

Data Structures and Algorithms

LECTURE 07: GREEDY ALGORITHMS

- Greedy Algorithms
- Greedy Failure Cases
- Optimal Greedy Algorithms

Greedy Algorithms

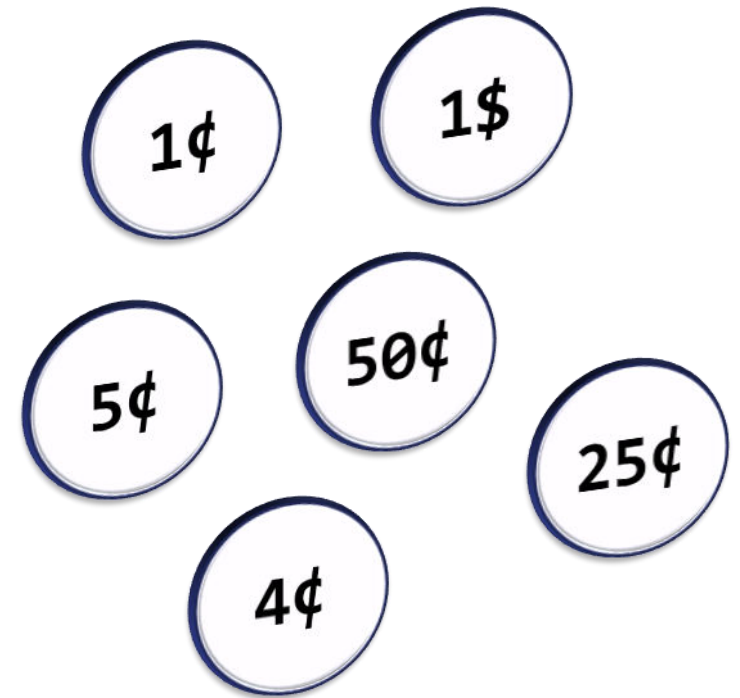
- Used for solving optimization problems
- Usually more efficient than the other algorithms
- Can produce a **non-optimal** (incorrect) result
- Pick the **best local** solution
 - The optimum for a **current** position and point of view
- Greedy algorithms assume that always choosing a **local** optimum leads to the **global** optimum

Optimization Problems

- Finding the best solution from all possible solutions
- Examples:
 - Find the **shortest** path from Sofia to Varna
 - Find the **maximum increasing subsequence**
 - Find the shortest route that visits each city and returns to the origin city

Problem: Sum of Coins

- Write a program, which gathers a sum of money, using the least possible number of coins
- Consider the US currency coins
 - 0.01, 0.02, 0.05, 0.10
- **Greedy algorithm** for "Sum of Coins":
 - Take the largest coin while possible
 - Then take the second largest
 - Etc.



Sum of Coins Visualization

Target: 18



Actual: 0

Sum of Coins Visualization

Target: 18

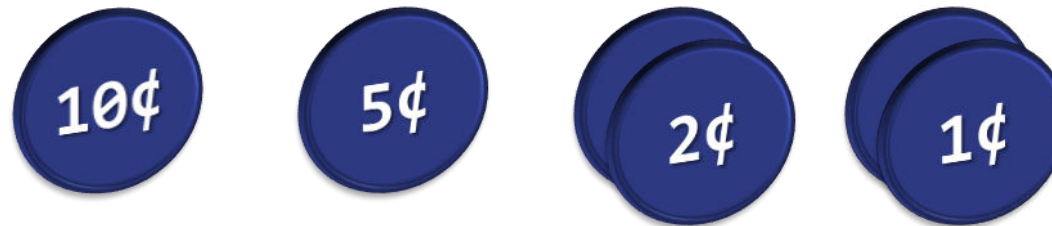


Actual: 10



Sum of Coins Visualization

Target: 18

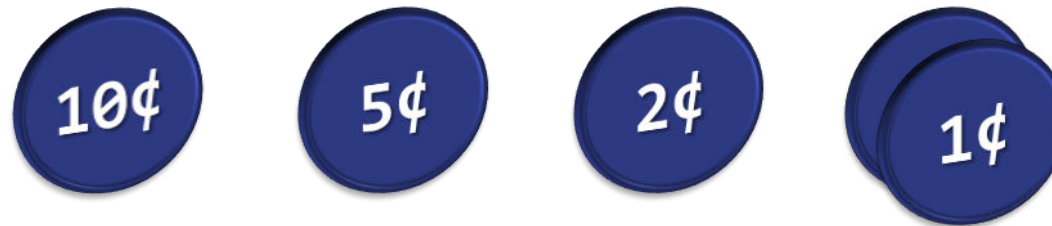


Actual: 15



Sum of Coins Visualization

Target: 18



Actual: 17



Sum of Coins Visualization

Target: 18



Actual: 18



Solution: Sum of Coins (1)

```
public static Map<Integer, Integer>
    chooseCoins(int[] coins, int targetSum) {
    List<Integer> sortedCoins = Arrays.stream(coins).boxed()
        .sorted(Collections.reverseOrder())
        .collect(Collectors.toList());
    Map<Integer, Integer> chosenCoins = new LinkedHashMap<>();
    int currentSum = 0; int coinIndex = 0;
    // Next slide
    if (currentSum != targetSum)
        throw new IllegalArgumentException();
    return chosenCoins;
}
```

Solution: Sum of Coins (2)

```
while (currentSum != targetSum && coinIndex < sortedCoins.size()) {  
    int currentCoin = sortedCoins.get(coinIndex);  
    int remainder = targetSum - currentSum;  
    int numberOfCoins = remainder / currentCoin;  
    if (currentSum + currentCoin <= targetSum) {  
        chosenCoins.put(currentCoin, numberOfCoins);  
        currentSum += numberOfCoins * currentCoin;  
    }  
    coinIndex++;  
}
```

Problem: Set Cover

- Write a program that finds the smallest subset of S , the union of which = U (if it exists)
- You will be given a **set** of integers U called "**the Universe**"
- And a set S of n integer sets whose union = U

Universe: 1, 2, 3, 4, 5

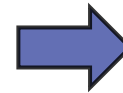
Number of sets: 4

1

2, 4

5

3



Sets to take (4):

{ 2, 4 }

{ 1 }

{ 5 }

{ 3 }

Solution: Set Cover (1)

```
public static List<int[]> chooseSets(  
    List<int[]> sets, List<Integer> universe) {  
    List<int[]> selectedSets = new ArrayList<>();  
    Set<Integer> universeSet = new HashSet<>();  
    for (int element : universe) { universeSet.add(element);}  
    while (!universeSet.isEmpty()) {  
        // Next Slide  
    }  
    return selectedSets;  
}
```

Solution: Set Cover (2)

```
int notChosenCount = 0;
int[] chosenSet = sets.get(0);
for (int[] set : sets) {
    // Next slide
}
selectedSets.add(chosenSet);
for (int elem : chosenSet) {
    universeSet.remove(elem);
}
```

Solution: Set Cover (3)

```
int count = 0;
for (int elem : set) {
    if (universeSet.contains(elem)) {
        count++;
    }
}
if (notChosenCount < count) {
    notChosenCount = count;
    chosenSet = set;
}
```


Greedy Failure Cases

- Greedy Algorithms Often Fail

Sum of Coins Failure

Target: 18



Actual: 0

Sum of Coins Failure

Target: 18

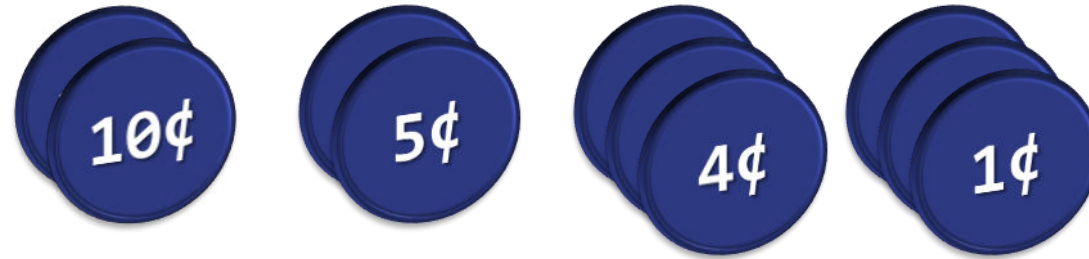


Actual: 10



Sum of Coins Failure

Target: 18



Actual: 15



Sum of Coins Failure

Target: 18



Actual: 16



Sum of Coins Failure

Target: 18



Actual: 17



Sum of Coins Failure

Target: 18



Actual: 18



Sum of Coins Failure

Target: 18



Optimal Greedy Algorithms

- Optimal Substructure and Greedy Choice Property

Optimal Greedy Algorithms

- Suitable problems for greedy algorithms have these properties:
 - **Greedy choice property**
 - **Optimal substructure**
- Any problem having the above properties is guaranteed to have an optimal greedy solution

Greedy Choice Property

- **Greedy choice property**
 - **A global optimal solution** can be obtained by greedily selecting a **locally optimal** choice
 - Sub-problems that arise are solved by consequent greedy choices
 - Enforced by optimal substructure

Optimal Substructure Property

- **Optimal substructure property**
 - After each greedy choice the problem remains an optimization problem of the same form as the original problem
 - **An optimal global solution contains the optimal solutions of all its sub-problems**

Greedy Algorithms: Example

- The "**Max Coins**" game
 - You are given a set of coins
 - You play against another player, alternating turns
 - Per each turn, you can take up to three coins
 - Your goal is to have as many coins as possible at the end



Max Coins – Greedy Algorithm

- A simple **greedy strategy** exists for the "Max Coins" game

At each turn take the maximum number of coins

- Always choose the local maximum (at each step)
 - You don't consider what the other player does
 - You don't consider your actions' consequences
- The **greedy algorithm** works optimally here
 - It takes as many coins as possible

Summary

- Greedy Algorithms
- Optimal Greedy Algorithms