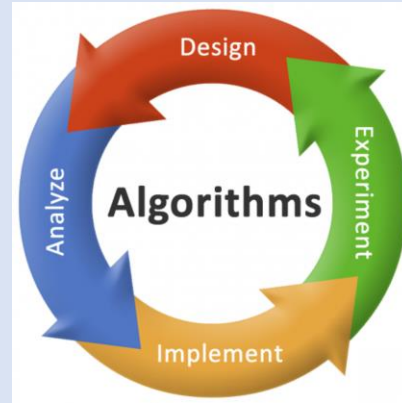


# Backtracking

COP 3503  
Fall 2021

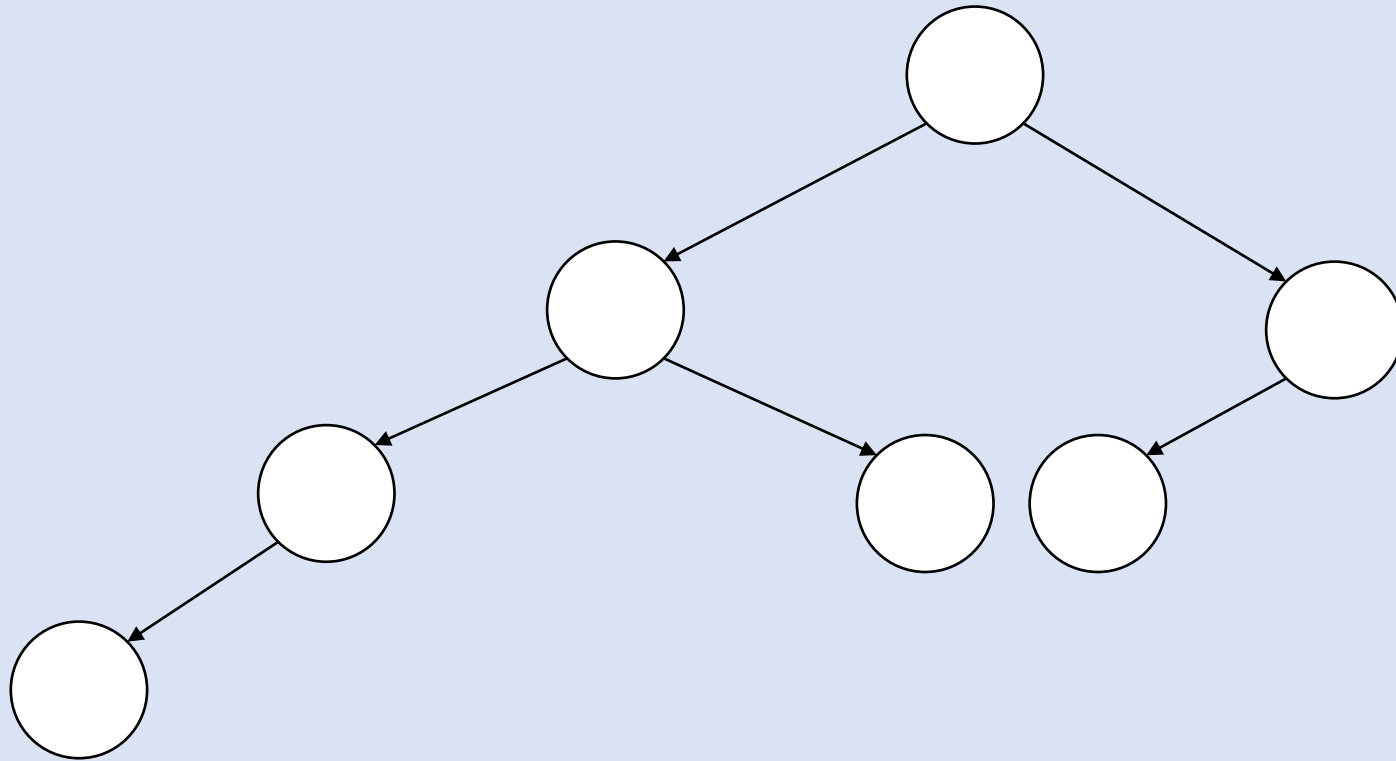
Department of Computer Science  
University of Central Florida  
Dr. Steinberg



# Introduction

- We just observed two common approaches to designing algorithms
  - Brute Force (BF)
  - Divide and Conquer (DC)
- BF and DC both yield correct results to problems. Both have their pros and cons that were discussed in previous lectures.
- There can be more than one correct result, however one may be better than the other.
- In this lecture we will discuss a technique called Backtracking

# Quick Review of Tree Terminology



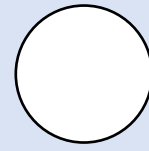
# Backtracking

- Uses a Search Tree
- The Root is the starting state before the search for solutions
- Nodes on the first level, choices made for the first component of the solution
- Nodes on the second level, choices for second component of the solution
- The pattern continues...
- Each node on the tree is considered a potential solution, if it corresponds to a partially constructed solution that may still lead to a full solution.
- Leaves in the tree are dead ends or complete solution.

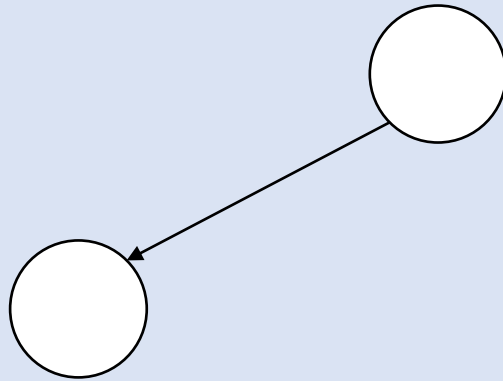
# The Search Tree Construction

- If the current node is considered a potential solution, its child is generated by adding the first remaining legitimate option for the next component of a solution
  - The algorithm moves onto the child
- If the current node is NOT considered a potential solution, then the algorithm “backtracks” to the parent to consider the next potential solution.
  - If no option is possible, then the algorithm backtracks up on more level in the search tree.
- If the potential solution turns out to be the complete solution, then the algorithm stops (assuming we are searching for one solution).

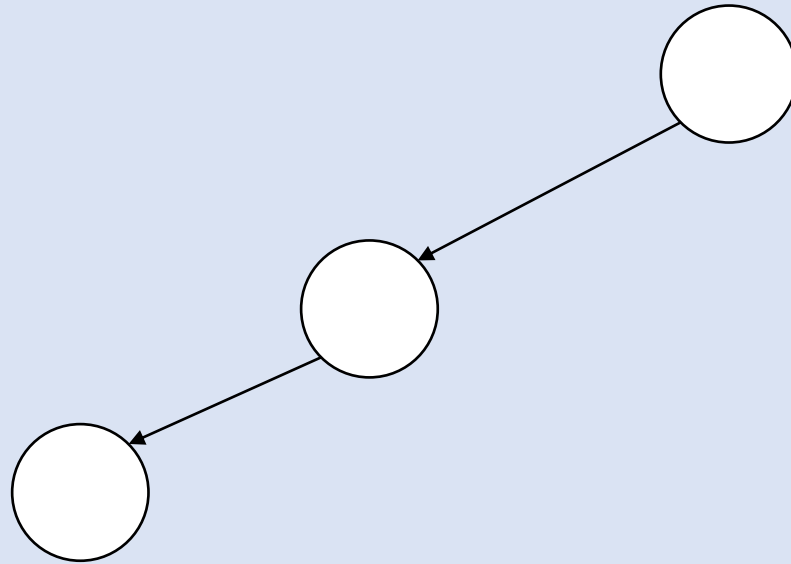
# The Big Picture of Backtracking



# The Big Picture of Backtracking

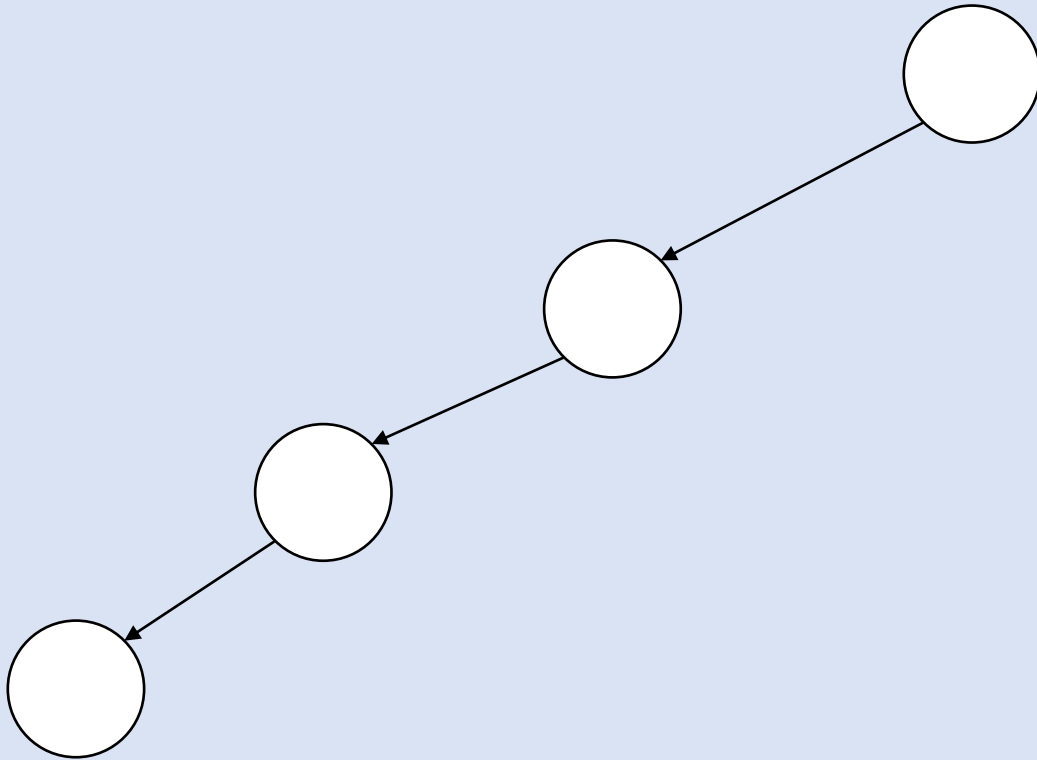


# The Big Picture of Backtracking

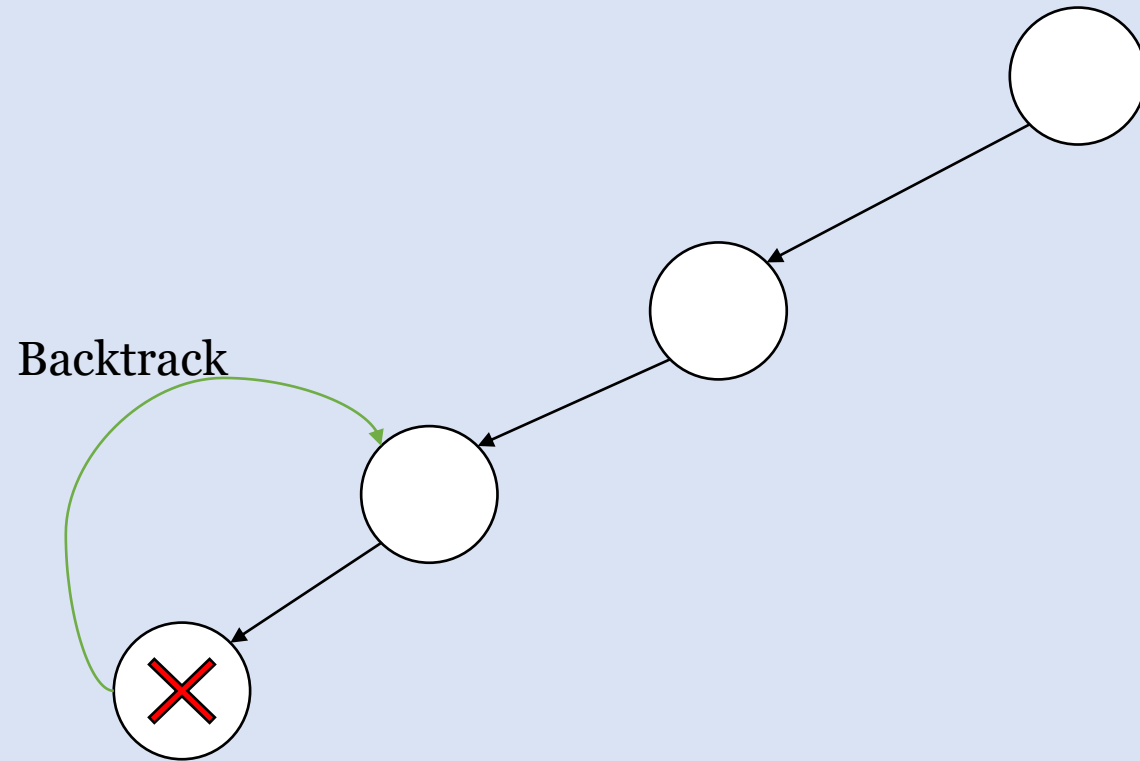




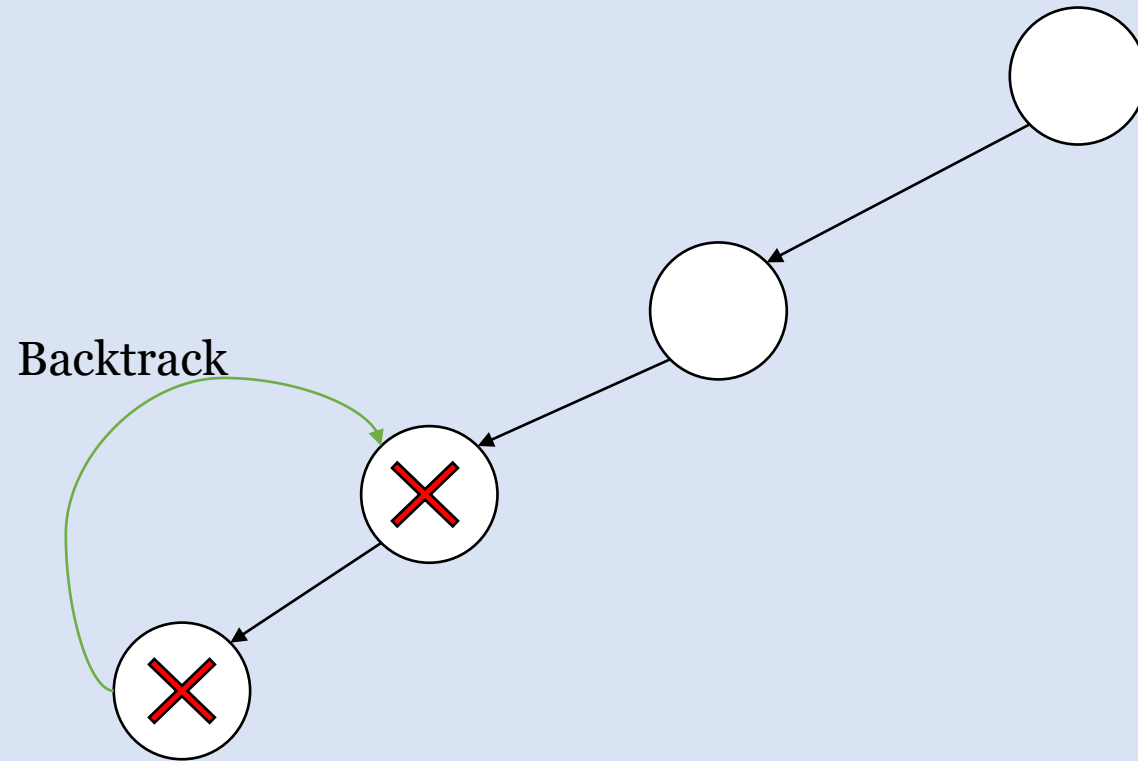
# The Big Picture of Backtracking



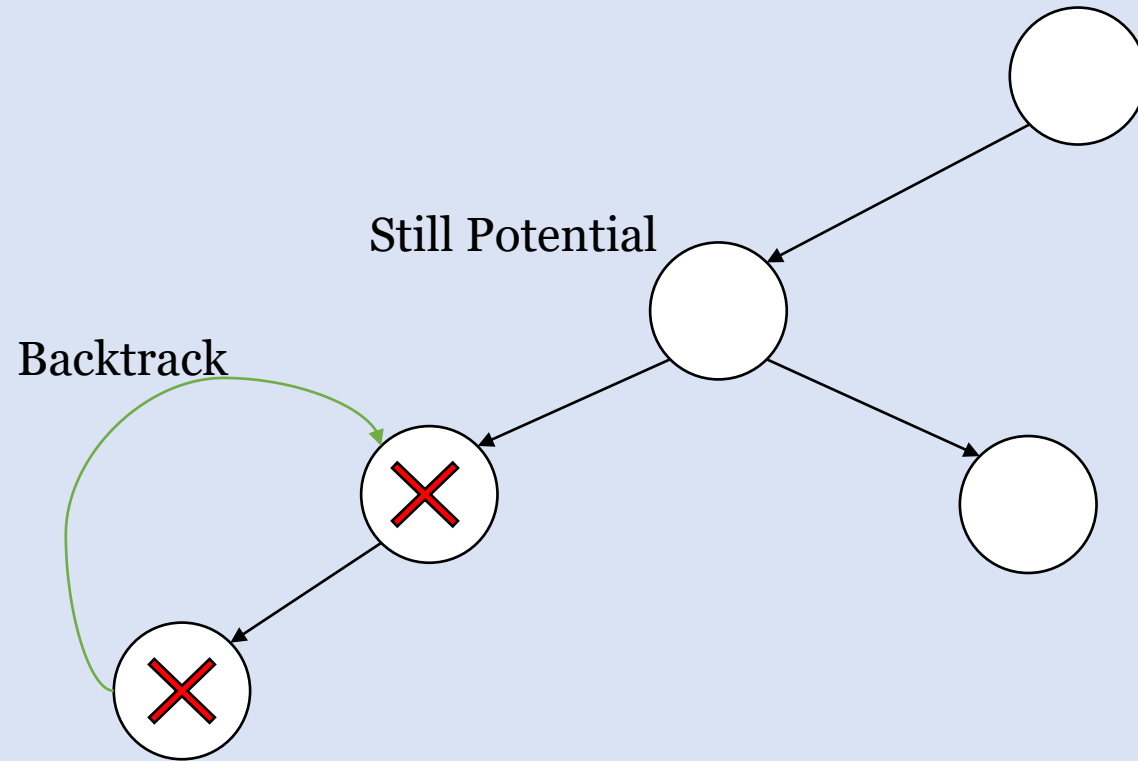
# The Big Picture of Backtracking



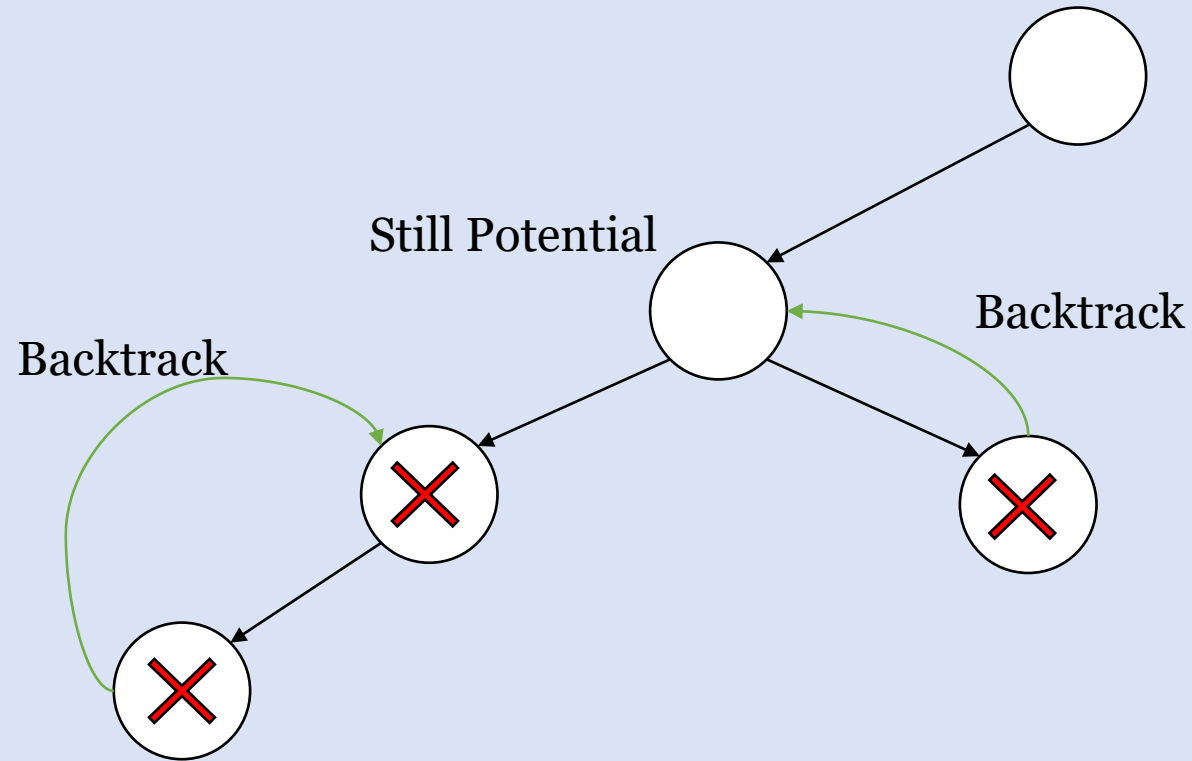
# The Big Picture of Backtracking



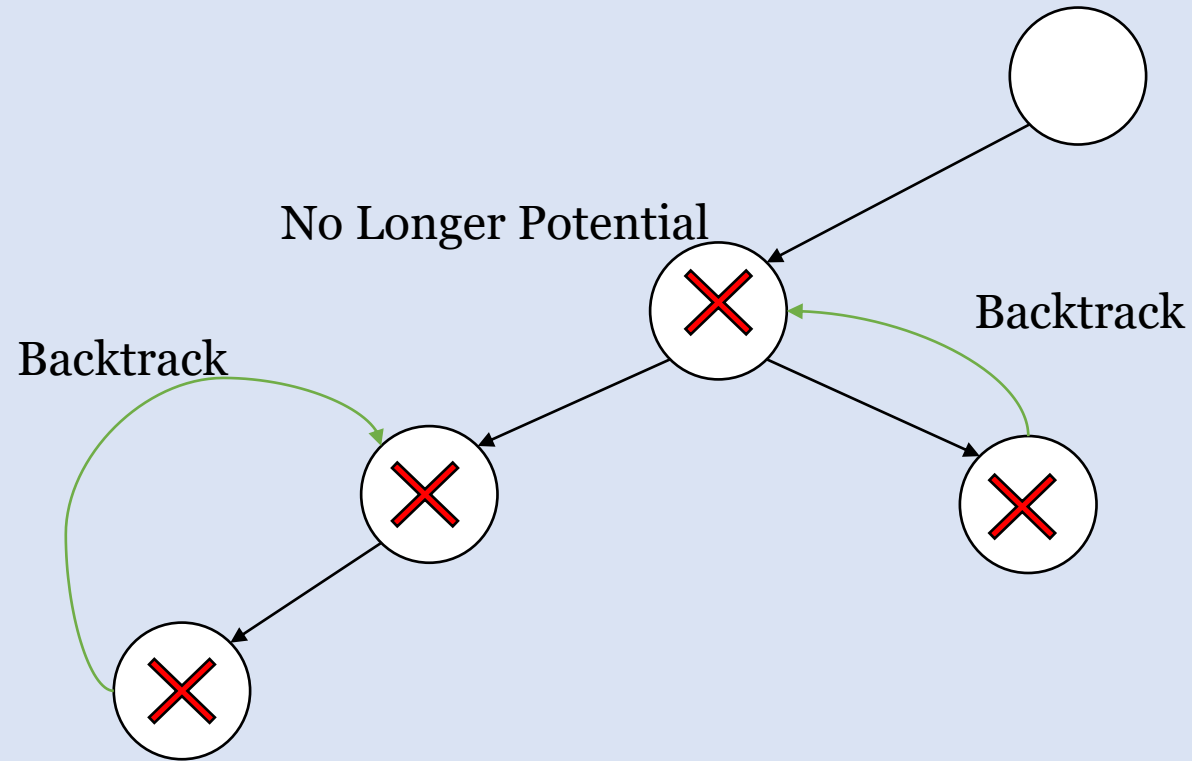
# The Big Picture of Backtracking



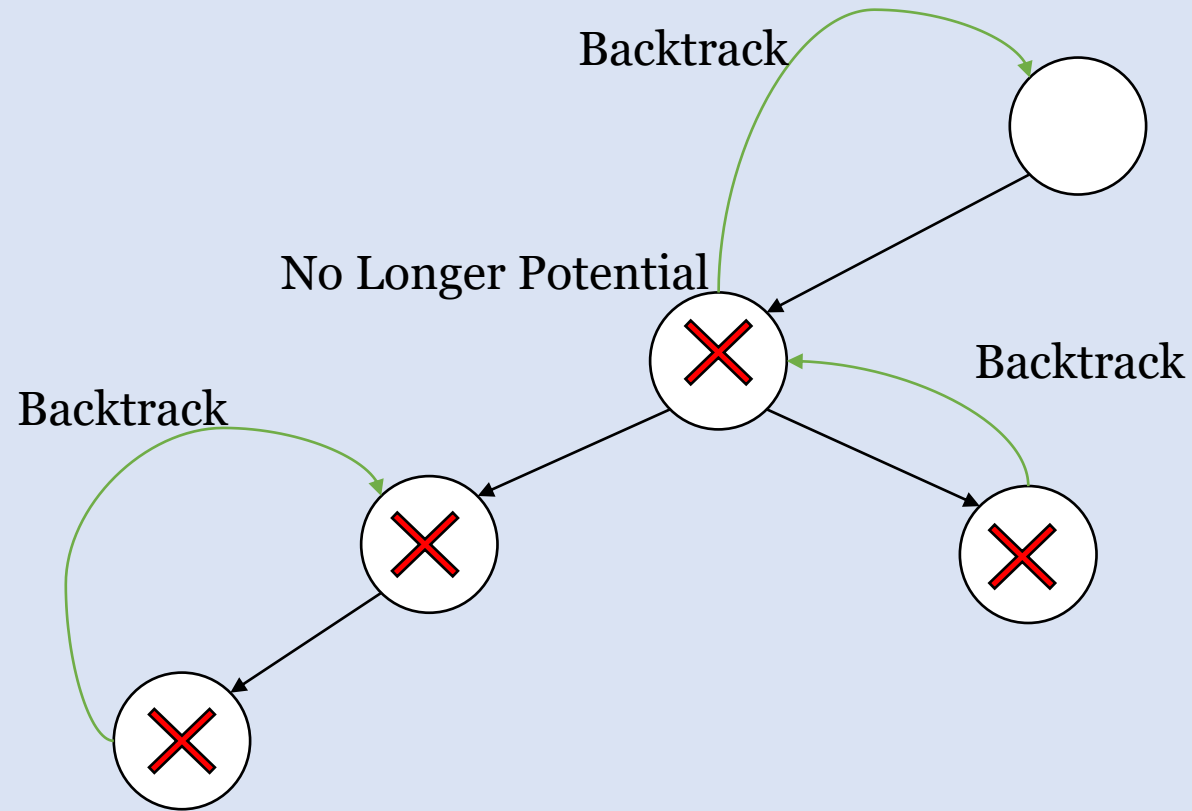
# The Big Picture of Backtracking



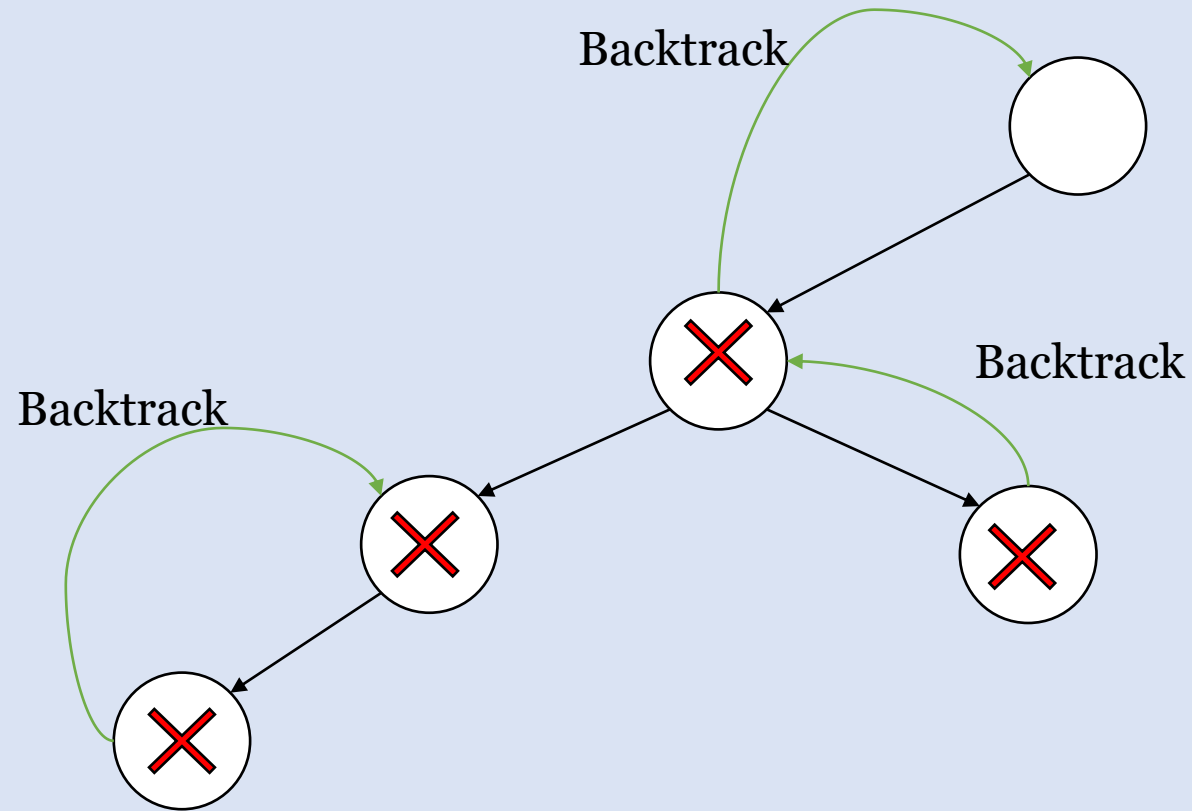
# The Big Picture of Backtracking



# The Big Picture of Backtracking

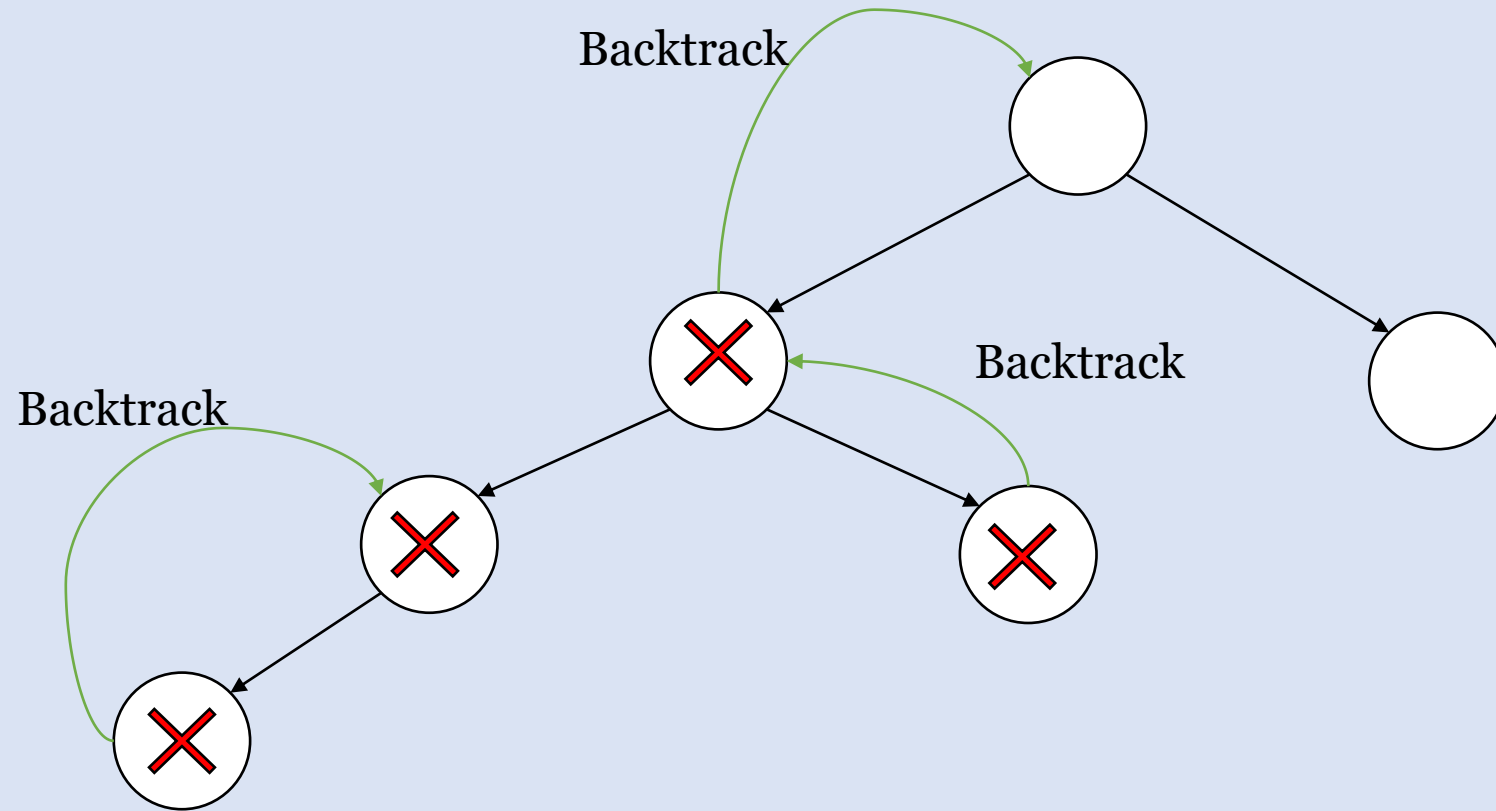


# The Big Picture of Backtracking

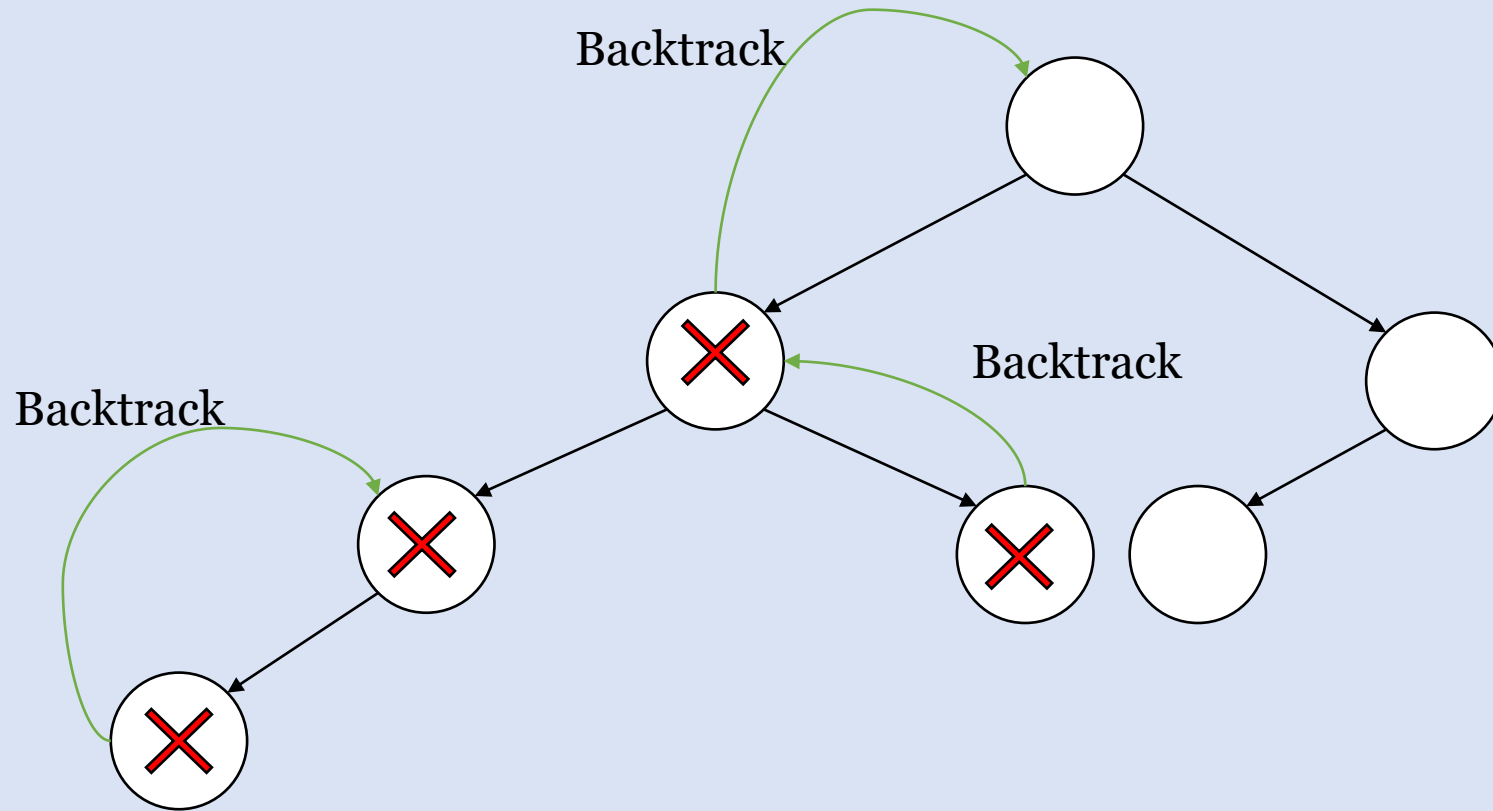




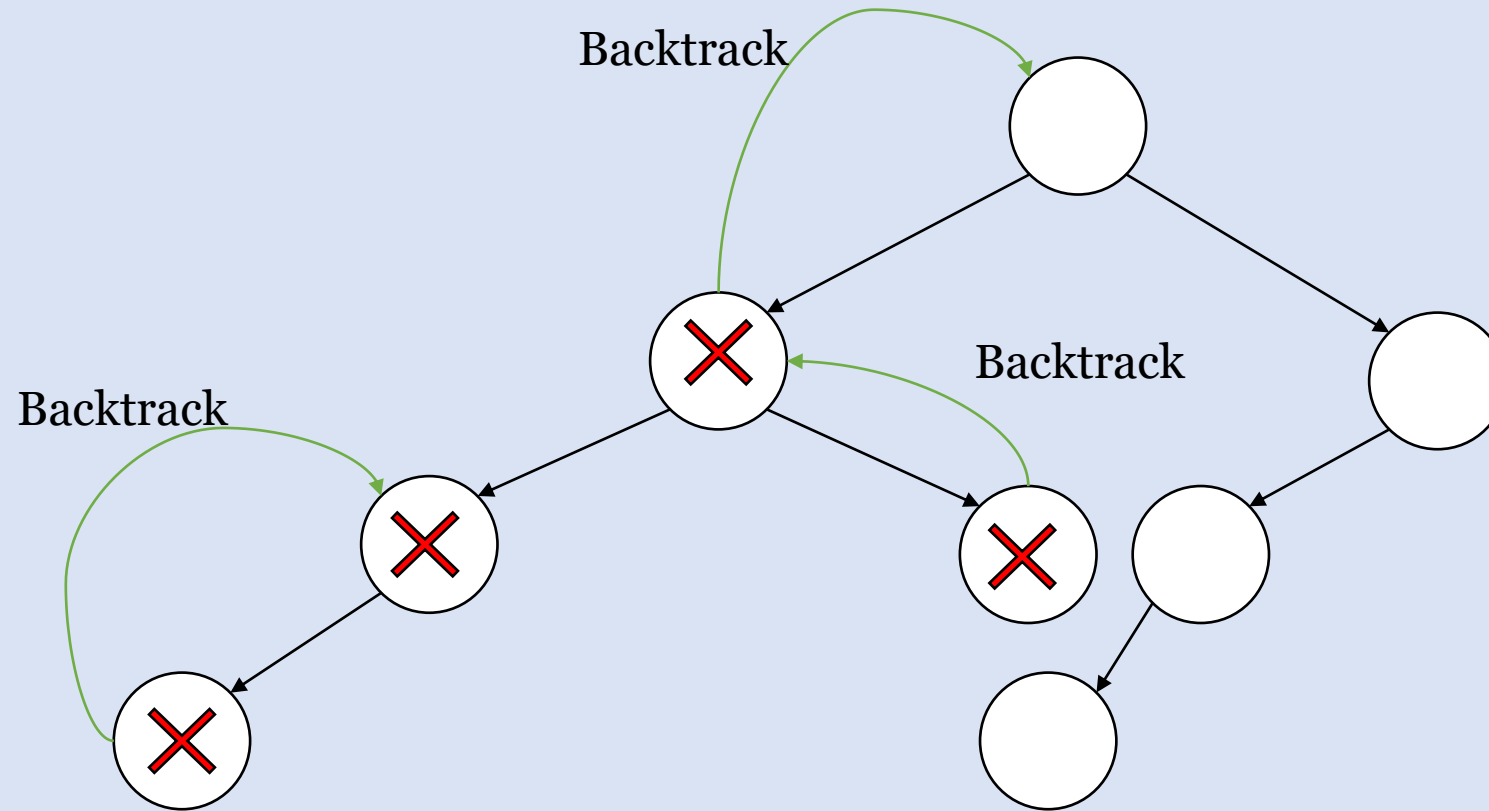
# The Big Picture of Backtracking



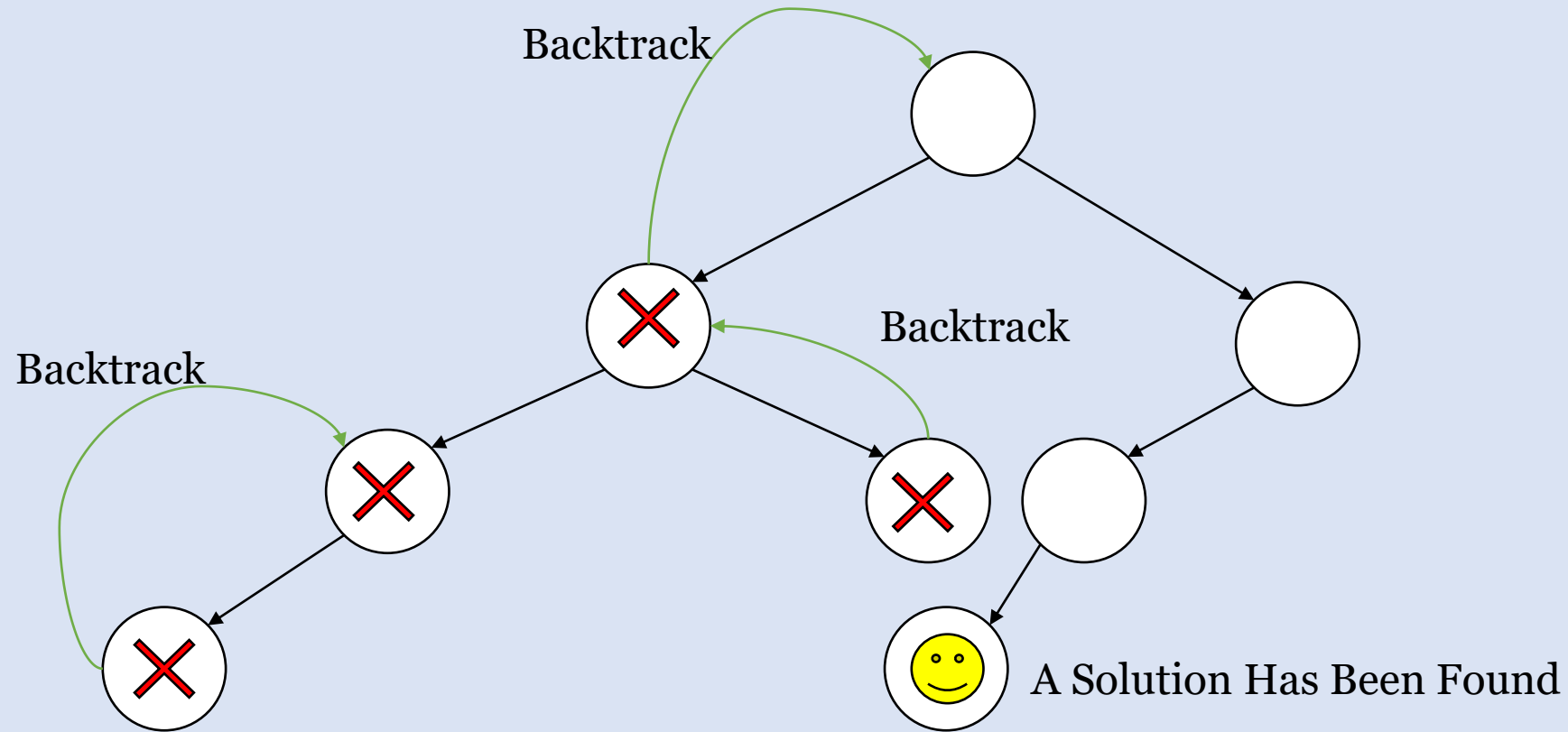
# The Big Picture of Backtracking



# The Big Picture of Backtracking



# The Big Picture of Backtracking



Based on observation of Backtracking what kind of problems can we apply this technique too? From the description, are there limitations?

Let's Observe the Classic N-Queens  
Problem

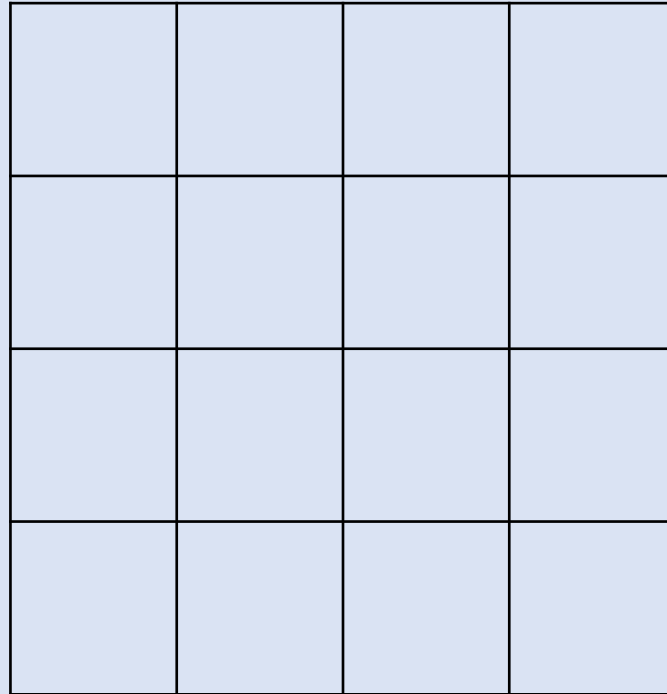
# Problem Definition

The  $n$ -queens problem is to place  $n$  queens on an  $n \times n$  board so that no two queens are in the same row, column, or diagonal.

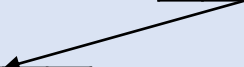
Sample  
4 queens on 4 x 4

		X	
X			
			X
	X		

# Search Tree Example of 4 queens on 4 x 4





X			



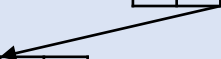

X			



X			
	X		

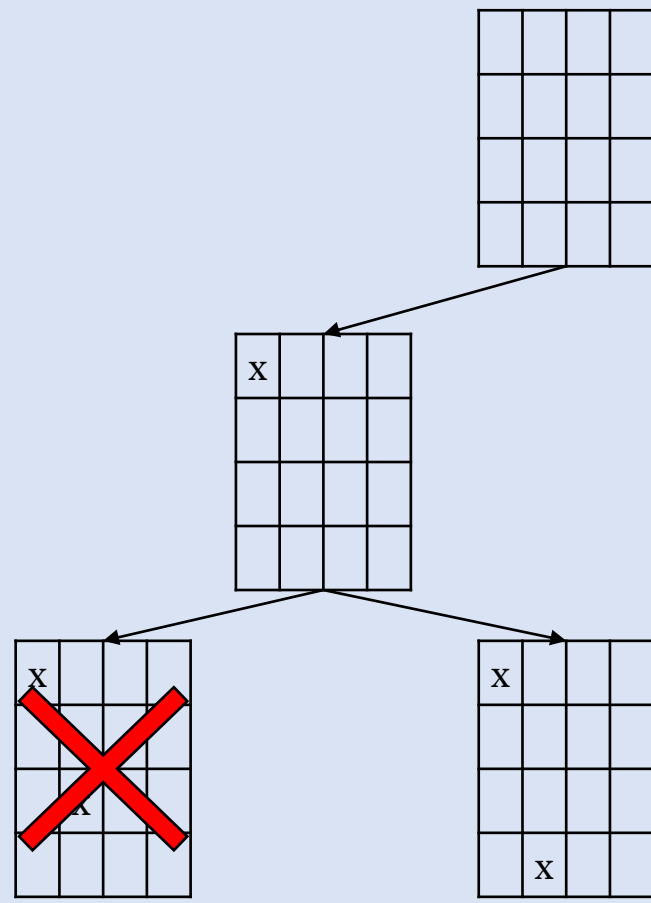


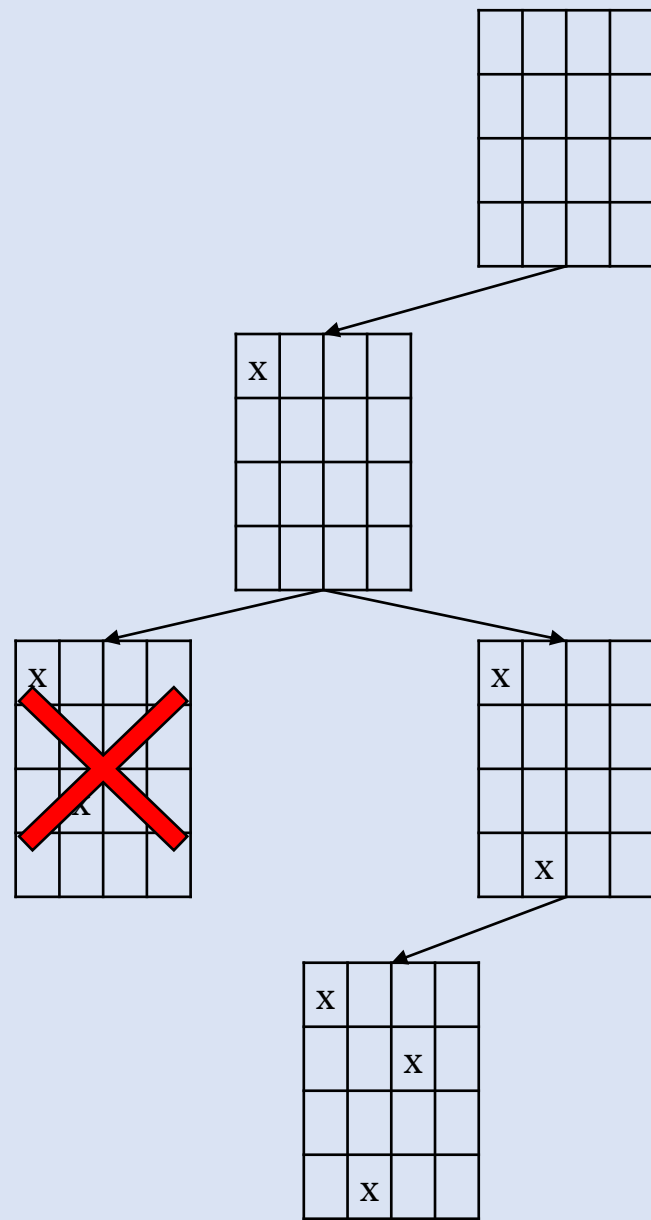

x			

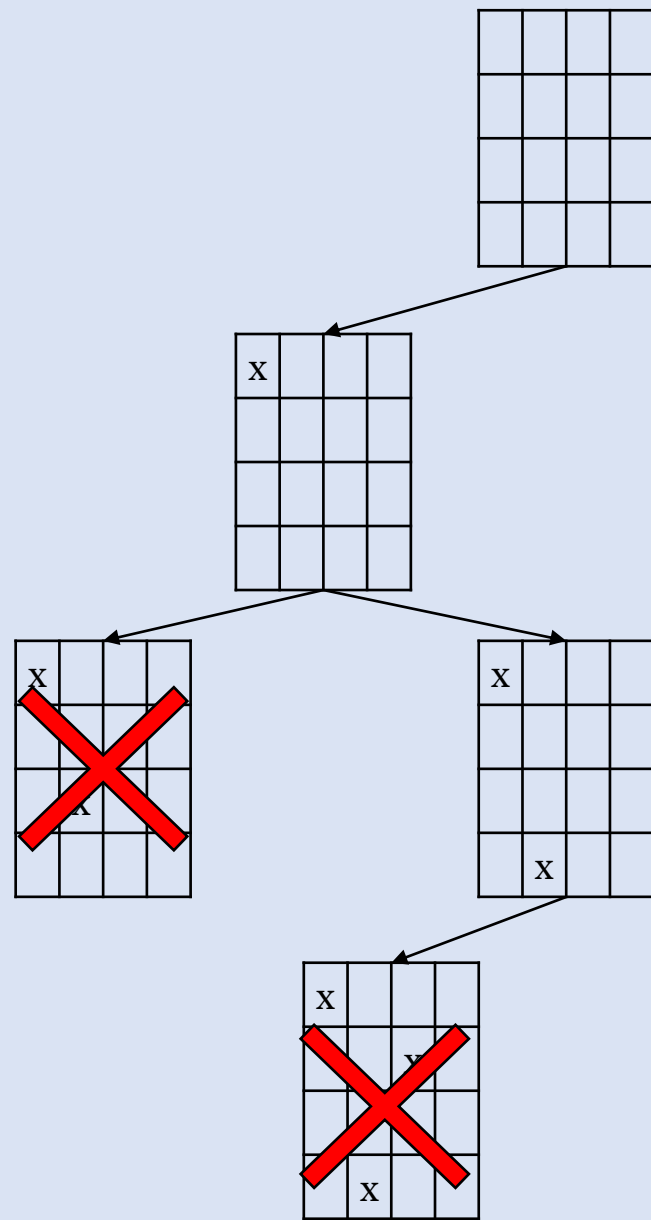


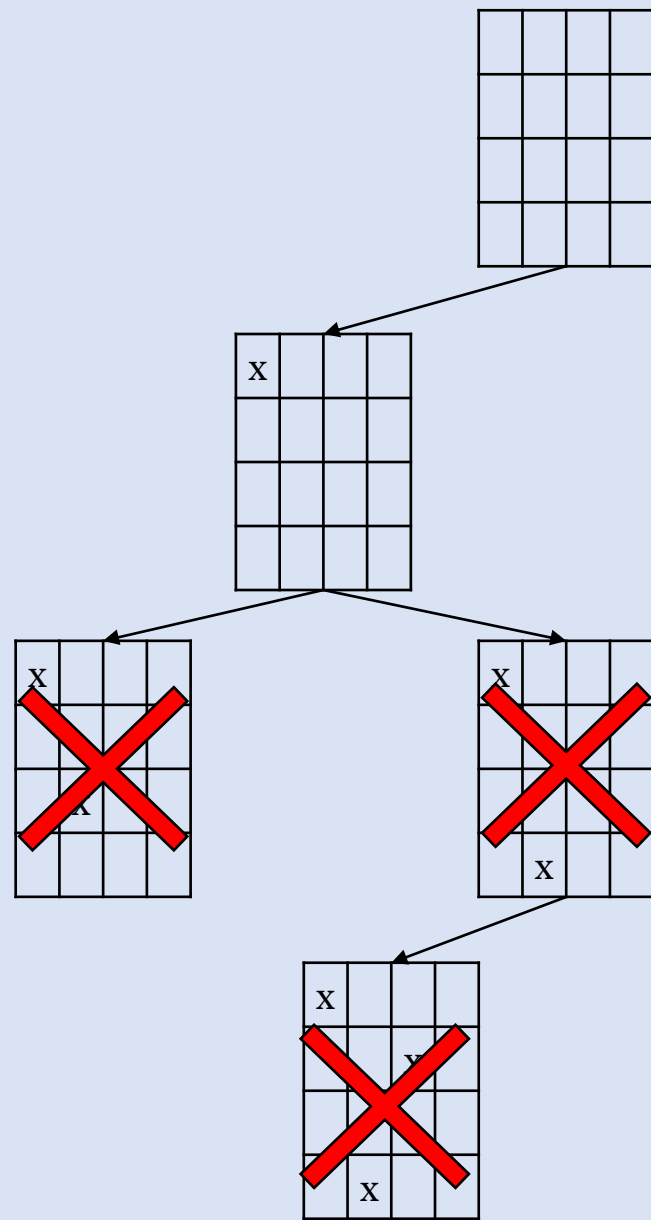
x			

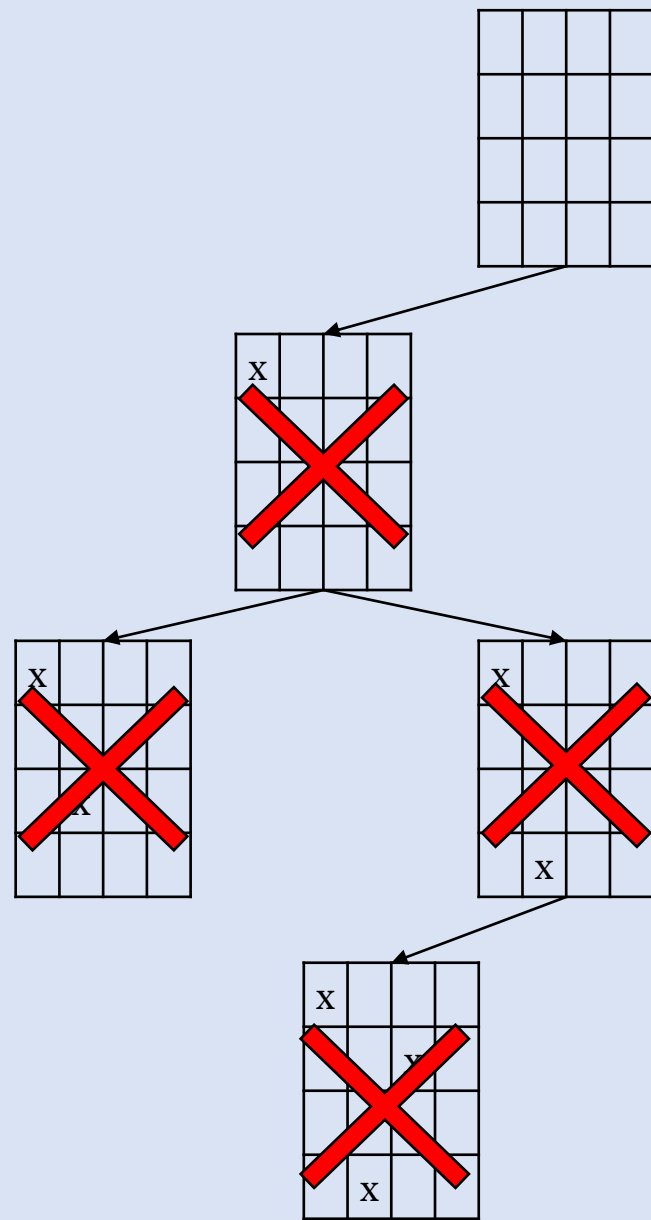
A large red 'X' mark is drawn over the entire bottom grid, crossing from the top-left to the bottom-right and from the top-right to the bottom-left.



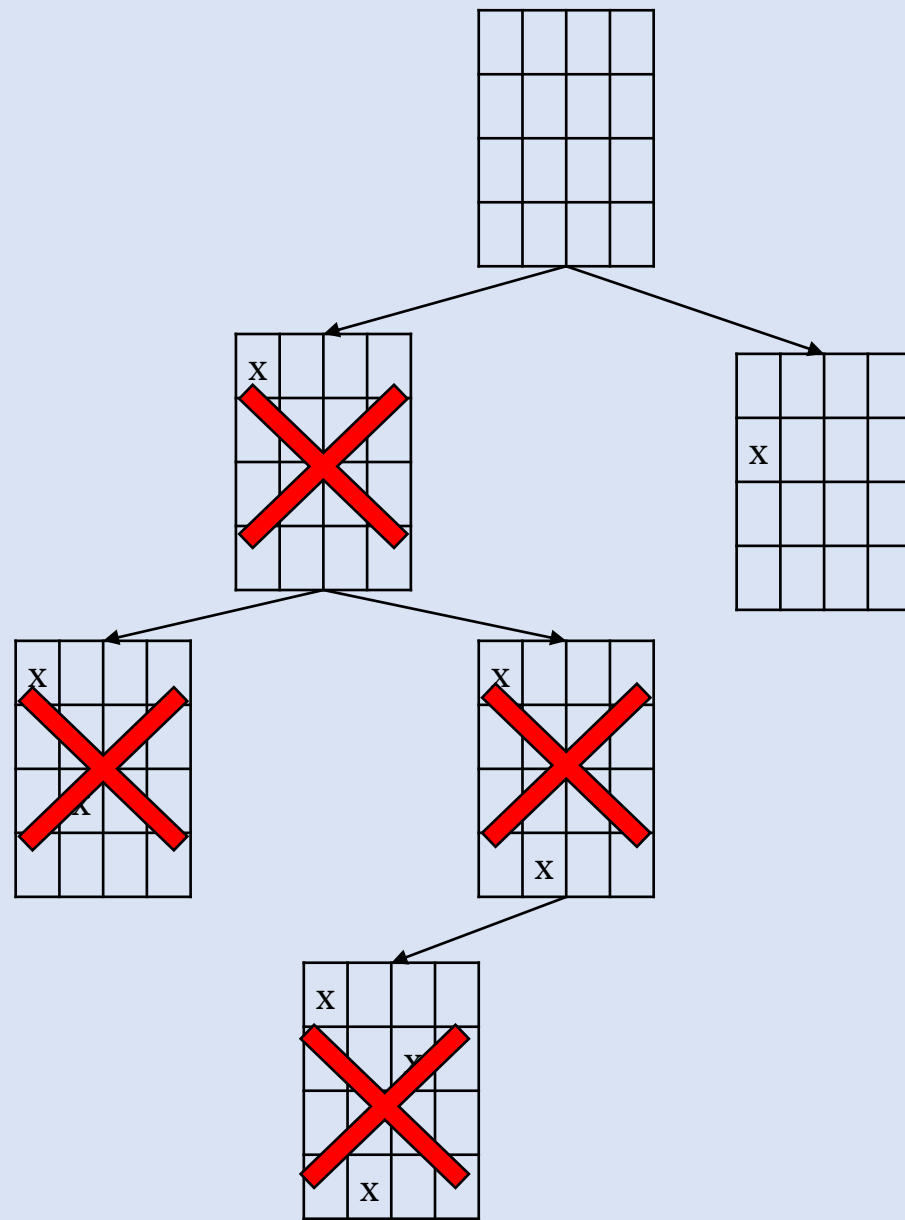


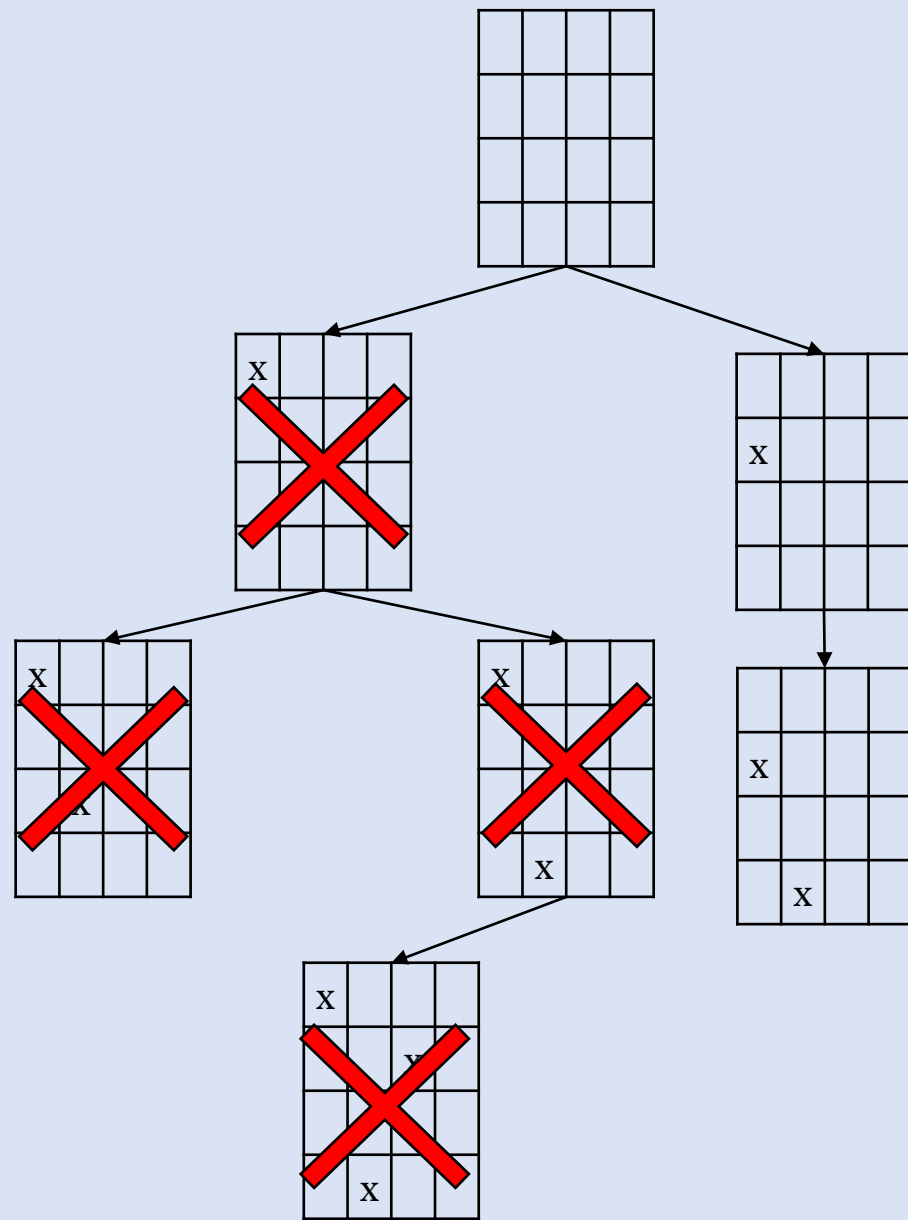


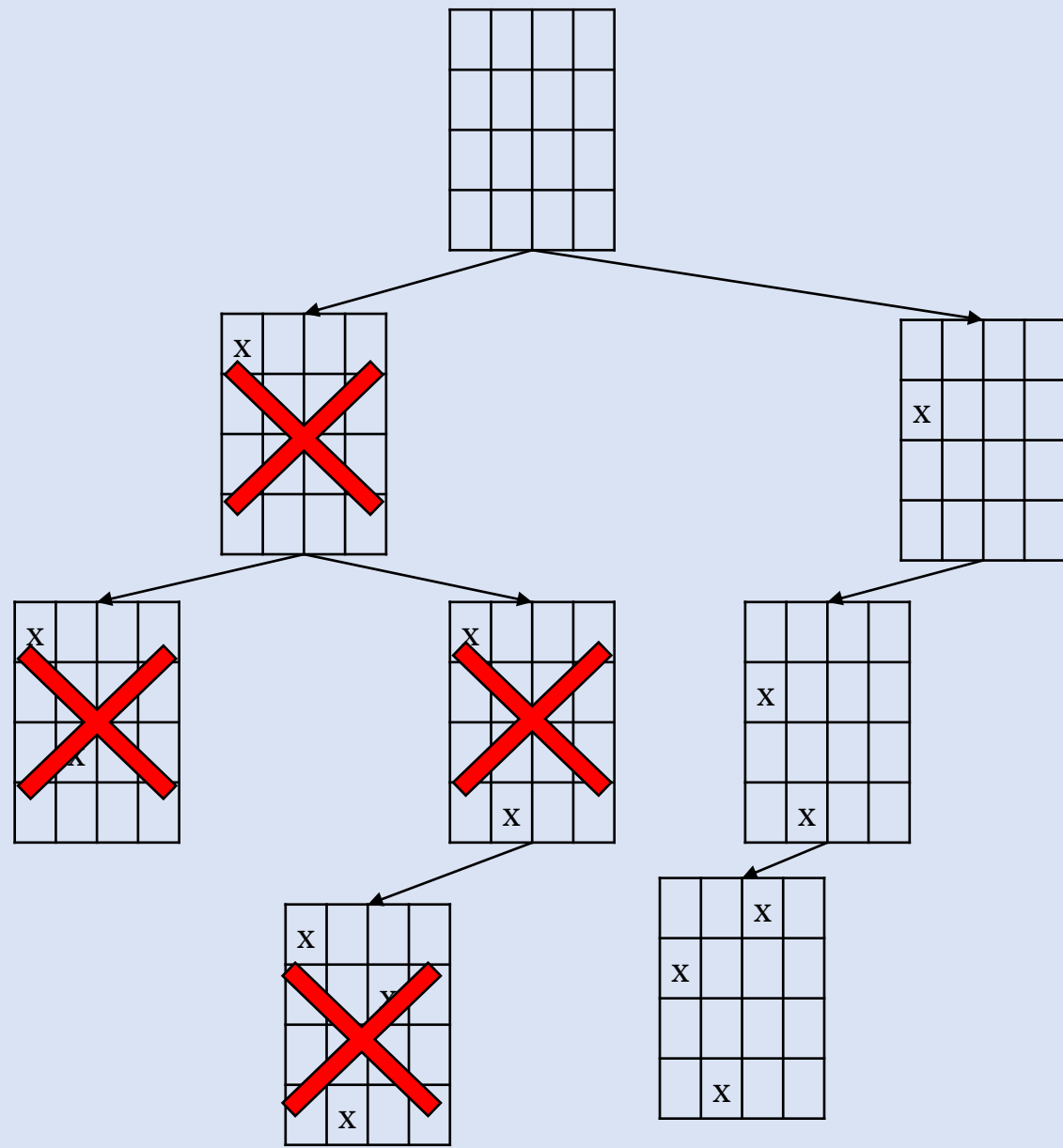


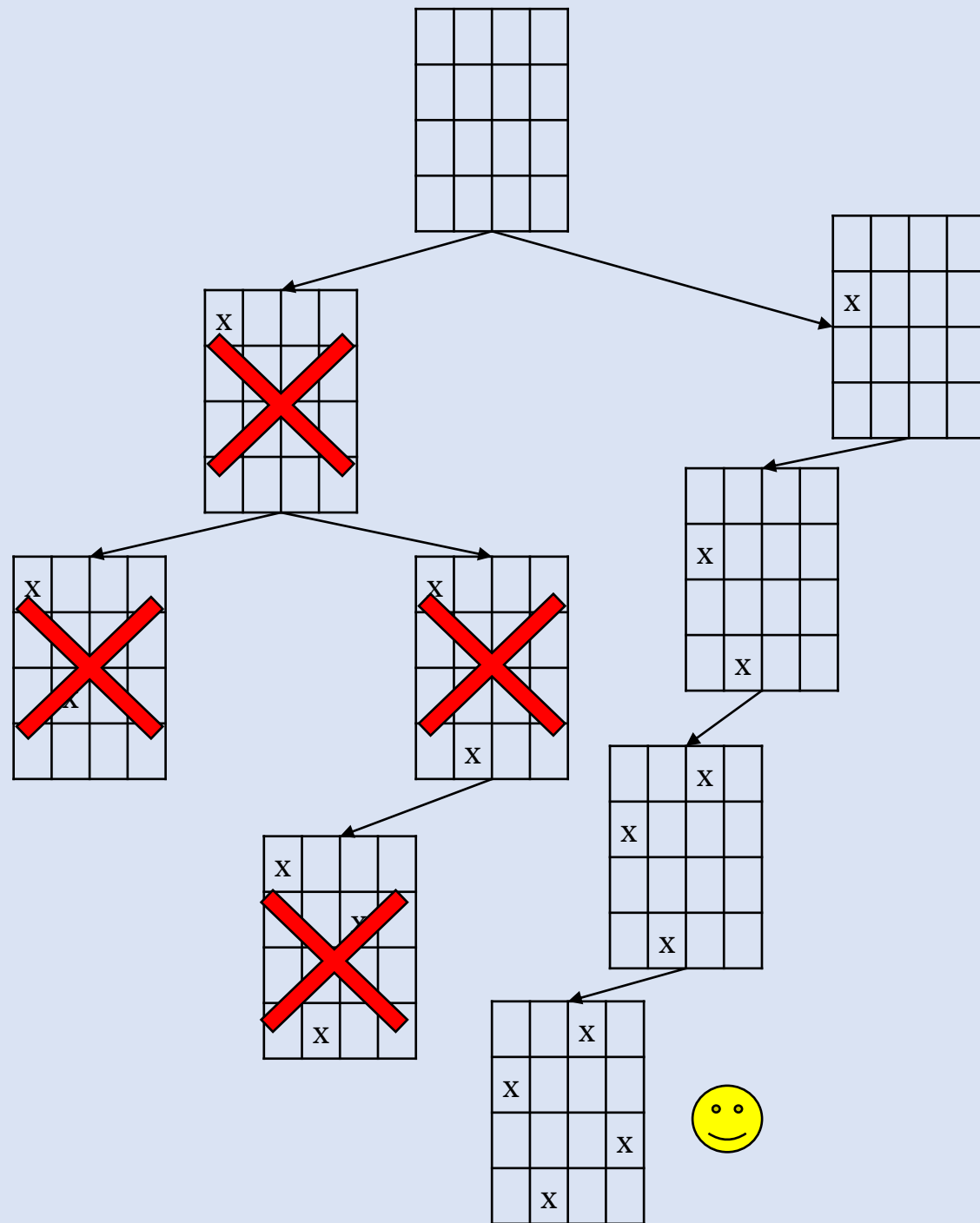












Now Lets Derive the Backtracking  
Algorithm for N-Queens

Now Lets Derive the Running Time  
Analysis

# The Skeleton Backtracking Algorithm

- This is a skeleton backtracking algorithm you can utilize in design your backtracking solutions to various problems
- `bound()` tests for a partial solution

```
backtrack(n)  
  backtrack(1,n)
```

```
rbacktrack(k, n)  
  for each  $x[k] \in S$   
    if (bound(k))  
      if (k == n)  
        //output the solution  
        for i = 1 to n  
          print(x[i] + “ ”)  
        println()  
      else  
        rbacktrack(k+1,n)
```