¹
FO(TC)	SO(Krom)	2SAT	NL comp.	NL
FO(DTC)			2COLOR	L
FO(REGULAR)				NC¹
FO(COUNT)				ThC⁰
FO	LOGTIME Hierarchy			AC⁰