

## Coping with NP-HARDNESS

- Strategies to deal with hard problems
- Algorithms for Special Cases
- Fixed Parameter Tractability
- Intelligent Exhaustive Search
  - Backtracking
  - Branch and Bound
- Dynamic Programming based pseudo polynomial algorithm TSP

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## Approaches to tackle hard problems

- 1 Special Cases:** Relevant structure on which the problem is easy
  - Exact results in poly-time only for special cases or a range of parameters
- 2 Intelligent Exhaustive Search:** Exponential time in worst case
  - The base and/or exponent are usually smaller
  - could be efficient on typical more realistic instances
  - Backtracking, Branch-and-Bound
- 3 Nearly exact solutions:** Output is 'close' to exact (optimal) solution
  - **Approximation Algorithms:** Solutions of guaranteed quality in poly-time
  - **Heuristic Algorithms:** Solutions hopefully good in poly-time
- 4 Randomized Algorithms:** Use coin flips for making decisions
  - Typically used for approximation, also used for easy problems

# Intelligent Exhaustive Search

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- Specific structure in instances is helpful sometimes
  - e.g.  $\text{IND-SET}(G, k)$  for trees is easy
  - 2-SAT is easy
- Sometime even a well-characterized special structure does not help
  - $\text{IND-SET}(G, k)$  is NP-HARD even for planar graphs
  - 3-SAT is NP-HARD
- In many cases, we cannot neatly characterize the particular cases
- Can still avoid exp-time exhaustively searching with clever methods
- These algorithms are still exp time in the worst case
  - With the right ideas they are efficient on typical (likely) instances

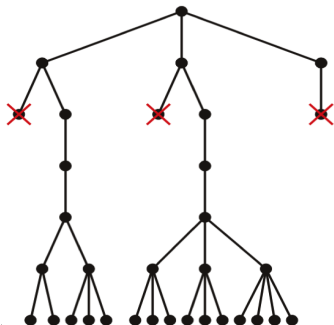
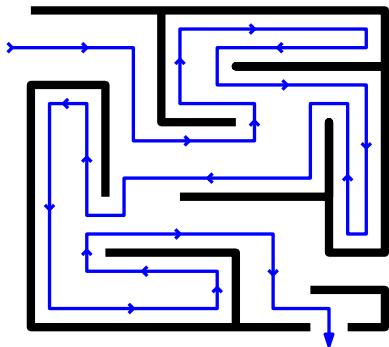
# Backtracking

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- Often solution to a problem can be made with a series of choices
- Each choice represents a partial solution
- These partial solutions form a tree (or DAG)
- Backtracking refers to a brute force solution where only feasible partial solutions are considered
- Feasibility and in-feasibility of partial solutions are determined given the specific problem in hand
- The idea in backtracking: **many partial solutions can be rejected quickly without completing it**

# Backtracking

Finding path in a maze - backtrack when you reach a dead-end



- Given a CNF formula  $f$  on  $n$  variables and  $m$  clauses
- **The brute force algorithm**
  - Check all  $2^n$  possible assignments to the  $n$  variables
  - Determine in  $O(m + n)$  whether an assignment is satisfying
  - Running time is  $O(2^n(n + m))$
- Visualize it as a complete full binary tree
  - Root of the tree correspond to variable  $x_1$
  - Left and right branches of root correspond to values of 1 and 0 for  $x_1$
  - Left and right subtrees are all possibilities for variables  $x_2, \dots, x_n$

- Do not consider all  $2^n$  branches of the binary tree (solution space)
- Carefully track each branch
- Stop when “get” a dead branch (cannot be extended to a solution)
  - $f = (\dots) \wedge \dots \wedge (x_6) \wedge \dots (\dots)$
  - Reject all solutions  $(x_1, \dots, x_n) \in \{0, 1\}^n$  with  $x_6 = 0$
  - Saves a lot- out of the  $2^n$  sized search space, we eliminated  $2^{n-1}$
- A more elaborate example follows

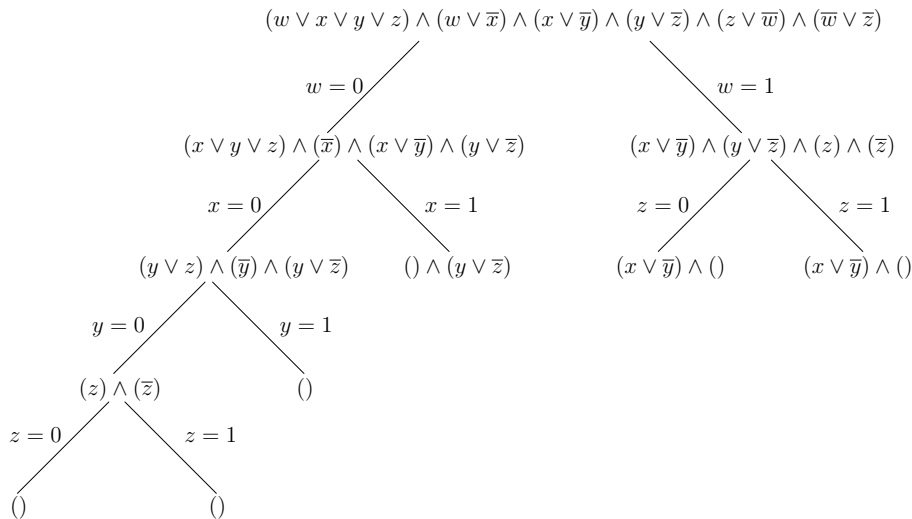
## Intelligent Exhaustive Search for SAT

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- Do not consider all  $2^n$  branches of the binary tree (solution space)
- Carefully track each branch
- Stop when “get” a dead branch (cannot be extended to a solution)
- When a literal in a clause is 1, the clause is satisfied we remove it
- When a literal in a clause is 0, the clause depends on other literals in it we remove the variable from it
- A partial assignment cannot satisfy the formula if there is an empty clause (no literal is 1)



# Intelligent Exhaustive Search for SAT



## General Backtracking Procedure

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A backtracking algorithm requires a test that looks at a subproblem and quickly declares one of three outcomes:

- **FAILURE**: the subproblem has no solution
- **SUCCESS**: a solution to the subproblem is found
- **UNCERTAINTY**: not yet clear if it is either - need to explore further

## Backtracking algorithm for problem P

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**Algorithm** Backtracking procedure for Problem P, Instance  $I_0$

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$\mathcal{S} \leftarrow \{I_0\}$

**while**  $\mathcal{S} \neq \emptyset$  **do**

    Choose a subproblem instance  $I \in \mathcal{S}$

$\mathcal{S} \leftarrow \mathcal{S} \setminus \{I\}$

    EXPAND  $I$  into  $\{I_1, I_2, \dots, I_k\}$

**for each**  $I_j$  **do**

**if**  $\text{TEST}(P_j) = \text{SUCCESS}$  **then**

**return** the current solution

**else if**  $\text{TEST}(P_j) = \text{FAILURE}$  **then**

**return NF**

**else**

$\mathcal{S} \leftarrow \mathcal{S} \cup \{I_j\}$

**return NF**

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## Backtracking for 3-SAT

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- Exhaustive search takes  $O(2^n \cdot (n + m))$  for a 3-CNF formula  $f$  on  $n$  variables and  $m$  clauses
- The previous approach was more variable centric
- Consider a more clause centric approach
- View a 3-CNF formula  $f$  as  $(l_1 \vee l_2 \vee l_3) \wedge (f')$  (unless  $f$  is empty)
- $f'$  too is a (possibly empty) 3-CNF formula
- By the distributive law we get

$$f = (l_1 \vee l_2 \vee l_3) \wedge (f')$$

$$\implies f = (l_1 \wedge f') \vee (l_2 \wedge f') \vee (l_3 \wedge f')$$

## Backtracking for 3-SAT

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$$f = (l_1 \vee l_2 \vee l_3) \wedge (f') \implies f = (l_1 \wedge f') \vee (l_2 \wedge f') \vee (l_3 \wedge f')$$

- $f[x = \mathbf{true}]$  ( $f$  with the value of  $x$  plugged in as **true**)

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### Algorithm Backtracking for 3-SAT

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**function** CHECK-SAT( $f$ )

**if**  $f$  is empty **then**

**return true**

**else**

    Let  $f = (l_1 \vee l_2 \vee l_3) \wedge (f')$

**if** CHECK-SAT( $f'[l_1 = \mathbf{true}]$ ) **then**

      ▷ implies  $l_1 \wedge f' = \mathbf{true}$

**return true**

**if** CHECK-SAT( $f'[l_2 = \mathbf{true}]$ ) **then**

**return true**

**if** CHECK-SAT( $f'[l_3 = \mathbf{true}]$ ) **then**

**return true**

**return false**

## Backtracking for 3-SAT

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$T(n)$ : runtime of this algorithm for a  $f$  on  $n$  variables with  $m$  clauses

$$T(n) = \begin{cases} 3T(n-1) + O(\text{poly}(n, m)) & \text{if } n \geq 1 \\ 1 & \text{otherwise} \end{cases}$$

$$T(n) = O(3^n \cdot \text{poly}(n, m))$$

▷ Simple recursion tree expansion

- Even worse that the variable centric brute-force search
- Observe the overlap in the subproblems - unnecessary repetitions

- Need to make these subproblems mutually exclusive
- Every satisfying assignment this algorithm finds (since it satisfies the clause  $(l_1 \vee l_2 \vee l_3)$ ) must be exactly one of the following types
  - $l_1 = \mathbf{true}$
  - $l_1 = \mathbf{false} \wedge l_2 = \mathbf{true}$
  - $l_1 = \mathbf{false} \wedge l_2 = \mathbf{false} \wedge l_3 = \mathbf{true}$
- We can pinpoint any of these three types of satisfying assignments to three literals in exactly one of the recursive calls

- Here is the clause centric algorithm based on this idea

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**Algorithm** Backtracking for 3-SAT

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```
function CHECK-SAT( $f$ )  
  if  $f$  is empty then  
    return true  
  else  
    Let  $f = (\ell_1 \vee \ell_2 \vee \ell_3) \wedge (f')$   
    if CHECK-SAT( $f'[\ell_1 = \text{true}]$ ) then  
      return true  
    if CHECK-SAT( $f'[\ell_1 = \text{false} \wedge \ell_2 = \text{true}]$ ) then  
      return true  
    if CHECK-SAT( $f'[\ell_1 = \text{false} \wedge \ell_2 = \text{false} \wedge \ell_3 = \text{true}]$ ) then  
      return true  
    return false
```

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Fixing values of  $k$  literals reduced number of variables in  $f$  by  $k$

$$T(n) = \begin{cases} T(n-1) + T(n-2) + T(n-3) + O(\text{poly}(n, m)) & n \geq 1 \\ 1 & \text{else} \end{cases}$$

Closed form of this recurrence is  $T(n) = O(1.84^n)$

- This is substantially faster than the  $O(2^n)$  algorithm
- Even for  $n \sim 100$  this is more than 4180 times faster