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# GENERATING HARD BUT SOLVABLE SAT FORMULAS

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## OUTLINE

- ① The 3-SAT Problem
- ② Instance Generation with Hidden Satisfying Assignment
  - Random Clause Generation (Uniform)
  - Hidden Complementary Assignment
  - 3-SAT Spin-Glass Model
  - 3-XOR-SAT Spin-Glass Model
- ③ Conclusions

## THE 3-SAT PROBLEM

→ Given formula  $F$  over set of  $N$  boolean variables (in CNF)

$$\{x_i | i = 1, \dots, N\}$$

→  $F$  consists of conjunction of  $M$  logical clauses

$$F = C_1 \wedge C_2 \wedge \dots \wedge C_M, \quad C = \{C_\mu | \mu = 1, \dots, M\}, \quad \alpha = M/N$$

→ Each clause is disjunction of 3 *literals*

$$C_\mu = (l_\mu^1 \vee l_\mu^2 \vee l_\mu^3), \quad l_\mu^i = x_k, \bar{x}_k$$

→ Exists assignment  $x_i \mapsto \{\text{true}, \text{false}\}$  that satisfies  $F$ ? (NP-complete)

→ Complete (zChaff, Satz) vs. incomplete (WalkSAT, RRT, SP) solvers

→ Problem: Generate test instances for solvers:

a) hard   b) have known truth assignment   c) easy and fast to generate

→ Consider other applications: e.g. cryptography (one-way functions)

## INSTANCE GENERATION

→ First candidate:

Pick truth assignment  $A$  uniformly at random;  
**foreach** clause  $C_\mu$  **do**  
 pick indices  $i, j, k$  uniformly at random;  
**foreach**  $i, j, k$  **do**  
 flip coin: negated or not;  
**if**  $A$  evaluates  $C_\mu$  to false **then** discard  $C_\mu$ ;  
**else** accept;

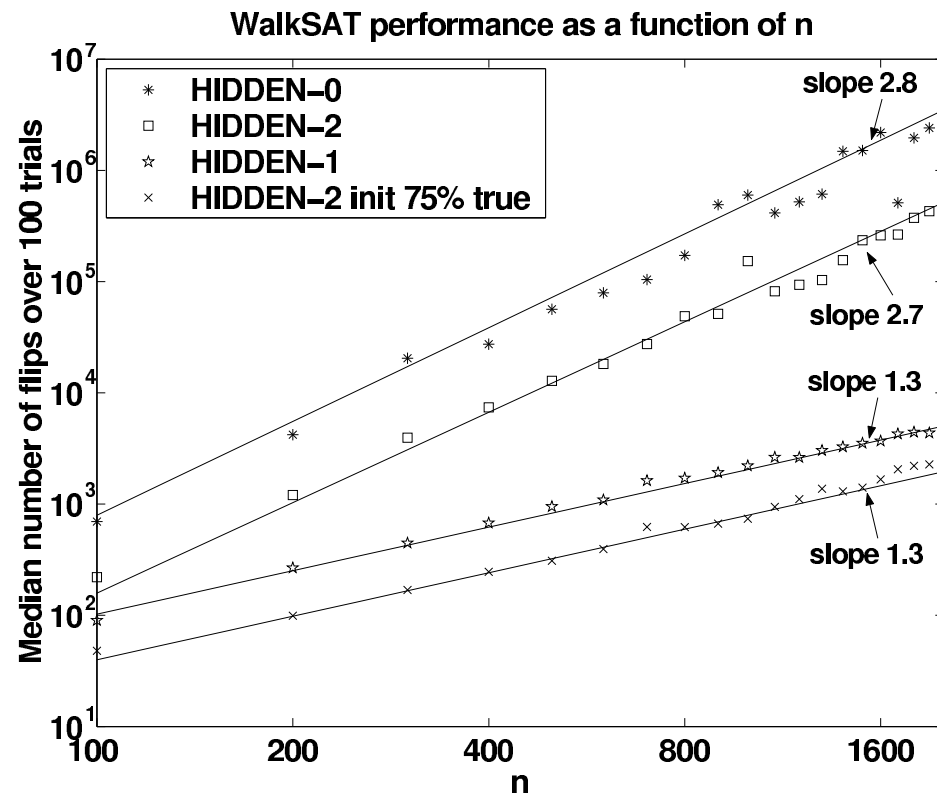
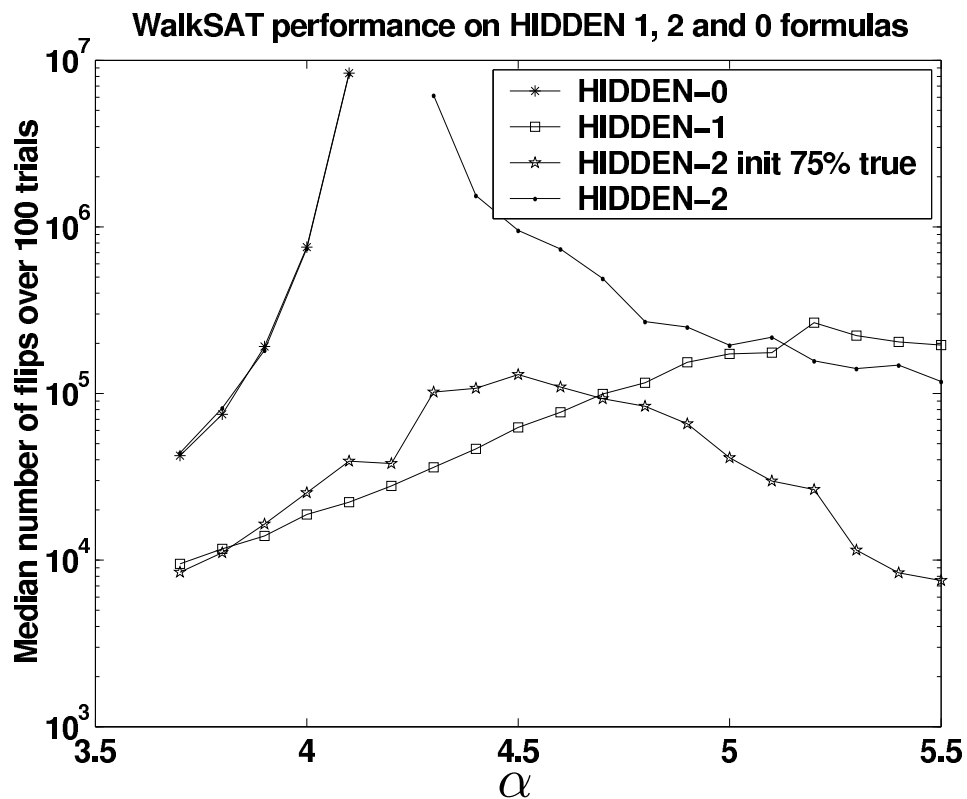
→ Potential problem: (hidden  $A = \{x_1 \leftarrow \text{true}, x_2 \leftarrow \text{true}, x_3 \leftarrow \text{true}\}$ )

$(x_1 \vee x_2 \vee x_3)$	$(x_1 \vee x_2 \vee \bar{x}_3)$	$(x_1 \vee \bar{x}_2 \vee x_3)$	$(x_1 \vee \bar{x}_2 \vee \bar{x}_3)$
$(\bar{x}_1 \vee x_2 \vee x_3)$	$(\bar{x}_1 \vee x_2 \vee \bar{x}_3)$	$(\bar{x}_1 \vee \bar{x}_2 \vee x_3)$	<del><math>(\bar{x}_1 \vee \bar{x}_2 \vee \bar{x}_3)</math></del>

→ Picking clauses uniformly at random introduces drift towards the correct assignment → formulas become “easier” (in some sense)

→ Idea: Hide  $A$  and complementary  $\bar{A}$  [Achlioptas, Jia, Moore, J. Artif. Intell. Res. (2005)]

# RESULTS FOR WALKSAT



Plot for  $\alpha = 4.25$

## SPIN-GLASS MODEL FOR 3-SAT

→ Map boolean  $x_i = 0, 1$  to Ising spins  $s_i = (-1)^{1-x_i} = -1, +1$ ; denote:

$$c_\mu^{(i)} = +1 \text{ if } x_i \text{ directly in } C_\mu, \quad c_\mu^{(i)} = -1 \text{ if } x_i \text{ negated in } C_\mu$$

→ Hamiltonian  $H$  defined to count unsatisfied clauses:

$$H(s) = C - \sum_{i=1}^N H_i s_i - \sum_{i<j} T_{ij} s_i s_j - \sum_{i<j<k} J_{ijk} s_i s_j s_k$$

→ One obtains (collecting constant and first-order terms):

$$H(s) = \sum_{\mu=1}^M \frac{1}{8} \prod_{i=1}^N (1 - c_\mu^{(i)} s_i), \quad C = \frac{M}{8} = \frac{\alpha}{8} N, \quad H_i = \frac{1}{8} \sum_{\mu=1}^M c_\mu^{(i)}$$

→ Collecting higher-order terms:

$$T_{ij} = -\frac{1}{8} \sum_{\mu} c_\mu^{(i)} c_\mu^{(j)}, \quad J_{ijk} = \frac{1}{8} \sum_{\mu} c_\mu^{(i)} c_\mu^{(j)} c_\mu^{(k)}$$

## SPIN-GLASS MODEL FOR 3-SAT CONT.

→ Idea: Generate clauses according to probability distribution, observe resulting spin-glass model (phase transition. . .) [Barthel et al., Phys. Rev. Lett. (2002)]

→ Clause probabilities  $p_i$  s.t.  $p_0 + 3p_1 + 3p_2 = 1$  (hiding  $x_i = 1 \forall i$ ):

$$p_0 : \quad (x_i \vee x_j \vee x_k)$$

$$p_1 : \quad (x_i \vee x_j \vee \bar{x}_k) \quad (x_i \vee \bar{x}_j \vee x_k) \quad (\bar{x}_i \vee x_j \vee x_k)$$

$$p_2 : \quad (x_i \vee \bar{x}_j \vee \bar{x}_k) \quad (\bar{x}_i \vee x_j \vee \bar{x}_k) \quad (\bar{x}_i \vee \bar{x}_j \vee x_k)$$

→ Averages resulting for spin-glass model:

$$\Pr(x_i \text{ or } \bar{x}_i \text{ appears in clause } \mu) = \frac{\binom{N-1}{2}}{\binom{N}{3}} = \frac{3}{N}$$

$$\overline{H_i} = \frac{1}{8} M E(c_\mu^{(i)}) = \frac{3\alpha}{8} \left( p_0 \cdot 1 + 3p_1 \cdot \frac{2-1}{3} + 3p_2 \cdot \frac{1-2}{3} \right) = \frac{3\alpha}{8} (p_0 + p_1 - p_2)$$

→ Similarly:  $\overline{T_{ij}} = \frac{3\alpha}{4N} (-p_0 + p_1 + p_2)$ ,  $\overline{J_{ijk}} = \frac{3\alpha}{4N^2} (p_0 - 3p_1 + 3p_2)$

## EFFECT OF VARIOUS CHOICES FOR $p_i$

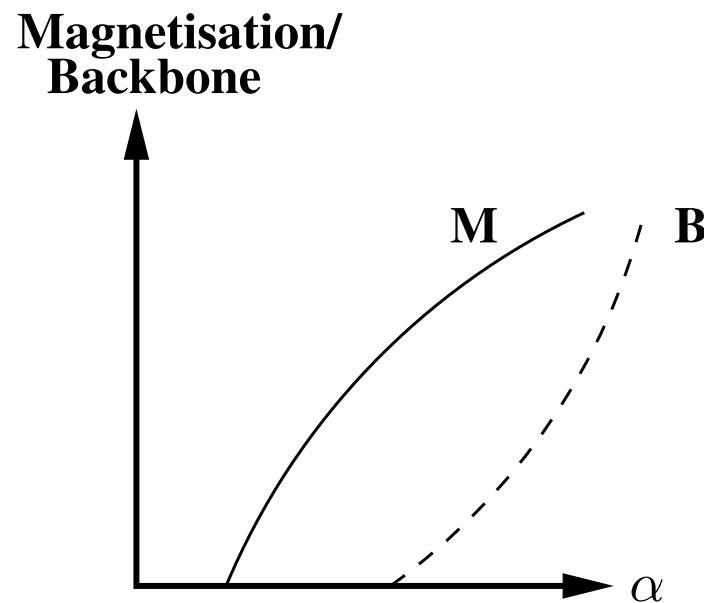
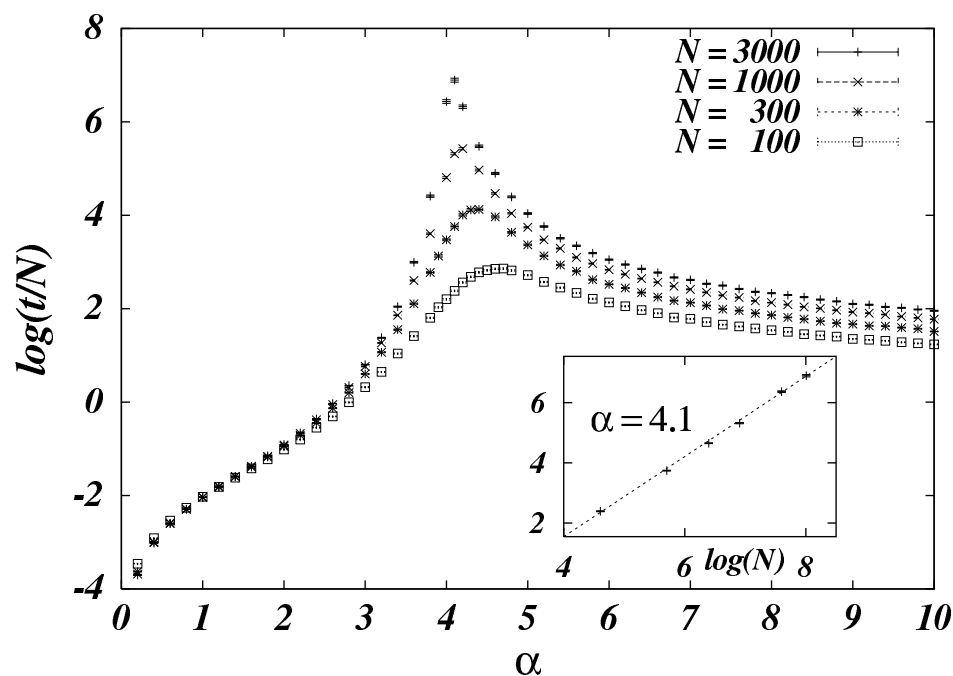
- ① Uniformly at random, rejecting clause  $(\bar{x}_i \vee \bar{x}_j \vee \bar{x}_k)$
- $x_i = 1 \forall i$  is satisfying
  - Case  $p_0 = p_1 = p_2 = 1/7 \Rightarrow \overline{H_i} = \frac{3\alpha}{56}$
  - Solvers (e.g. local solvers s.a. WalkSAT) guided by local field
  - Therefore: Set  $\overline{H_i} = \frac{3\alpha}{8}(p_0 + p_1 - p_2) = 0$
  - Resulting restrictions for probabilities  $p_i$ :

$$0 \leq p_0 \leq \frac{1}{4}, \quad p_1 = \frac{1 - 4p_0}{6}, \quad p_2 = \frac{1 + 2p_0}{6}$$



## EFFECT OF VARIOUS CHOICES FOR $p_i$ (CONT.)

- ② Uniformly at random, rejecting clauses  $(\bar{x}_i \vee \bar{x}_j \vee \bar{x}_k)$ ,  $(x_i \vee x_j \vee x_k)$ 
  - Both,  $x_i = 1 \forall i$  and  $x_i = 0 \forall i$ , are satisfying
  - Case  $p_0 = 0, p_1 = p_2 = 1/6 \Rightarrow \overline{J_{ijk}} = \frac{3\alpha}{4N^2} (p_0 - 3p_1 + 3p_2) = 0$
  - Seems more difficult, but WalkSAT shows avg. runtime  $O(N^c)$



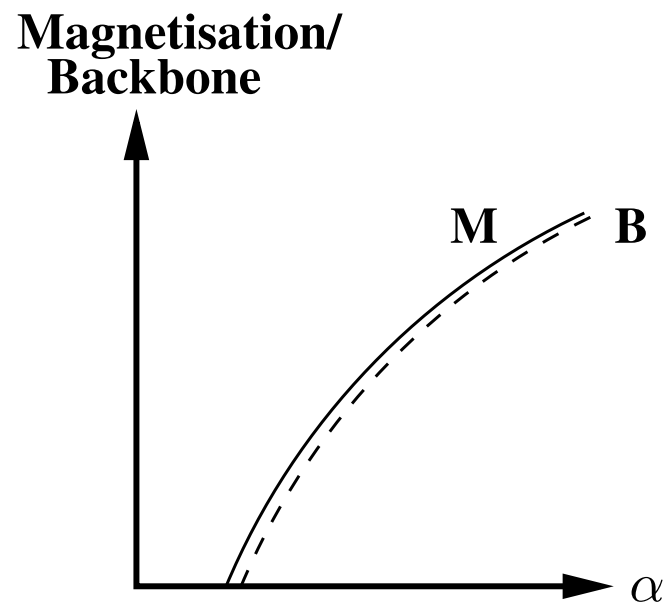
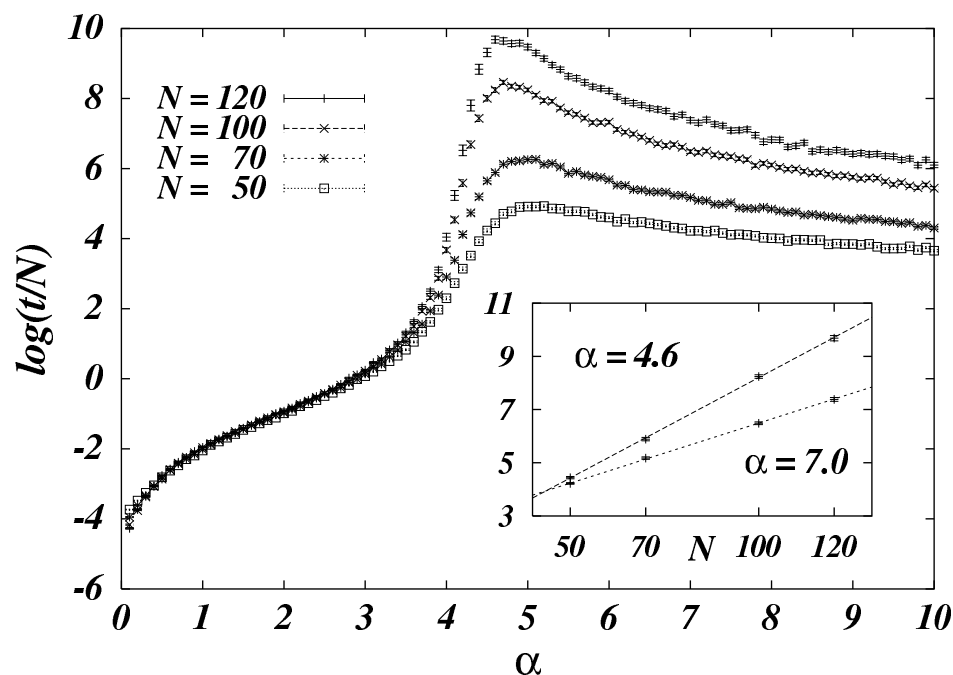
## EFFECT OF VARIOUS CHOICES FOR $p_i$ (CONT.)

### ③ Random satisfiable 3-XOR-SAT

→ Case  $p_0 = p_2 = 1/4, p_1 = 0$

→ Can be solved in  $O(N^c)$  time (Gauss), but difficult for solvers

→ WalkSAT shows exponential running time close to phase transition

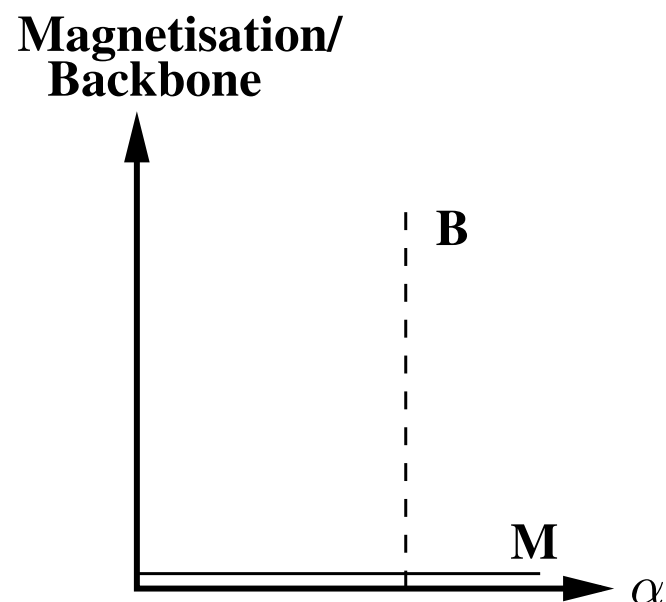


## EFFECT OF VARIOUS CHOICES FOR $p_i$ (CONT.)

④ Range  $0.077 \lesssim p_0 < 1/4$

→ Discontinuous appearance of backbone

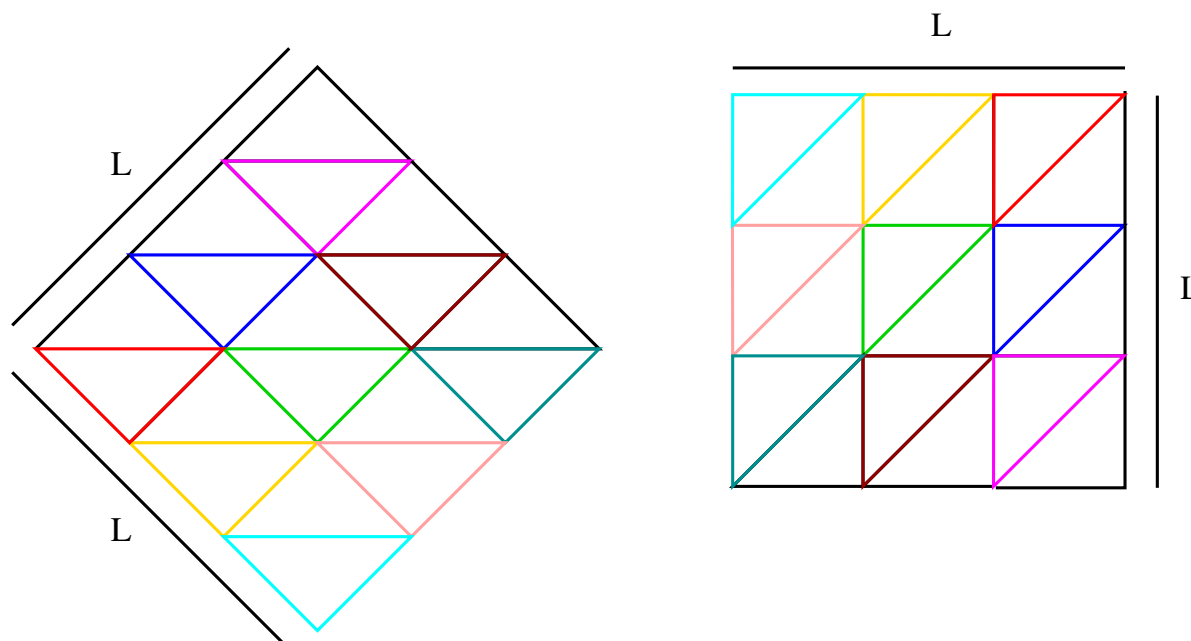
→ Leads to hard instances



## 3-XOR-SAT

- Triangular lattice spin-glass model [Jia, Moore, Selman, Th. Appl. o. Sat. Testing (2005)]
- Nearest neighbour interaction and short loops → glassy
- Lattice:  $L \times L$  rhombus with periodic boundary conditions
- Hamiltonian

$$H = \frac{1}{2} \sum_{i,j=0}^{L-1} s_{i,j} \cdot s_{i,j+1 \bmod L} \cdot s_{i+1 \bmod L,j}$$



## 3-XOR-SAT CONT.

→ Rewriting in terms of booleans  $x_{i,j} = \frac{1}{2}(s_{i,j} + 1)$  then (up to a constant)

$$H = \sum_{i,j=0}^{L-1} \left( (x_{i,j} + x_{i,j+1 \bmod L} + x_{i+1 \bmod L,j}) \bmod 2 \right)$$

→ Corresponds to  $L^2$  3-XOR-SAT clauses of the form

$$\overline{x_{i,j} \oplus x_{i,j+1 \bmod L} \oplus x_{i+1 \bmod L,j}}$$

$$\equiv (\bar{x}_{i,j} \vee x_{i,j+1 \bmod L} \vee x_{i+1 \bmod L,j}) \wedge (x_{i,j} \vee \bar{x}_{i,j+1 \bmod L} \vee x_{i+1 \bmod L,j}) \wedge (x_{i,j} \vee x_{i,j+1 \bmod L} \vee \bar{x}_{i+1 \bmod L,j}) \wedge (\bar{x}_{i,j} \vee \bar{x}_{i,j+1 \bmod L} \vee \bar{x}_{i+1 \bmod L,j})$$

→ 3-SAT formula,  $L^2$  variables,  $4L^2$  clauses,  $H$  counts unsatisfied clauses

→ Satisfying:  $x_{i,j} = 0 \forall i, j$  (unique if  $L = 2^k$  [Newman, Moore, Phys. Rev. Lett., (1999)])

→ Hidden assignment is arbitrary: define  $y_{i,j} = x_{i,j} \oplus a_{i,j}$  (can be found e.g. by Gauss elimination mod 2)

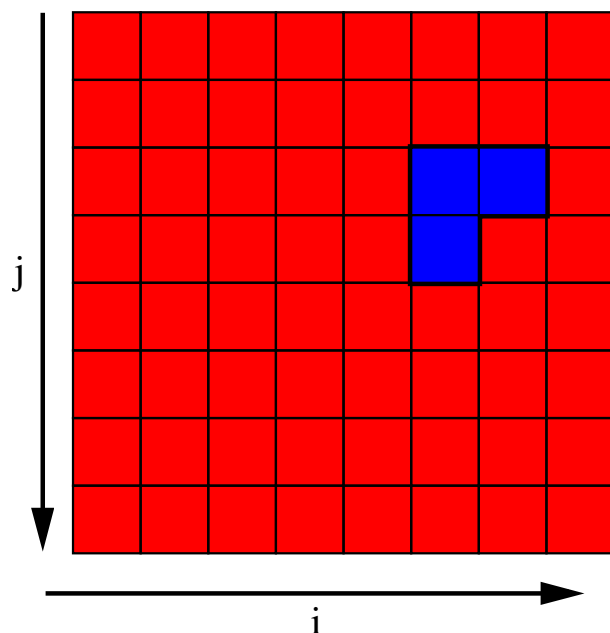
## HARDNESS OF FORMULAS FOR SAT SOLVERS

→ Satisfying assignment implies recurrence equation

$$x_{i,j} \oplus x_{i,j+1 \bmod L} \oplus x_{i+1 \bmod L,j} \equiv 0 \Rightarrow x_{i,j+1 \bmod L} = x_{i,j} \oplus x_{i+1 \bmod L,j}$$

⇒ Truth values are given by Pascal's triangle mod 2!

→ ■: undetermined, ■: fixed to 1 (defect), ■: 1, : 0



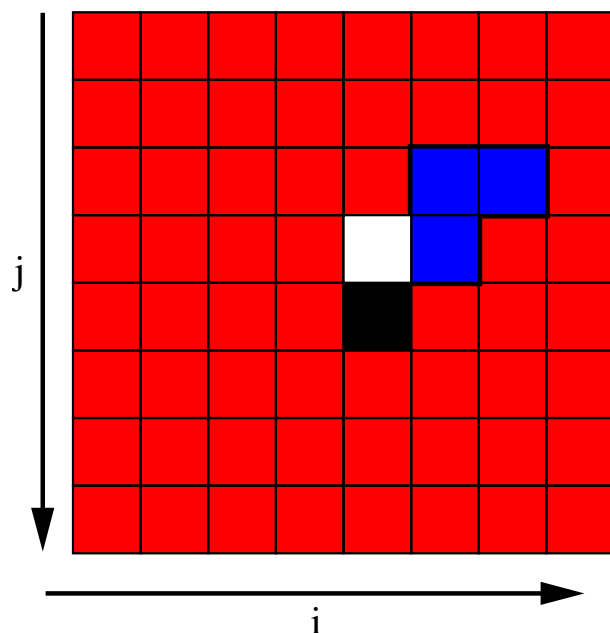
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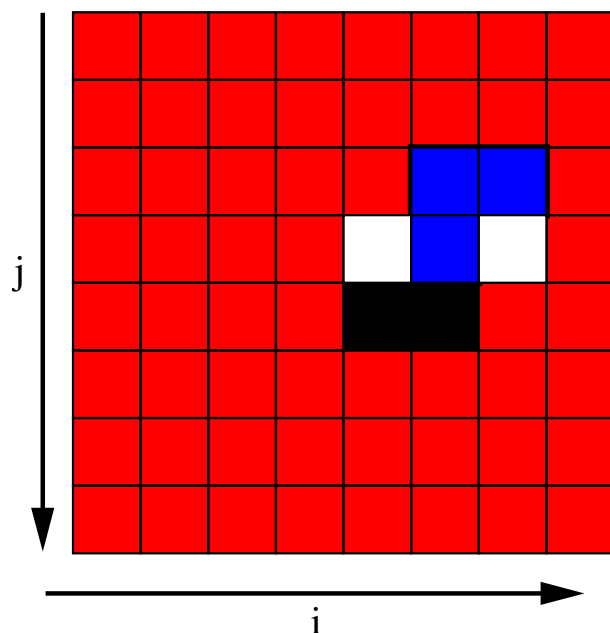
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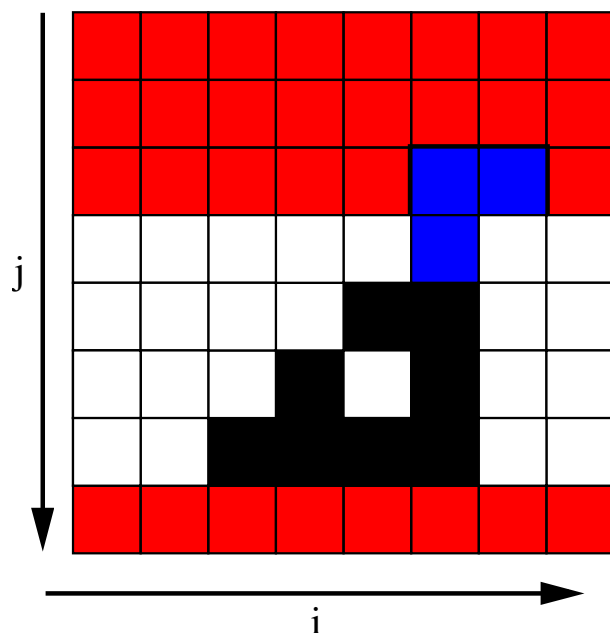
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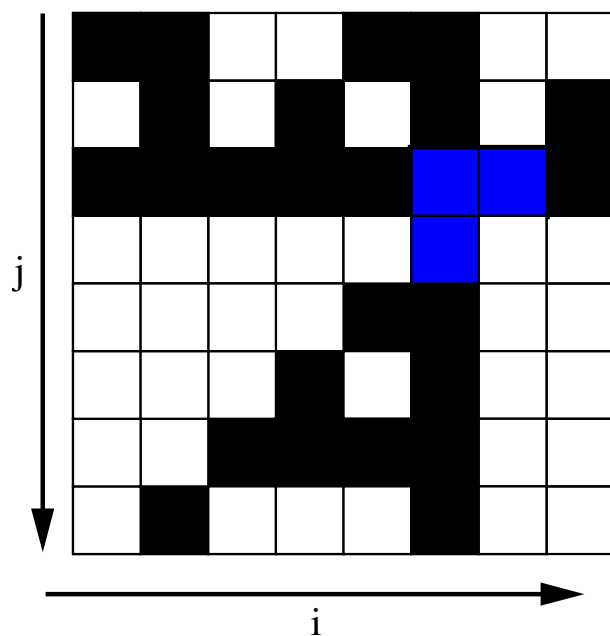
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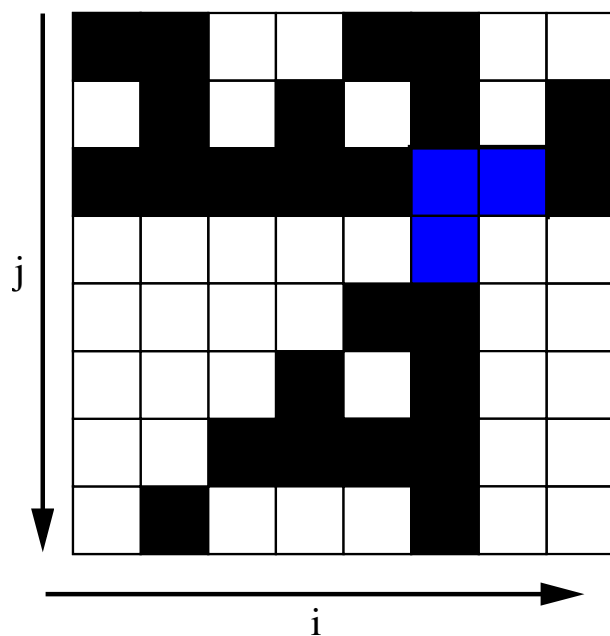
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Analytical calculations show

→ Hamming distance to sat. assignment:  $L^{\log_2 3}$  (#ones)

→ Energy barrier towards sat. assignment:  $O(\log_2 L)$

→ #local minima:  $O(\kappa^{L^2})$ ,  $\kappa \approx 1.395$  (*hard hexagon constant*)

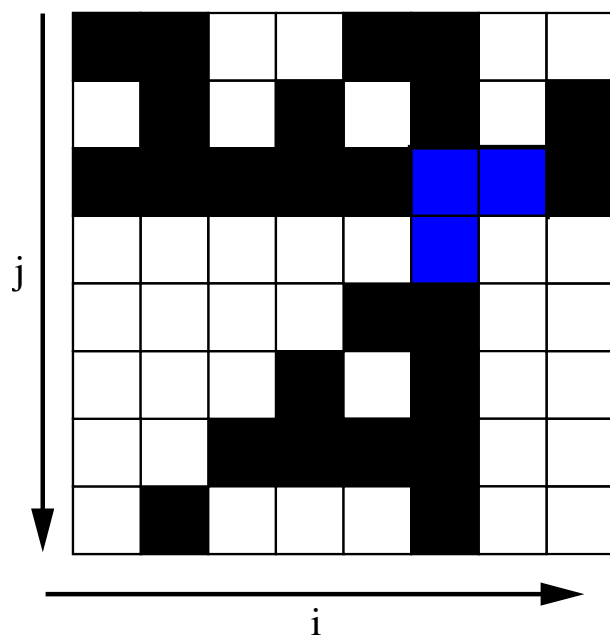
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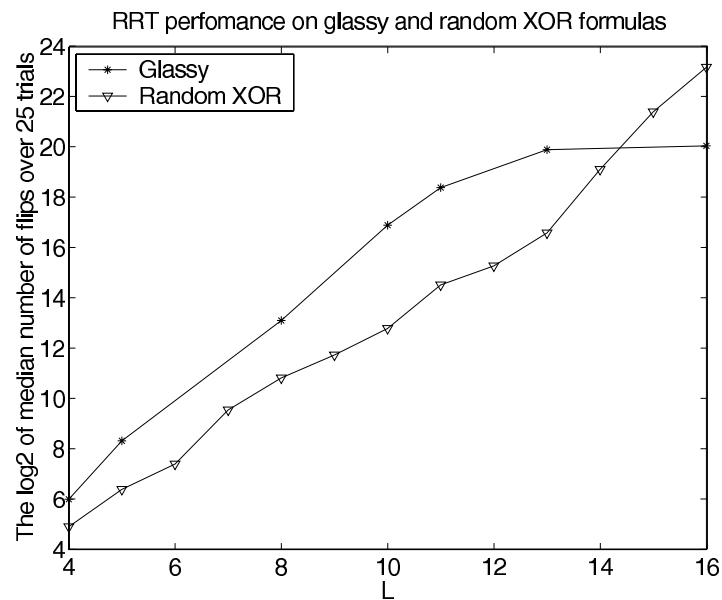
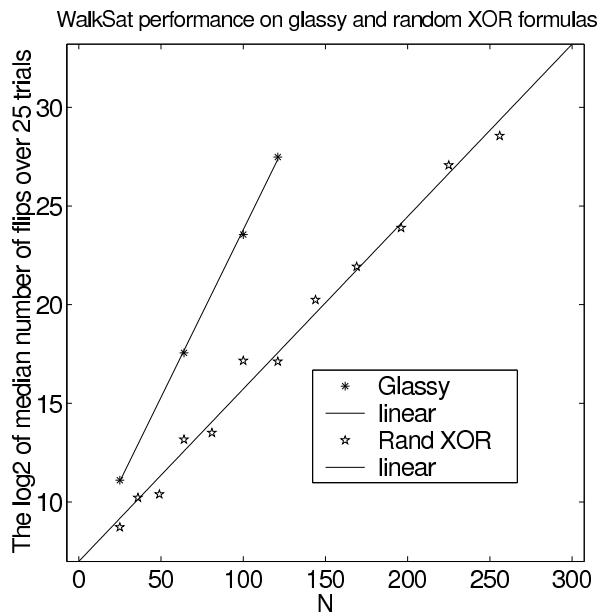
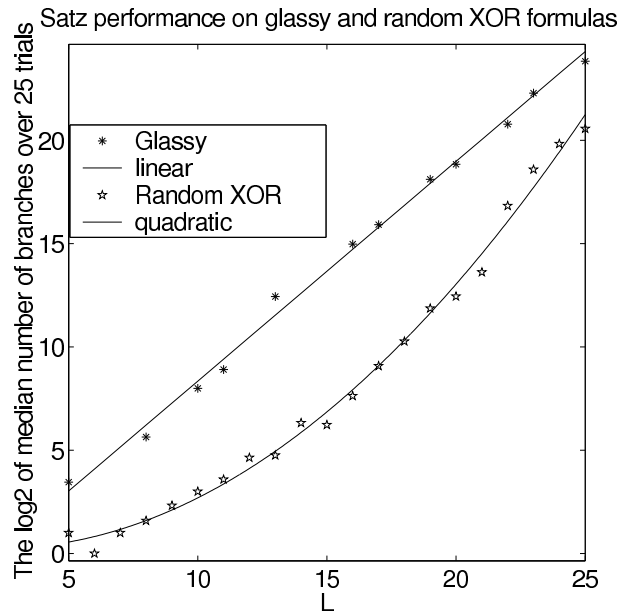
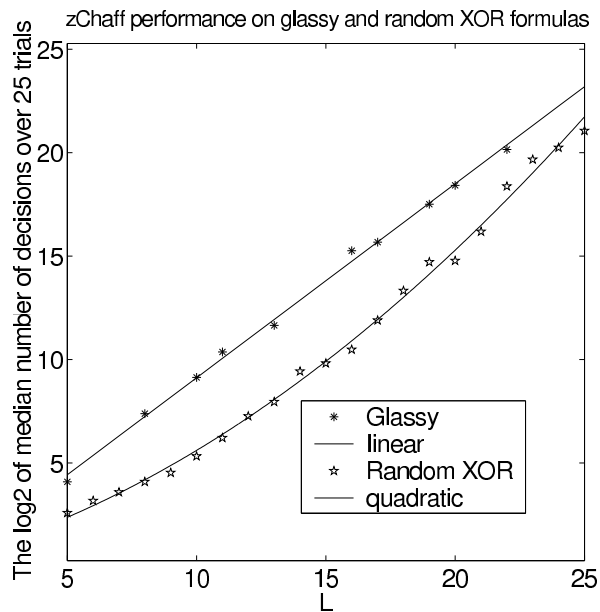
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⇒ hard for local algorithms!

# EXPERIMENTAL RESULTS [1]



## CONCLUSIONS

- Generation of 3-SAT instances important for evaluation of solvers
- Not all “random” generation schemes result in hard formulas
- Parameterised generation of instances can reveal interesting aspects of problem structure
- Instances can be difficult for local solvers even though efficiently solvable by other means (XOR-SAT)
- Wide range of other generation methods, e.g. regular XOR-SAT based on 3-regular constraint graphs [Haanpää, Jarvisalo, Kaski, and Niemelä, Journal on Satisfiability, Boolean

Modeling and Computation (2006)]

## COMMENTS, QUESTIONS?

# Thank you for your attention.

## References

- [1] Jia, H., Moore, C., Selman, B.: From spin glasses to hard satisfiable formulas. In: Theory and Applications of Satisfiability Testing. Volume 3542/2005., Springer Berlin / Heidelberg (2005) 199–210
- [2] Barthel, W., Hartmann, A.K., Leone, M., Ricci-Tersenghi, F., Weigt, M., Zecchina, R.: Hiding solutions in random satisfiability problems: A statistical mechanics approach. Phys. Rev. Lett. **88**(18) (2002) 188701
- [3] Newman, M., Moore, C.: Glassy dynamics in an exactly solvable spin model. Phys. Rev. Lett. **60** (1999) 5068–5072
- [4] Achlioptas, D., Jia, H., Moore, C.: Hiding satisfying assignments: Two are better than one. J. Artif. Intell. Res. (JAIR) **24** (2005) 623–639