

Power of reasoning over richer domains

Antonina Kolokolova (MUN)

NII Shonan meeting
“Logic and computational complexity”,
Sep 21, 2017

VNC¹



$$\begin{pmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{pmatrix}$$

Problem representations

Given a problem with its instance(s), how to state it to make it easier to solve?

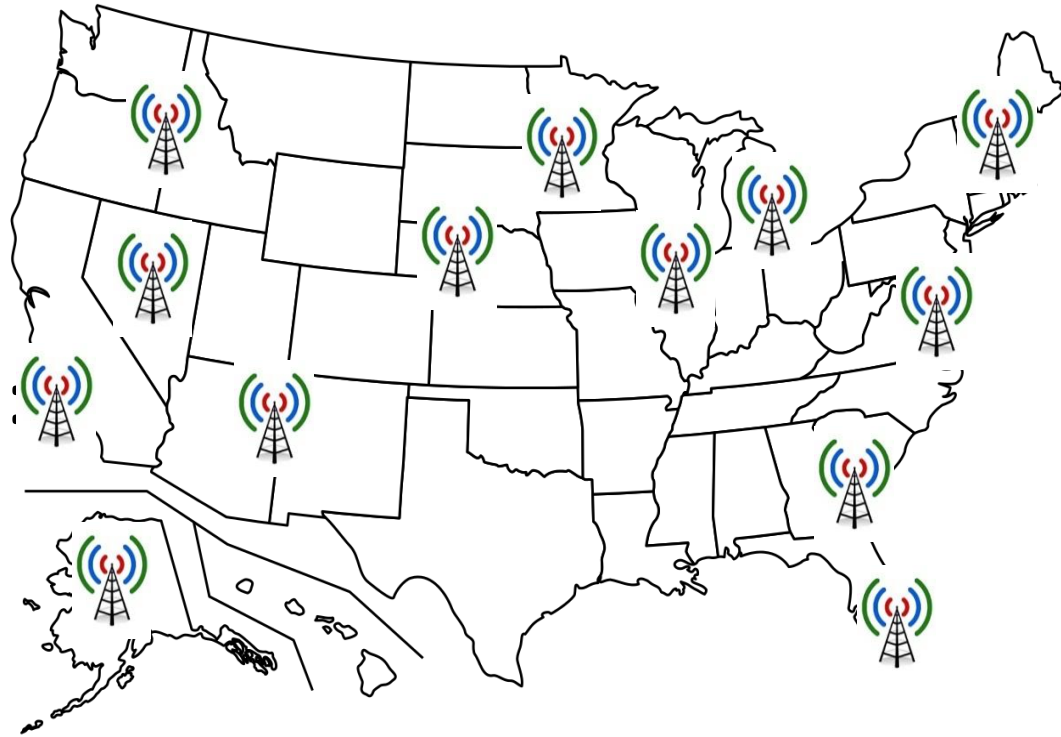


Example: SAT vs. Integer LP

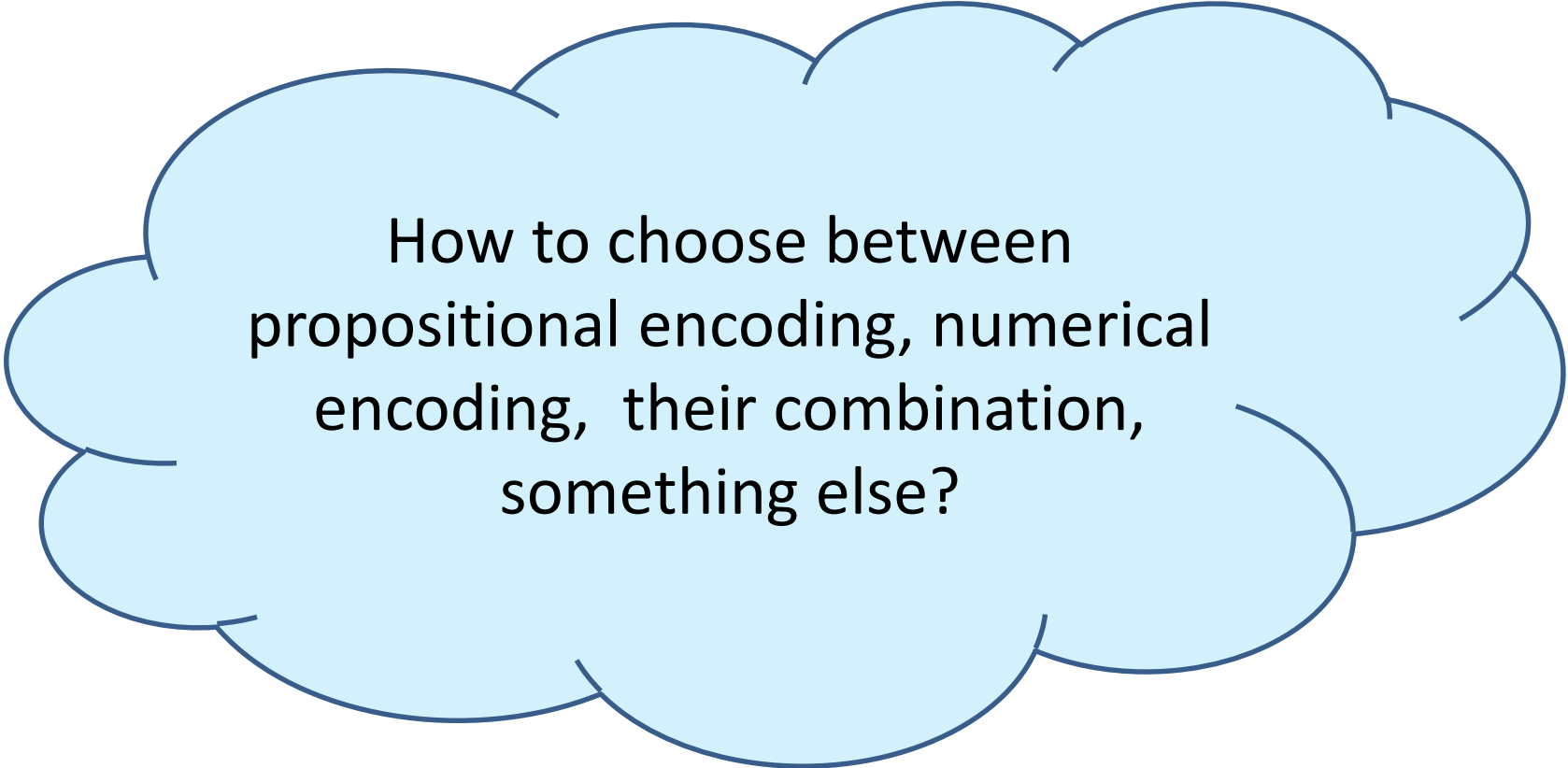
- FCC spectrum auction:
 - Essentially a colouring problem
 - ILP: poor
 - SAT: good

- TravelingSalesman:

- ILP: good
- SAT: poor



Problem representations



How to choose between
propositional encoding, numerical
encoding, their combination,
something else?

Combinatorial vs. algebraic proofs

- Algebraic

- Uses algebraic concepts

- determinants, eigenvalues...

- Relies on their properties for analysis

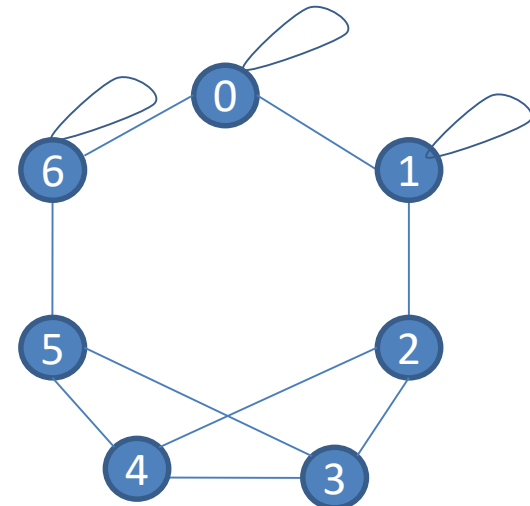
$$\begin{pmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{pmatrix}$$

- Combinatorial

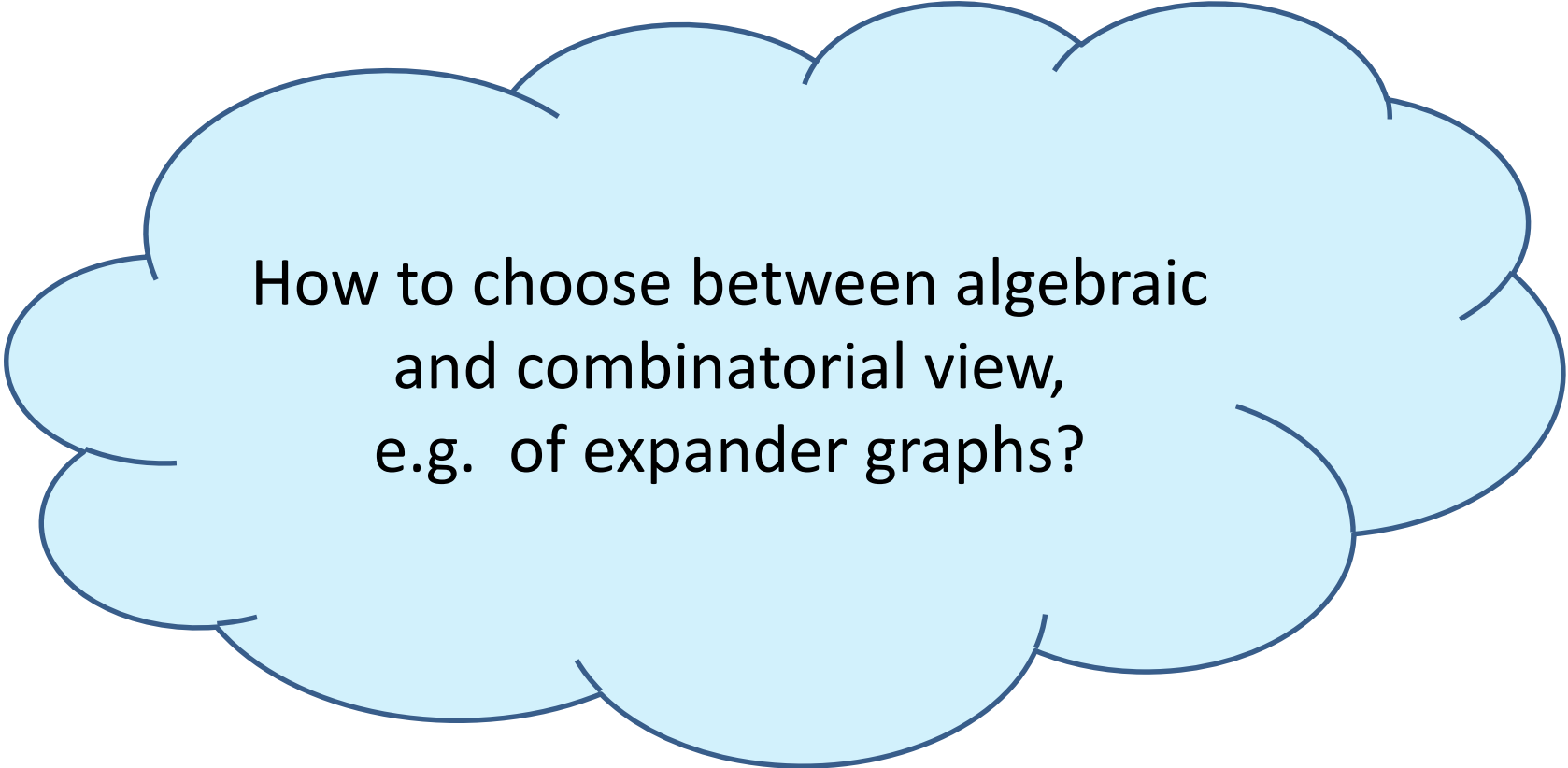
- Uses “simple to define” properties

- Avoids algebra even in proofs

- Algorithms of lower complexity!



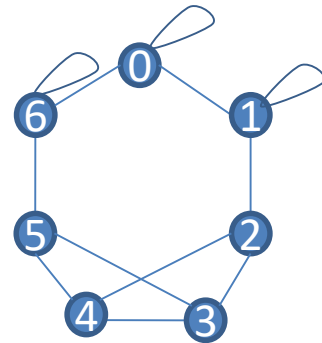
Problem representations

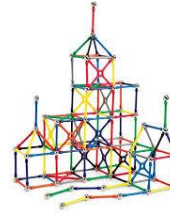
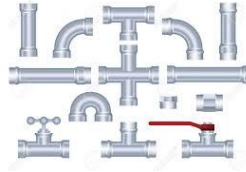


How to choose between algebraic
and combinatorial view,
e.g. of expander graphs?

Some results

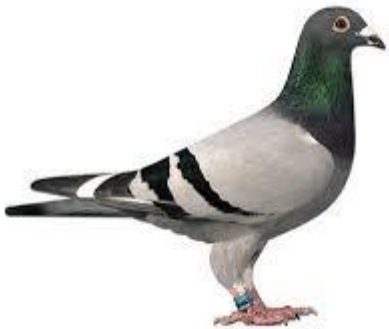
- Proof complexity of SMT:
 - Combining resolution with theories over underlying domain
 - Linear arithmetic, equality, uninterpreted functions with equality (EUF)...
 - Models satisfiability modulo theories solvers like resolution models SAT solvers
 - With EUF, can polynomially simulate Frege.
- Complexity of expander-based reasoning:
 - Can prove existence of expander graphs using purely combinatorial reasoning.
 - Corollary: monotone Frege is as powerful as non-monotone.





Proof complexity of Satisfiability Modulo Theories

with Robert Robere and Vijay Ganesh







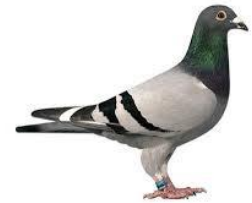
PigeonHolePrinciple



- PigeonHole Principle: there is no injective function from $[n]$ to $[n-1]$

- PHP:

$$\bigwedge_{i \leq n} (\bigvee_{j < n} p_{i,j}) \wedge \bigwedge_{i \neq k, j} (\neg p_{i,j} \vee \neg p_{k,j})$$



- ==-PHP:

$$\bigwedge_{i \leq n} (\bigvee_{j < n} (p_i = h_j)) \wedge \bigwedge_{i < k \leq n} (p_i \neq p_k)$$



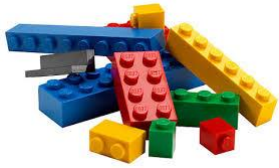
- EUF-PHP:

$$\bigwedge_{x \in [n]} (f(x) \neq 0) \wedge \bigwedge_{x, y \in [n]} (x \neq y \rightarrow f(x) \neq f(y))$$

- LA-PHP:

$$\bigwedge_{i \leq n} (\sum_{j < n} x_{i,j} \geq 1) \wedge \bigwedge_{\substack{i, k \leq n, \\ j < n}} (x_{i,j} + x_{k,j} \leq 1)$$





PigeonHolePrinciple



- PigeonHole Principle: there is no injective function from $[n]$ to $[n-1]$

- PHP:

$$\bigwedge_{i \leq n} (\bigvee_{j < n} p_{i,j}) \wedge \bigwedge_{i \neq k, j} (\neg p_{i,j} \vee \neg p_{k,j})$$

- =-PHP:

$$\bigwedge_{i \leq n} (\bigvee_{j < n} (p_i = h_j)) \wedge \bigwedge_{i < k \leq n} (p_i \neq p_k)$$

- EUF-PHP:

$$\bigwedge_{x \in [n]} (f(x) \neq 0) \wedge \bigwedge_{x, y \in [n]} (x \neq y \rightarrow f(x) \neq f(y))$$

- LA-PHP:

$$\bigwedge_{i \leq n} (\sum_{j < n} x_{i,j} \geq 1) \wedge \bigwedge_{\substack{i, k \leq n, \\ j < n}} (x_{i,j} + x_{k,j} \leq 1)$$

- Propositional

- Theory of equality:

- $(a = b \wedge b = c \rightarrow a = c)$

- Equality with uninterpreted functions (EUF)

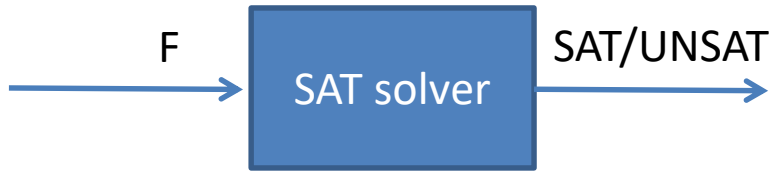
- equality axioms

- Ackermann axioms: $(a = b \rightarrow f(a) = f(b))$

- Linear arithmetic

SAT vs. SMT

Propositional



- PHP:

$$\bigwedge_{i \leq n} (\bigvee_{j < n} p_{i,j}) \wedge \bigwedge_{i \neq k, j} (\neg p_{i,j} \vee \neg p_{k,j})$$

- =-PHP:

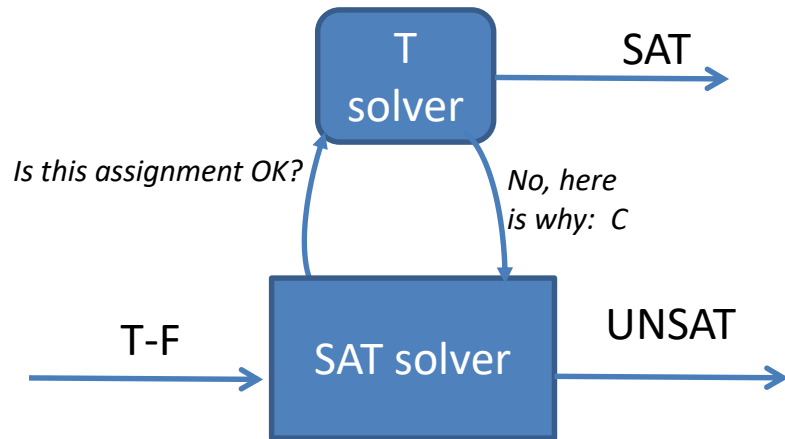
$$\bigwedge_{i \leq n} (\bigvee_{j < n} (p_i = h_j)) \wedge \bigwedge_{i < k \leq n} (p_i \neq p_k)$$

- EUF-PHP:

$$\bigwedge_{x \in [n]} (f(x) \neq 0) \wedge \bigwedge_{x, y \in [n]} (x \neq y \rightarrow f(x) \neq f(y))$$

- LA-PHP:

$$\bigwedge_{i \leq n} (\sum_{j < n} x_{i,j} \geq 1) \wedge \bigwedge_{\substack{i, k \leq n, \\ j < n}} (x_{i,j} + x_{k,j} \leq 1)$$

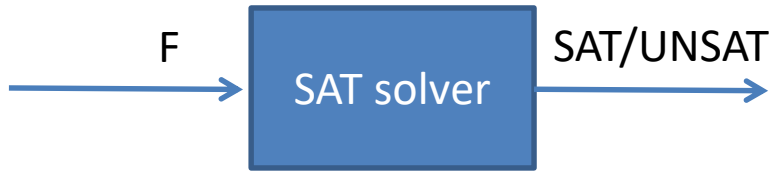


For which theory T
can a SAT solver
with a T solver
simulate Extended Frege?



Res(T)

Propositional



Resolution

Res(T):

Literals are atoms of the theory.

Rules of inference:

1. Resolution rule

$$\frac{(C \vee x) \quad (D \vee \neg x)}{(C \vee D)}$$

2. Clauses derivable from T:

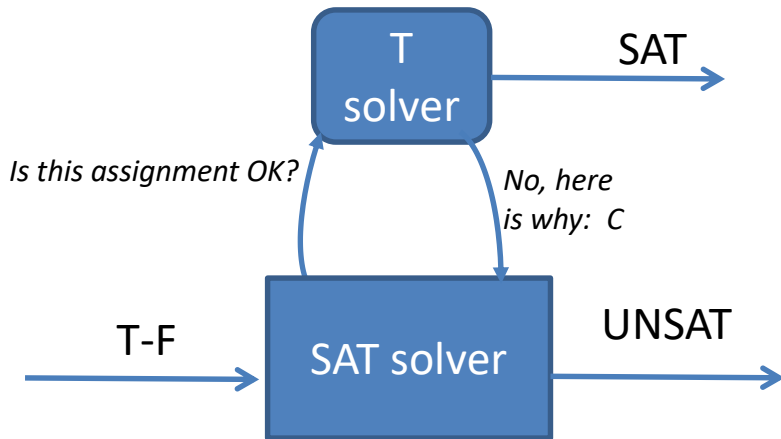
- Eg: T is a theory of equality:

- $(a \neq b \vee b \neq c \vee a = c)$

- Eg: T is linear arithmetic:

- $(a \leq c \vee b \leq d \vee a + b > c + d)$

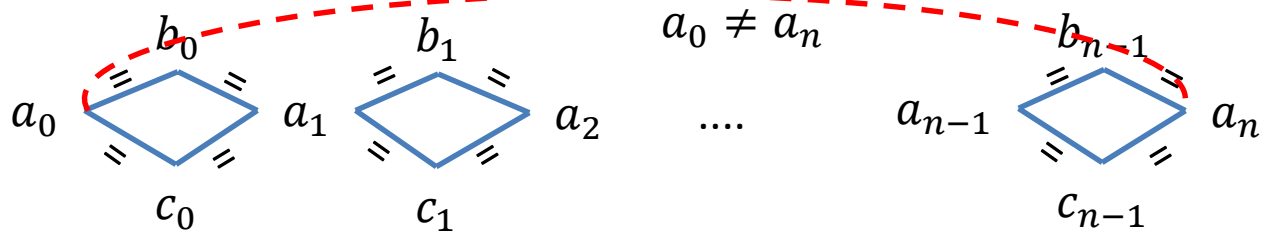
Can introduce new literals



Res(T)

New literals

- Theory solver has to be able to return a clause using literals not in the original formula:
 - if F contained $a=b$ and $b=c$, T returns a clause $(a \neq b \vee b \neq c \vee$



Res(T) vs. SMT solvers

- CDCL (conflict-driven clause learning with restarts)
 - Repeat:
 - Assign some variables
 - Do unit clause propagation (set literals in unit clauses)
 - If there is an unsatisfied clause, backtrack and learn the conflict as a clause
 - Maybe restart, removing variable assignment, but keeping learned clauses
- CDCL(T):
 - Also check whether assignment makes sense for T
 - If not, learn a conflict clause.
- Resolution captures CDCL
 - Pipatsrisawat/Darwiche'11.
- Res(T) captures CDCL(T)
 - Generalizing Pipatsrisawat/Darwiche'11.

Power of Res(T)

- Res(Theory of Equality) is no more powerful than Resolution
 - Add all n^3 equality axioms to F, then solve.
- Res(LA) polynomially simulates R(lin)
- Resolution over Equality with Uninterpreted Functions theory, Res(EUF), can effectively p-simulate Frege.
 - Conjunctions of EUF atoms are decidable in $O(n \log n)$ time!
 - Using a variant of Union-Find algorithm.

Equality with uninterpreted functions theory (EUF)

- Signature:
 - uninterpreted function symbols of bounded arity
 - constants a, b, c, \dots
- Terms: constants, and inductively $f(\bar{t})$ for functions.
- Atoms: equalities/disequalities over terms: $t_1 = t_2, t_1 \neq t_2$
- Formulas: conjunctions of atoms

$$(f(a) = b) \wedge (b = c) \wedge (g(f(a)) \neq c)$$

- Axioms:
 - Equality: $(a = b \wedge b = c \rightarrow a = c)$
 - Ackermann: $\bar{a} = \bar{b} \rightarrow f(\bar{a}) = f(\bar{b})$
- Can decide in near-linear time if a given EUF formula is satisfiable:
 - Downey-Sethi-Tarjan congruence closure (based on Union-Find)

Sequent calculus (LK)

- Equivalent to Frege.
 - Natural deduction
- Sequents: $A_1, \dots, A_n \longrightarrow B_1, \dots, B_m$
 - $A_1 \wedge \dots \wedge A_n \rightarrow B_1 \vee \dots \vee B_m$
 - Axioms $A \rightarrow A, 0 \rightarrow S, S \rightarrow 1$.
 - Rules for \vee, \wedge, \neg and cut

$$\frac{F \rightarrow G, A \quad A, F \rightarrow G}{F \rightarrow G}$$

$$\frac{F \rightarrow G, A}{\neg A, F \rightarrow G}$$

$$\frac{F \rightarrow G, A \quad F \rightarrow G, B}{F \rightarrow G, A \wedge B}$$

$$\frac{A, B, F \rightarrow G}{A \wedge B, F \rightarrow G}$$

- Proof size: total number of symbols.

Res(EUF) simulates LK

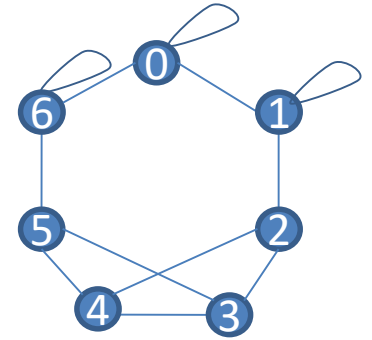
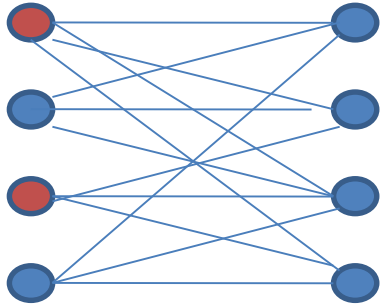
- Suppose there is an LK proof of $F \rightarrow 0$
 - An LK-refutation of F
- Add to F :
 - Two constants: $e_0 \neq e_1$
 - Definitions of N , O , A (and, or, not):
 - $N(e_0) = e_1, N(e_1) = e_0, O(e_1, e_0) = e_1, \dots$
 - Bounded variable range: $\bigwedge (x_i = e_0 \vee x_i = e_1)$
- Now simulate an LK proof by constructing terms for all formulas in the proof inductively
 - Prove that at each step of LK proof: $A_1 \dots A_k \rightarrow B_1 \dots B_\ell$
 - Either one of the A terms is e_0 or one of the B terms is e_1
 - Also for each subformula in proof so far, its term = e_0 or = e_1

Open problems

- Is it better to use SMT than propositionalize completely? If so, when?
 - Flatten:
 - replace nested terms by new variables
 - Bit blast:
 - represent each variable by $\log n$ bits.
 - add all relevant axioms explicitly.
- How to choose T given a problem and class of instances?
 - And how to choose T -representation?

For which theory T would $\text{Res}(T)$ effectively p -simulate Extended Frege?





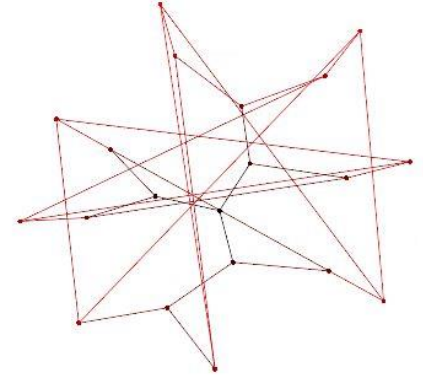
Complexity of Expander-Based Reasoning and the Power of Monotone Proofs

with Sam Buss, Valentine Kabanets
and Michal Koucky

VNC¹

$$\begin{pmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{pmatrix}$$

Expander graphs



- Graphs which are both
 - sparse (usually constant degree)
 - and well connected (log length path between any two points).
- Expander graphs are pseudorandom objects. A random graph is an expander with high probability.
- Random walk on an expander converges fast.

Uses of expanders

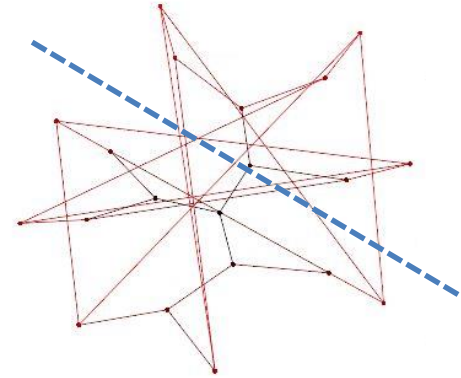
- As pseudorandom objects
 - One-way functions of Goldreich'2000
 - Cryptographic hash functions: Charles/Goren/Lauter...
 - Error-correcting codes, derandomization...
- In complexity theory
 - Reingold and Rozenman/Vadhan: USTCON in LogSpace
 - Dinur: combinatorial proof of the PCP theorem
 - Ajtai/Komlos/Szemerédi: AKS sorting networks

Combinatorial definition of expanders

- d-regular undirected (multi)graphs
- Edge expansion:
 - min fraction of edges crossing a cut (normalized by smaller side size).

$$- h(G) = \min_{\{\emptyset \neq U, |U| \leq \frac{n}{2}\}} \frac{|E(U, U^c)|}{|U|}$$

- Expander: $h(G)$ is constant.



Algebraic definition of expanders

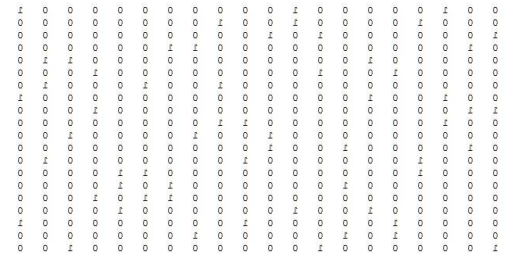
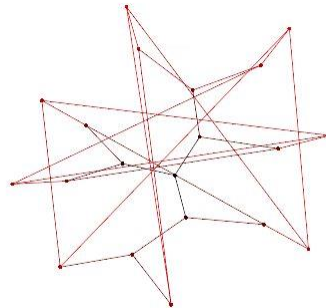
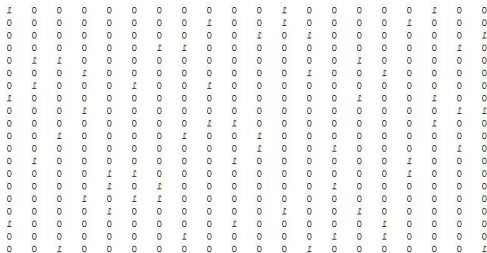
1	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	1	0	0	0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
0	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
0	1	0	0	0	0	1	0	0	1	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
0	0	0	0	1	0	0	0	0	0	1	1	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0
0	0	0	0	1	1	0	0	0	0	0	0	0	0	0	0	0	1	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	1
0	0	1	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0

- Spectral gap: $d - \lambda_2$,
 - d is the degree of G
 - λ_2 is the second largest eigenvalue of adjacency matrix M_G of G .
- Expander (λ -expander):
 - A graph that has a constant spectral gap.

Combinatorial vs. algebraic

- Cheeger inequality:

$$\frac{d - \lambda_2}{2} \leq h(G) \leq \sqrt{2d(d - \lambda_2)}$$



- So constant spectral gap \sim constant expansion
- Most proofs use algebraic definition
 - Some loss in parameters in combinatorial setting
- Combinatorial definition allows lower complexity algorithms

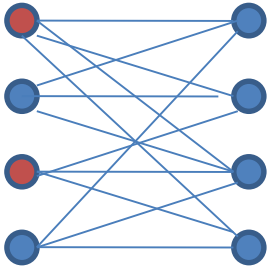
Formalizing “combinatorial”

- Take a system of reasoning which cannot define algebraic objects
 - No eigenvalues, determinants, etc
 - E.g., a system based on polynomial-size formulas (NC^1 -reasoning)
- Proofs in this system are combinatorial (unless algebra $\in NC^1$)
 - Combinatorial proofs of correctness of algorithms or existence of combinatorial objects.
 - Not known to prove $AB=I \Rightarrow BA=I$

Our results

- We give an NC^1 proof of existence of expander graphs of arbitrary size.
 - Includes a combinatorial analysis of a fully explicit expander construction.
 - And its formalization in an NC^1 theory
- **Corollary:** monotone proofs are as powerful as non-monotone.
 - Monotone LK polynomially simulates LK.
 - By adding the last piece to [Atserias-Galesi-Gavaldà'01, Atserias-Galesi-Pudlak'02, Jerabek'11]

Expander constructions



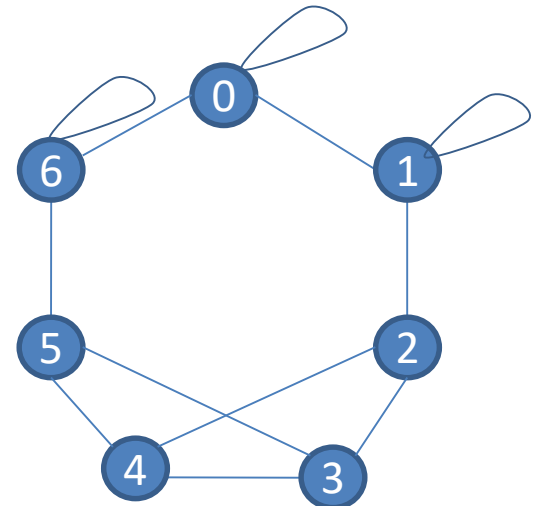
Example 1: Margulis, Gabber/Galil bipartite expanders.

$(x,y) \rightarrow (x,y), (x,x+y), (x,x+y+1), (x+y,y), (x+y+1,y)$

Example 2: (from Hoory/Linial/Wigderson)

\mathbb{Z}_p : for every $v \neq 0$, connect v to $v-1, v+1$ and v^{-1} .

For $v=0$, connect v to $0, 1$ and $p-1$.



Iterative constructions

- Start with constant-size expanders. Obtain a large size graph by repeatedly applying:
 - Powering (to increase expansion)
 - Zig-zag or replacement product (to reduce degree)
 - Tensoring (to grow quickly).
- Originally by [Reingold-Vadhan-Wigderson'02]
 - Zig-zag product. Proof uses spectral gap.
 - [Alon, Schwartz, Shapira'08] Replacement product with its combinatorial analysis.
- Explicit: given vertex v and $i \in [1 \dots d]$, produce (w, j) such that w is the i^{th} neighbour of v , and v is the j^{th} neighbour of w in resulting graph.
 - In time $O(\log |G|)$.

Our variant of the construction

- Start with $2d$ -regular G_0 with $h(G_0) = \epsilon = 1/1296$ and d -regular H , $h(H) = 1/3$.
- Apply the following $\sim \log n$ times:
 1. Add self-loops to double the degree; tensor with itself
 2. Add self-loops again and power to a constant c
 3. Replace each vertex with H .
- Each G_i has $h(G_i) = \epsilon$ and size $>$ squared.
- Fully explicit: NC^1 algorithm to compute k^{th} neighbour w of v in the final G , and its edge index j from (v,k)

Powering: $M_{G'} = M_G^k$

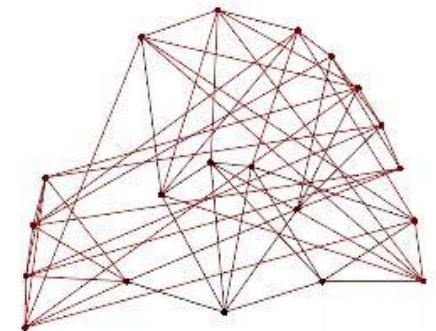
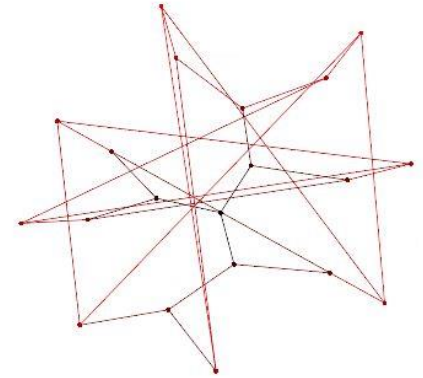
⊙ Easy with eigenvalues: $\lambda_2 \rightarrow \lambda_2^k$

⊙ Combinatorially, let $h(G)=\epsilon$.

- First, add d self-loops to G .
- Using [Mihail'89] mixing lemma
 - and mixing \rightarrow expansion

• Get $h(G') = \frac{1}{2} \left(1 - \left(1 - \frac{\epsilon^2}{4}\right)^{k/2}\right)$

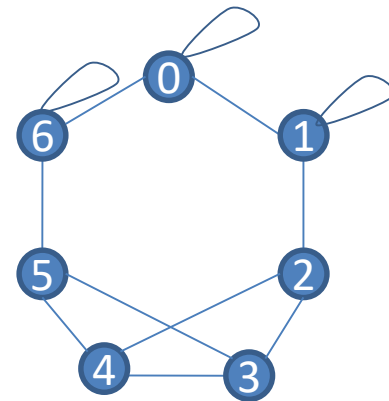
Proof : Cauchy-Schwartz and sums.



Mihail'89 mixing lemma

- A random walk on an expander converges to uniform distribution exponentially fast.
- *More precisely, let*
 - *G be a d -regular graph with edge expansion ϵ .*
 - *Add d self-loops to each vertex of G to obtain G'*
 - *A be a normalized adjacency matrix of G'*
 - *π be any distribution on vertices of G'*
 - *u the uniform distribution on vertices of G'*

- Then
$$\|A^k \pi - u\|^2 \leq \left(1 - \left(\frac{\epsilon^2}{4}\right)^k\right) \|\pi - u\|^2$$



Constructiveness

- For formalization, need an NC^1 algorithm:
 - Given a non-expanding set U' in G'
 - Produce a non-expanding set U in G .
- From [Mihail'89] proof:
 - Sort vertices in decreasing $\pi - u$ order
 - If some U' in G' is non-expanding, then so is a set of first k vertices in G for some k . Test which one.
 - Both sorting and testing are in NC^1

Formalizing “combinatorial”

- Bounded arithmetic:
 - Theories \sim complexity classes.
 - For a class C , a theory $V-C$ can reason about C -definable concepts (numbers and strings)
- Eigenvalues, determinants, etc are not known to be computable in NC^1
- So proofs in $V-NC^1$ are “combinatorial” in a strict sense.
 - If $V-NC^1$ cannot prove existence of eigenvalues
 - Then it cannot formalize proofs relying on eigenvalues, even in disguise.

Bounded arithmetic theories

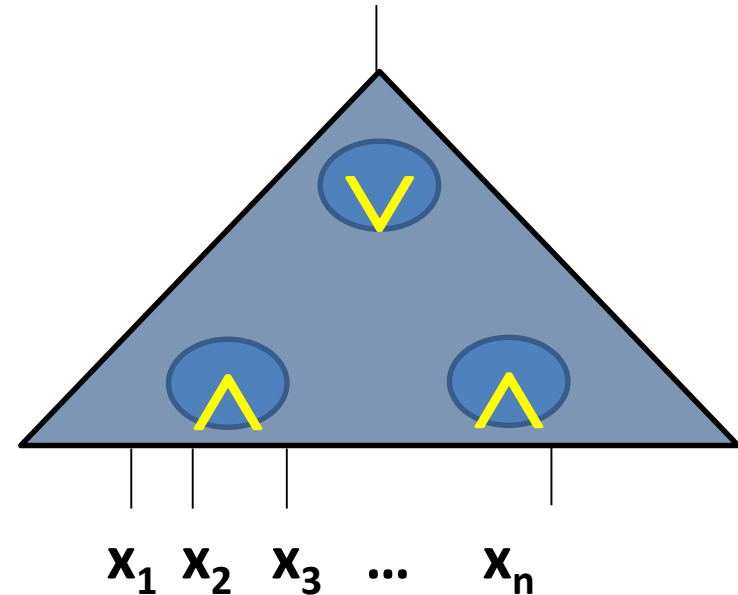
- V^0 : first-order reasoning
- VTC^0 : V^0 + “exists numones(y, X)= z ”
- VNC^1 : V^0 + “exists an evaluation of a Boolean formula”
 - Not known to prove $AB=I \rightarrow BA = I$
 - Uniform version of Frege/Sequent Calculus LK
- $V^1 \approx S_2^1 \dots$

Formalizations

- Approximate counting, randomized computation, PRGs (Jerabek), PCPs (Pich), Toda's theorem in higher complexity theories.
- Assuming existence of expanders, correctness of AKS sorting networks is provable in a (slightly non-uniform version of) VNC^1 (Jerabek)
- **Our result:** Theory VNC^1 proves existence of expanders of arbitrary size.
 - Thus, NC^1 reasoning is enough to prove correctness of AKS sorting networks.

Complexity in monotone

- Monotone functions:
 - $\forall x, y, x \subseteq y \Rightarrow f(x) \leq f(y)$
 - Majority, Threshold, Clique...
- Monotone circuits:
 - AND, OR gates.
 - $\text{Clique}_{k,n}$ requires monotone circuits of size $\geq 2^{\epsilon\sqrt{k}}$ for some ϵ .



Monotone proof complexity?

Monotone sequent calculus (MLK)

- Monotone version of LK [Buss-Pudlak'95]
- Sequents: $A_1, \dots, A_n \longrightarrow B_1, \dots, B_m$
 - all A_i, B_j are formulas over \wedge, \vee .
 - Axioms $A \rightarrow A, 0 \rightarrow S, S \rightarrow 1$.
 -
 - Rules for \vee, \wedge and cut
 - No rule for \neg

$$\frac{F \rightarrow G, A \quad F \rightarrow G, B}{F \rightarrow G, A \wedge B}$$

$$\frac{A, B, F \rightarrow G}{A \wedge B, F \rightarrow G}$$

$$\frac{F \rightarrow G, A \quad A, F \rightarrow G}{F \rightarrow G}$$

- Non-uniform version of VNC^1
- Polynomial-size proofs of PHP

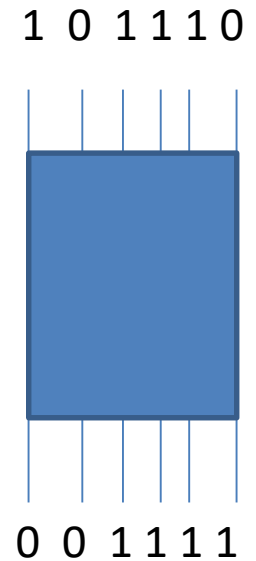
MLK polynomially simulates LK

- [Atserias-Galesi-Gavaldà'01, Atserias-Galesi-Pudlak'02]:
 - Simulate $\neg x$ using threshold formulas:
 - if k 1s in the input, and still k 1s with x_i replaced by 0, then $x_i = 0$
 - Slice functions idea.
 - Recursive definition of thresholds gives quasipolysize proofs.
 - Monotone NC^1 threshold functions?
 - AKS sorting networks

AKS sorting networks

- Sorting network:
 - n inputs, n outputs (Boolean)
 - Outputs input bits in sorted order

- [Ajtai-Komlos-Szemerédi'83]
 - Monotone log-depth sorting networks
 - Based on expanders



AKS sorting networks

- [Jerabek'11] Properties of AKS sorting networks are in (slightly non-uniform) VNC^1 .
So
 - if VNC^1 proves that expanders exist,
 - get polysize proofs for properties of thresholds
 - and polynomial simulation of LK by MLK.
- Here: VNC^1 proves that expanders exist.

Open problems

- Can existence of expanders be proven in VTC^0 ?
- Complexity of $USTCONN \in L$?
 - Our analysis needs both initial graphs to be expanders.
- Proof complexity of other results that now rely on algebra?

