

• Branch and Bound BFS

It works by dividing the problem into sub problems, or branches, then eliminating certain branches based on bounds of optimal solutions. This process continues until best solution is found or all branches are explored.

• 0/1 Knapsack Using Branch & Bound

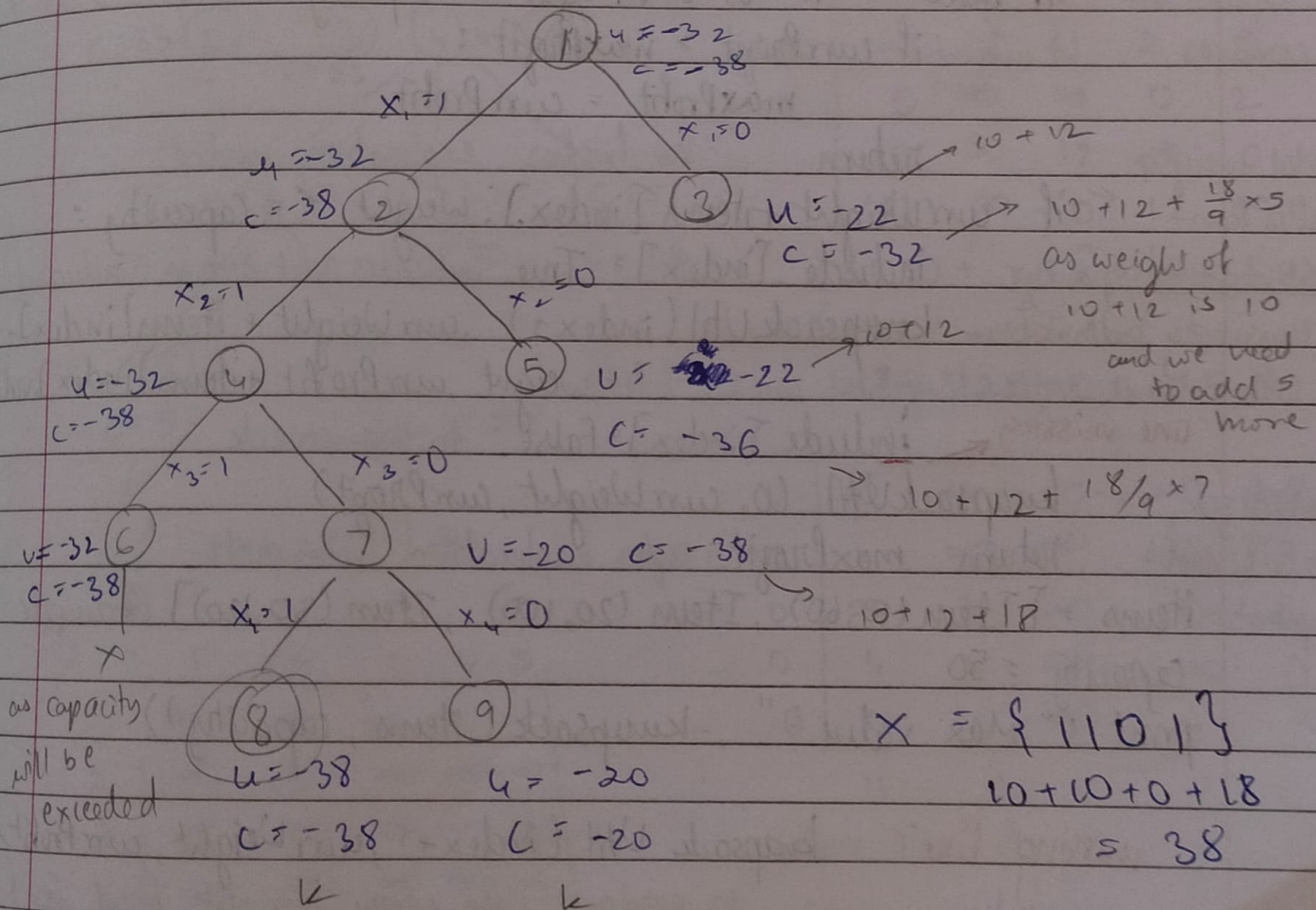
	1	2	3	4
Profit:	10	10	12	18
Weight:	2	4	6	9

$m=19 \rightarrow$ weight

$n=4$

upper bound - $u = \sum_{i=1}^n P_i x_i$

cost $C = \sum_{i=1}^n P_i x_i$ with fraction



Time complexity : $O(2^n)$

Space : $O(n)$

class Item : represents each item
~~def~~ def __init__(self, weight, value):
 constructor for ← self.weight = weight
 initialization self.value = value

def knapsack(items, capacity):

n = len(items) no. of items in the list

maxProfit = 0 currWeight = 0 currProfit = 0

include = [False] * n indicates whether each item is included or not

items sorted ← items.sort(key = lambda x: x.value / x.weight,
 based on $\lambda = \text{value} / \text{weight}$ descending reverse = True)

def knapsackUtil(index, currWeight, currProfit):
 recursive fn to explore all combos
 not loaded maxProfit and updated max Profit
 if index == n or currWeight == capacity: knapsack is full
 of current item if currProfit > maxProfit:
 maxProfit = currProfit

total weight after item ← return no further exploration needed.

if currWeight + items[index].weight ≤ capacity:

doesn't exceed capacity include[index] = True indicates that item is included

recursively calls it ← knapsackUtil(index + 1, currWeight + items[index].weight,

with next index & updated curr wt & Profit weight, currProfit + items[index].value

line missing ← include[index] = False to backtrack

initializes ← knapsackUtil(0, currWeight, currProfit)

sorts & starts return maxProfit after exploration is complete.

exploration items = [Item(10, 60), Item(20, 100), Item(30, 120)] Example

capacity = 50

print("Max value:", knapsack(items, capacity))

missing line: knapsackUtil(index + 1, currWeight, currProfit)
 to explore case where current item isn't
 included in the knapsack

Time Complexity: $O(n!)$
Space Complexity: $O(n^2)$

def TSP(n, distanceMatrix):

bestTour = [] list to find best tour found so far

bestTourLength = float('inf') initialized to +ve ∞ to ensure shorter length

recursive def search(k, tour):

explores all non local bestTour, bestTourLength

tours starting from city 0 if k == n: all cities visited

tourLength = calculate TL (tour, distanceMatrix) calculate current TL

if tourLength < bestTourLength:

if current tour length < best, we update it bestTour = tour.copy()

bestTourLength = tourLength

return

for i in range(k): for each city k, it tries inserting into all posⁿ in current tour

copies current tour newTour = tour[:k]

inserts city k at i newTour.insert(i, k)

before exploring further, it if calculate Lower Bound (newTour, distanceMatrix)

checks if lower bound of tour < best TL, < bestTourLength:

if not, the branch is pruned search(k+1, newTour) → next city & new tour

search(1, 10)

return bestTour, bestTourLength

def calculateTourLength^{TL}(tour, distanceMatrix): calculates TL considering

iterates over city index tourLength = 0

distance from last to starting city

i to tour, except last city for i in range(len(tour) - 1):

for each city, it adds dist from tourLength += distanceMatrix[tour[i]][tour[i+1]]

current to next city tourLength += distanceMatrix[tour[-1]][tour[0]]

in TL return tourLength

for remaining TL

def calculateLowerBound(tour, distanceMatrix): by summing min

tourLength = calculate TL (tour, distanceMatrix) distances from

missing Cities = set(range(len(distanceMatrix))) - set(tour)

identifies set of cities not included in tour

each unvisited city to nearest unvisited city

it finds the best city cities excluded

Lower Bound = tour Length \rightarrow break missing city
for city in missing Cities:

if finds this from \leftarrow min Distance = $\min(\text{distance Matrix}[\text{city}][\text{otherCity}])$
that city to another city in the set of missing for other city in missing Cities if otherCity != city
cities excluding itself
Lower Bound + min Distance
return Lower Bound \rightarrow min additional distance required to complete
tour assuming shortest paths are taken
b/w missing cities.

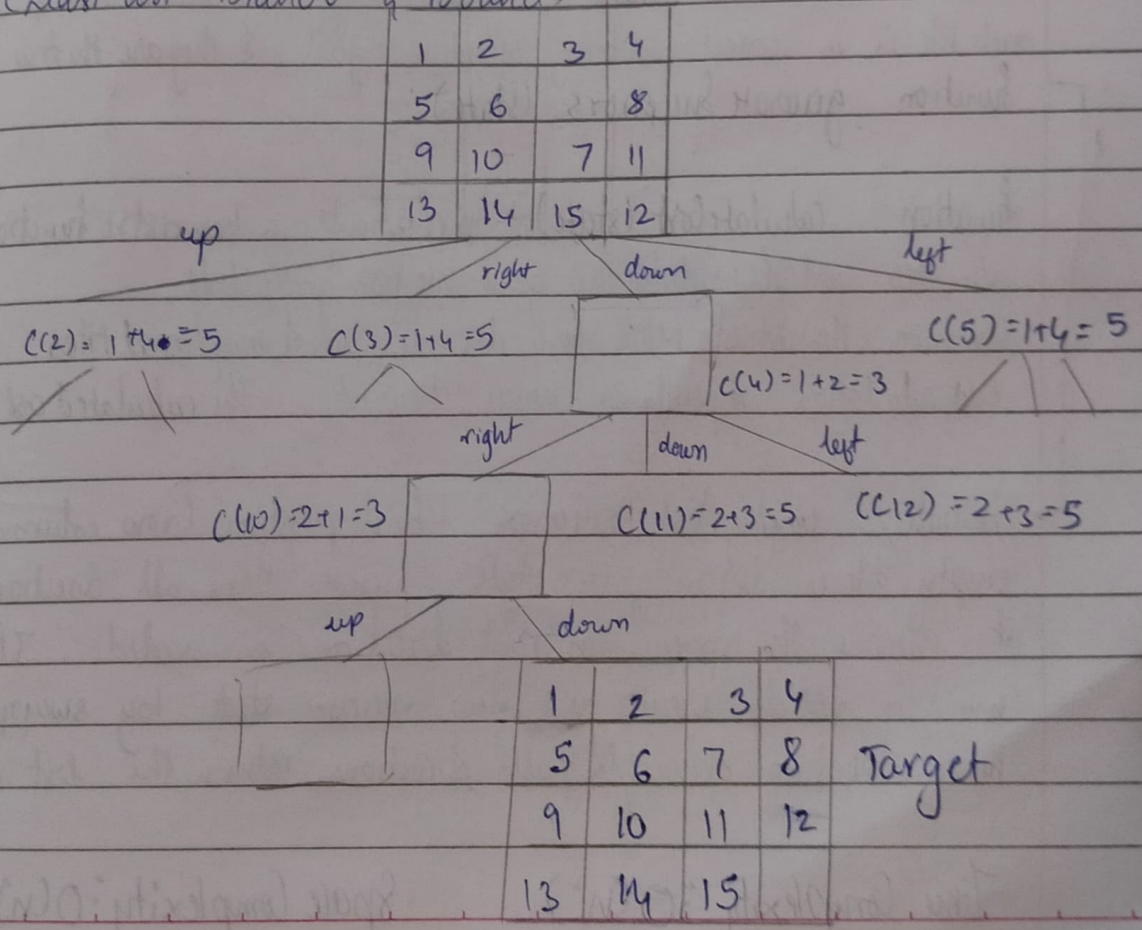
There's a very short algo in mam's PPT \hearts

15 - Puzzle Problem

Move the blank piece ~~to~~ one step each time to reach the final or correct stage.

Cost = $f(x) + g(x) \rightarrow$ how many pieces are away from final/correct position
 \rightarrow no. of moves made

Drawing every branch is hard, so we can follow LC-BB
(Least Cost Branch & Bound)



function solve15Puzzle(initialState):

openList = Priority Queue() to store partial solⁿ

openList.put(Node(initialState, 0)) Enqueue initial state

while not openList.empty():

currentNode = openList.get() Dequeue lowest cost node. ^{from openList}

check if current ← if isGoalState(currentNode.state):

state is goal node return currentNode.state return goal state

Generate successor states ← successors = generateSuccessors(currentNode.state)

for successor in successors:

calculate successor state cost ← successorCost = calculateCost(successor)

enqueue successor state ← openList.put(Node(successor, successor))

return None No solution found

class Node: to represent each state of the puzzle

def __init__(self, state, cost):

self.state = state state of the puzzle (board config)

self.cost = cost cost of reaching this state.

function isGoalState(state): iterate through each tile in state & check if

each tile is in correct position according to goal, if they are the true else false

function generateSuccessors(state):

function calculateCost(state): implements a heuristic function that estimates cost of reaching the goal state from given state.

Common Heuristics: Manhattan distance, no. of misplaced tiles

Cost calculation depends on chosen heuristic. The calculated cost is returned

→ Initialize empty list 'successors'. Find position (row column) of empty tile is (0) in given state, iterate thru all directions, check if moving the space in that direction is valid. If the move is valid, create a new successor state by swapping. After iterating through all directions, return the list of successors

Time Complexity: $O(n^2)$

Space Complexity: $O(n)$