



School of Computer Science, University of Adelaide



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# A Parameterized Runtime Analysis of Evolutionary Algorithms for MAX-2-SAT

Jareth Day, Andrew M. Sutton and Frank Neumann

School of Computer Science  
University of Adelaide, Australia

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## Introduction

- Introduce MAX-2-SAT and FPT
- Parameterized complexity analysis on MAX-2-SAT
  - Identify fitness landscape of MAX-2-SAT
  - Produce parameterized algorithms for MAX-(2,3)-SAT.





## MAX-2-SAT

- Maximum 2-Satisfiability Problem
- $\mathcal{C} = \{(l_{1,1} \vee l_{1,2}), (l_{2,1} \vee l_{2,2}), \dots, (l_{m,1} \vee l_{m,2})\}$
- eg  $\mathcal{C} = \{(\neg v_1 \vee v_2), (v_1 \vee v_3), \dots, (v_6 \vee v_n)\}$
- $m$  clauses,  $\mathcal{C}_i = \{l_1, \dots, l_m\}$
- $n$  Boolean variables  $v_i = \text{true/false}$





## MAX-2-SAT

- $\mathcal{C} = \{(\neg v_1 \vee v_2), (v_1 \vee v_3), \dots, (v_6 \vee v_n)\}$
- Given  $x \in \{0, 1\}^n$ ,  
 $x_i = 1$  corresponds to  $v_i = \text{true}$ ,  
 $x_i = 0$  corresponds to  $v_i = \text{false}$
- We want to maximize  $f: \{0, 1\}^n \rightarrow \{0\} \cup [m]$
- $f(x)$  = number of clauses satisfied by  $x$





## Parameterized Complexity

- MAX-2-SAT is NP-Hard.
- Standard algorithms:  $\exp(|x|)$
- Parameterized complexity: parameterization  $K$
- XP algorithms:  $|x|^{g(K(x))}$   
eg  $n^k$
- Fixed-parameter tractable (FPT) algorithms:  $g(K(x)) \cdot |x|^{O(1)}$   
eg  $n \cdot 2^k$





## MAX-2-SAT Algorithm

- We analyze the runtime of the (1+1) EA:

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**Algorithm 1:** The (1+1) EA.

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```
1 Choose  $x$  uniformly at random from  $\{0, 1\}^n$ ;  
2 repeat forever  
3    $x' \leftarrow \text{mutate}(x)$ ;  
4   if  $f(x') \geq f(x)$  then  $x \leftarrow x'$ 
```

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- where `mutate()` negates some elements of  $x$





## Uniform-Complement Mutation

- Traditional uniform mutation creates offspring by flipping each bit of  $x$  with probability  $1/n$
- Uniform-complement may, with uniform probability, produce the complement of  $x$  with probability  $\Theta(1)$ .
- Complement under uniform mutation: probability  $O(n^{-n})$





## Uniform-Complement Mutation: Fitness Landscape

- MAX-2-SAT fitness function and uniform-complement operator corresponds to an elementary landscape.
- Can reach solutions of certain quality in polynomial time by making local improvements.







## Uniform-Complement Mutation: Fitness Landscape

- Let  $N(x)$  be union of the Hamming neighbors of  $x$  and the complement of  $x$
- If the  $i$ -th clause is not satisfied by  $x$ , it is satisfied for three neighbors  $y \in N(x)$ :
  - The two Hamming neighbors of  $x$  that have the variables in the  $i$ -th clause negated, and
  - The complement of  $x$ .
- If the  $i$ -th clause is satisfied by  $x$ , at least one of its literals evaluates to true under  $x$ .
  - If only one true, clause is satisfied for all  $y \in N(x)$  except for the negation of variable involved in the true literal.
  - If both true, clause is satisfied for all  $y \in N(x)$  except for the complement.





## Uniform-Complement Mutation: Fitness Landscape

- If clause  $i$  unsatisfied by  $x$ , clause satisfied by three neighbors.
- If clause  $i$  satisfied by  $x$ , clause satisfied by  $|N(x)| - 1$  neighbors.
- Let  $c_i : \{0, 1\}^n \rightarrow \{0, 1\}$  if clause  $i$  is satisfied by  $x$ .

$$\sum_{y \in N(x)} c_i(y) = 3(1 - c_i(x)) + (|N(x)| - 1)c_i(x) = 3 + (n - 3)c_i(x)$$

- Since  $f(x)$  is the sum of the clauses satisfied in  $x$

$$\sum_{y \in N(x)} f(y) = \sum_{i=1}^m (3 + (n - 3)c_i(x)) = 3m + (n - 3)f(x)$$





## Uniform-Complement Mutation: Fitness Landscape

- Until no further improvements can be made, there are two cases in which an improvement is generated.
  - Complement is improving state.  
Probability  $\frac{1}{2}$  to choose.
  - Hamming neighbor is improving state.  
Probability  $\frac{1}{2}(n^{-1}(1 - n^{-1})^{n-1}) \geq (2en)^{-1} = \Omega(n^{-1})$
- Number of improvements bounded by number of clauses.
- Reaches state with no improvements in expected time bounded by  $O(mn)$





## Uniform-Complement Mutation: Fitness Landscape

- Reached solution  $x'$  s.t.  $f(x')$  has best fitness in neighborhood
- Current state  $x = x'$

$$\frac{1}{|N(x')|} \sum_{y \in N(x')} f(y) \leq f(x')$$

$$\frac{1}{|N(x')|} \sum_{y \in N(x')} (3m + (n-3)f(x')) \leq f(x')$$

$$\frac{3m}{(n+1)} + \frac{(n-3)}{(n+1)} f(x') \leq f(x')$$

$$f(x') \geq \frac{3}{4} m$$





## MAX-(2,3)-SAT

- Restricted problem: MAX-(2,3)-SAT
- Each variable may only appear in at most 3 clauses
- eg  $\mathcal{C} = \{(\neg v_1 \vee v_2), (v_1 \vee \neg v_3), (\neg v_1 \vee \neg v_5), \dots\}$
- Still NP-hard





## MAX-(2,3)-SAT

- Graph:  $G(V, E)$

$$|V| = n$$

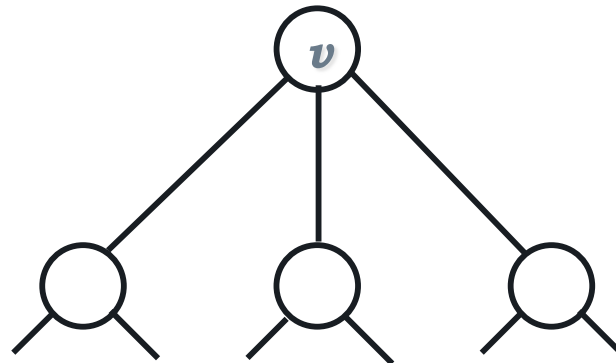
$$E = \{\{u, v\} \subset V \mid u \text{ and } v \text{ appear together in a clause}\}$$

- As long as there are two variables in a clause, there's an edge.
- Diameter of  $G$ : maximum shortest-path distance in any of the connected components
- Parameter: diameter of  $G$  is bounded by  $k$





## MAX-(2,3)-SAT



- $C(v)$  is a connected component containing node  $v$
- With diameter bounded by  $k$ , the number of nodes in  $C(v)$  is bound by:

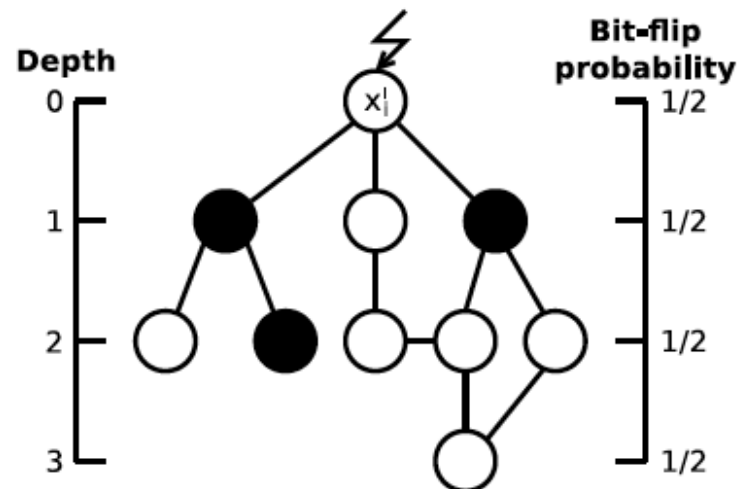
$$1 + \sum_{i=0}^{k-1} 3 \cdot 2^i = 3 \cdot 2^k - 2 \geq |C(v)|$$





## Basic FPT Algorithm

- Select  $v_i$  uniformly at random.
- Flip all bits in  $x$  associated with Boolean variables in  $C(v_i)$  with probability  $1/2$

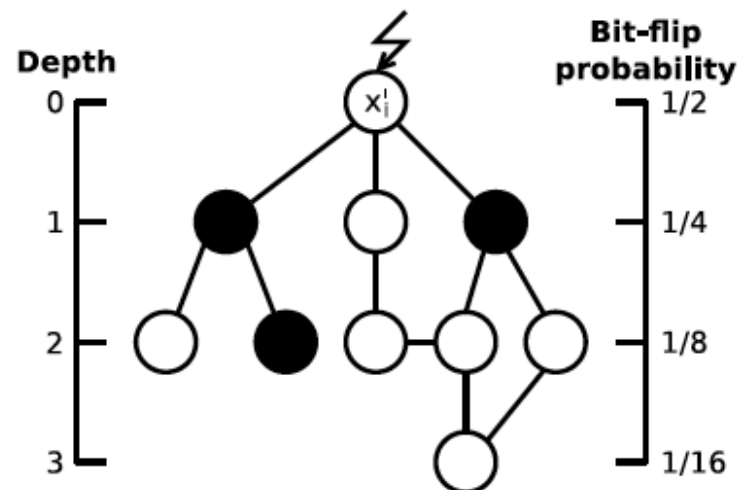






## Modified FPT Algorithm

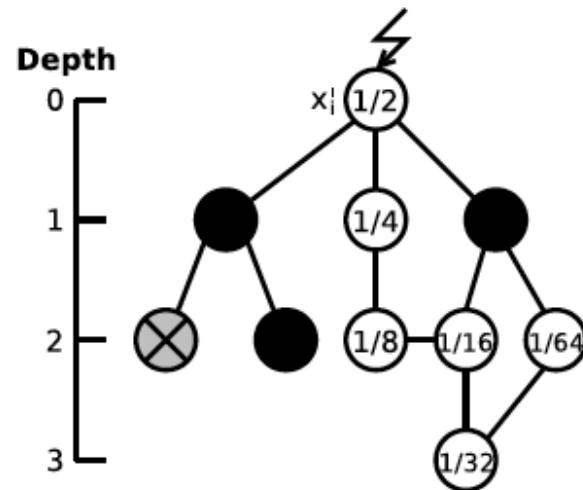
- Select  $v_i$  uniformly at random.
- Flip all bits in  $x$  associated with Boolean variables in  $C(v_i)$  with decreasing probability the further the distance from  $v_i$





## Propagation FPT Algorithm

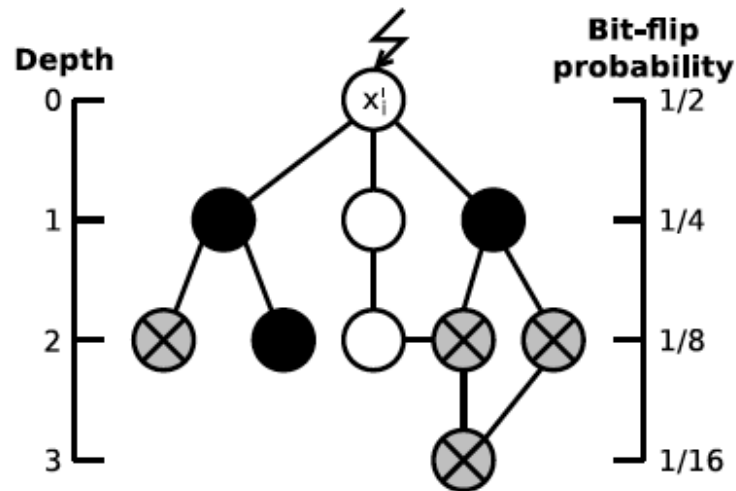
- Select  $v_i$  uniformly at random.
- Flip all bits in  $x$  associated with Boolean variables in  $C(v_i)$  with decreasing probability the further the distance from  $v_i$ . Only propagates to next bit if current bit flipped.





## Modified Propagation FPT Algorithm

- Select  $v_i$  uniformly at random.
- Flip all bits in  $x$  associated with Boolean variables in  $C(v_i)$  with decreasing probability the further the distance from  $v_i$ . Only propagates to next bit if current bit flipped. Only visits bits at increasing distance from  $v_i$ .





## Expected Time to Completion

- Number of nodes in connected component bound by  $g(k) = 3 \cdot 2^k - 2$

XP (in paper)	$O(n^{g(k)})$
Simple FPT	$O(n \log n \cdot 2^{g(k)})$
Modified FPT	$O(n \log n \cdot 2^{k \cdot g(k)})$
Propagation FPT	$O(n \log n \cdot 2^{g(k)^2})$
Modified Propagation FPT	$O(n \log n \cdot 2^{k \cdot g(k)})$





## Conclusion

- We analyzed MAX-2-SAT and MAX-(2,3)-SAT and determined:
  - Elementary landscape of MAX-2-SAT
  - XP evolutionary algorithm for MAX-(2,3)-SAT
  - FPT evolutionary algorithms for MAX-(2,3)-SAT

