

# Algorithms and Data Structures (CSci 115)

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# Learning outcomes

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- Greedy programming
  - Definitions and principles
  - Examples
- Rationale
  - Many graph walk algorithms use greedy approaches

# Introduction

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## ■ Optimization problems

➤ For most optimization problems, you desire to find ideally the **best** solution

○ Policy:

- Sequence of steps leading to the solution
  - You may not be able to isolate each step from the others (dynamic programming)
  - You may be able to pick what's the best right now ( #yolo)

## ■ A **greedy algorithm** sometimes works well for optimization problems

➤ Multiple steps:

1. Take the best you can get **right now**,
  - Without regard for future consequences.
2. Hope that by choosing a local optimum at each step, you will end up at a global optimum.

# Example: counting coins

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- Problem

- To count out a certain amount of money, using the **fewest possible bills and coins**

- Greedy algorithm:

- 1. At each step, take the **largest** possible bill or coin that does not overshoot

- Example: To make \$16.39, you can choose:

- A \$10 bill
    - A \$5 bill
    - A \$1 bill
    - A 25¢ coin
    - A 10¢ coin
    - Four 1¢ coins

- The greedy algorithm always gives the optimum solution with US bills and coins

# Example: counting coins

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- Same problem
  - Monetary system: “klop” come in 1 klop, 7 klops, and 10 klop coins
- With the greedy algorithm to count out 15 klops
  - A 10 klop piece
  - Five 1 klop pieces, for a total of 15 kops
  - → 6 coins
- Best solution
  - Two 7 klop pieces and one 1 klop piece
  - → 3 coins
- Conclusion
  - The greedy algorithm provides a solution, but not in an optimal solution

# Example: scheduling

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## ■ Problem

- Example with CPU scheduling
- You have to run 9 jobs
  - running times of 3, 5, 6, 10, 11, 14, 15, 18, and 20 minutes.
- You have 3 processors that can be used to run these jobs.

## ■ Strategy

- Do the **longest**-running jobs first, on whatever processor is available.
- Do the **shortest**-running jobs first, on whatever processor is available

# Strategy

- A greedy algorithm obtains an optimal solution to a problem
  - By making a **sequence of choices**.
  - At each decision, the algorithm makes choice that seems best at the moment.
- Warning:
  - This heuristic strategy does not always produce an optimal solution
    - **Heuristic:**
      - Technique designed for solving a problem more quickly, or for finding an approximate solution when regular methods fail to find any exact solution, or are too slow.
  - **but** sometimes it does.
- Steps:
  1. Determine the optimal substructure of the problem.
  2. Develop a **recursive** solution.
  3. Show that if we make the greedy choice, then only 1 sub-problem remains.
  4. Prove that it is always safe to make the greedy choice.
    - Steps 3 and 4 can occur in either order.
  5. Develop a recursive algorithm that implements the greedy strategy.
  6. Convert the recursive algorithm to an iterative algorithm.

# Design

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## ■ Steps

1. Cast the optimization problem as one in which we make a choice and are left with 1 subproblem to solve.
2. Prove that there is **always** an optimal solution to the original problem
  - Which makes the greedy choice → the greedy choice is always **safe**.
3. Demonstrate optimal substructure by showing:
  1. We made the greedy choice, rest is:
    - A subproblem with the property that if we combine an optimal solution to the subproblem with the greedy choice we have made,
    - We arrive at an optimal solution to the original problem.



# Greedy choice property

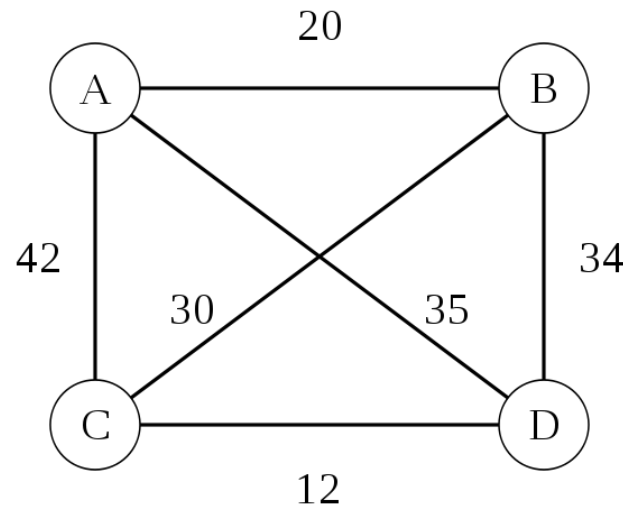
- Goal
  - Assemble a globally optimal solution by making locally optimal (greedy) choices
    - When considering which choice to make
    - → make the choice looking best in the current problem
      - **Without** considering results from subproblems.
- The choice made by a greedy algorithm may depend on choices so far
  - **But** it cannot depend on any future choices or on the solutions to subproblems
  - Progress in a **top-down** fashion
    - Reduce each given problem instance to a smaller one.
- A greedy algorithm makes its first choice before solving any subproblems
- Comparison with dynamic programming (DP)
  - A choice at each step **but** the choice usually depends on the solutions to subproblems.
  - → DP solution in a **bottom-up** manner
    - Progressing from smaller subproblems to larger subproblems.
  - DP: solves the subproblems **before** making the first choice

# Greedy

## ■ Advantage

➤ Used to get an approximation for Hard optimization problems

- Example: Traveling Salesman Problem (NP Hard problem)
  - Given a list of cities and the distances between each pair of cities
  - What is the shortest possible route that visits each city and returns to the origin city?



# Huffman coding compression algorithm

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- Huffman Coding (Huffman Encoding)
  - An algorithm for doing data compression
    - → the basic idea behind file compression.
- Fixed length
  - Every character is stored with a sequence of 0 and 1 using 8 bits
- Variable length encoding
  - It is to design an algorithm that can represent the same piece of text using lesser number of bits.
  - We assign variable number of bits to characters depending on their frequency in the given text. → some character might end up taking 1 bit, some might end up taking 2 bits, ...
  - The problem with variable length encoding lies in its decoding.

# Huffman coding

## ■ Variable Length codes

➤ Suppose the frequency distribution of the characters is:

| A       | B   | C   |
|---------|-----|-----|
| 999,000 | 500 | 500 |

➤ Encode:

| A | B  | C  |
|---|----|----|
| 0 | 10 | 11 |

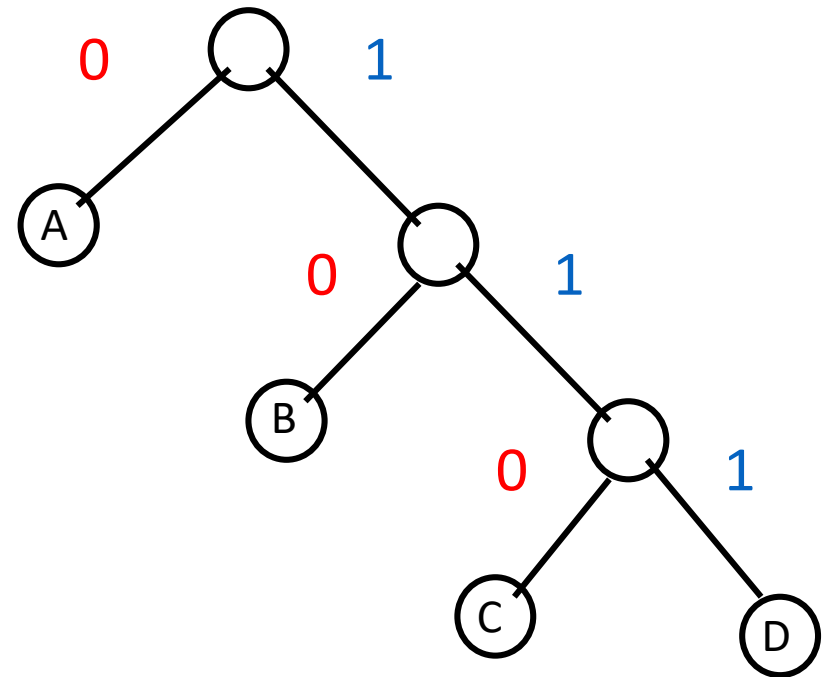
➤ The code of A: length 1, and the codes for B and C: length 2

- Fixed code:  $1,000,000 \times 2 = 2,000,000$
- Variable code:  $999,000 \times 1 + 500 \times 2 + 500 \times 2 = 1,001,000$

# Huffman coding

## ■ Decoding

- In the variable length code: **Prefix code**
  - Where no code is a prefix of another.
- Example: A = 0, B = 10, C = 11
  - None of the above codes is a prefix of another
- Example
  - Input: AAABBBCCCBCBAACC
  - Encoding: 0001010111111101110 0 01111
- Binary tree:
  - 0: left child
  - 1: right child



# Huffman coding

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## ■ Pseudo code

- Consider all pairs: <frequency, symbol>.
- Choose the 2 lowest frequencies:
  1. Create a node, set these pairs as children
  2. The node gets the combined frequency.
- Iterate

# Huffman coding

## ■ Example:

➤ Alphabet: A,B,C,D,E,F

➤ Frequency table:

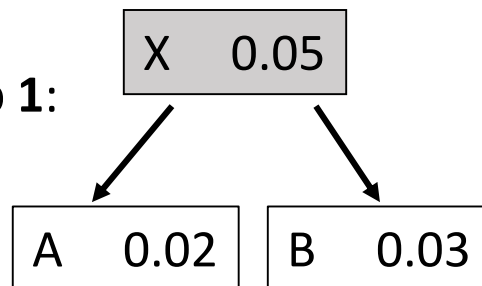
➤ Current values (key,frequency)

| A    | B    | C    | D    | E    | F    |
|------|------|------|------|------|------|
| 0.02 | 0.03 | 0.10 | 0.20 | 0.30 | 0.35 |

○ Step 0:

|        |        |        |        |        |        |
|--------|--------|--------|--------|--------|--------|
| A 0.02 | B 0.03 | C 0.10 | D 0.20 | E 0.30 | F 0.35 |
|--------|--------|--------|--------|--------|--------|

○ Step 1:

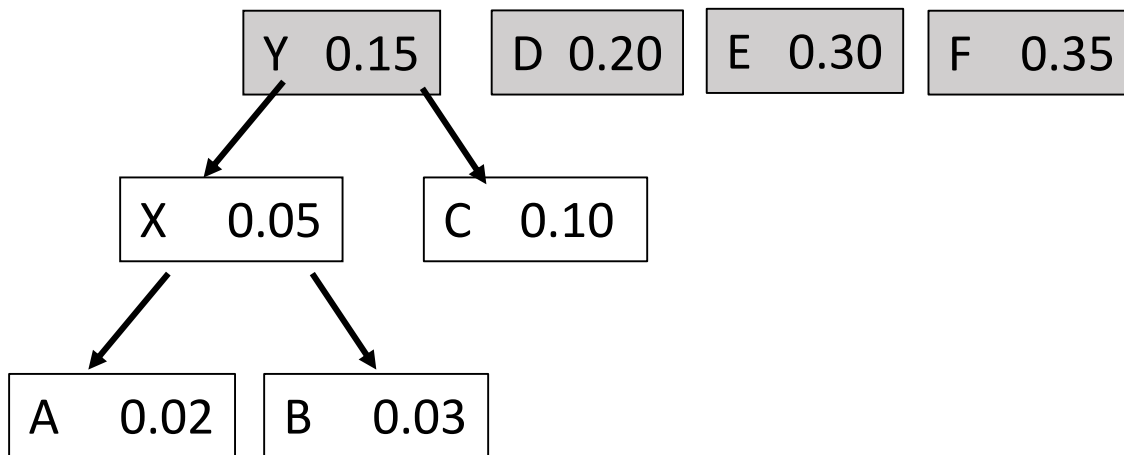


|        |        |        |        |
|--------|--------|--------|--------|
| C 0.10 | D 0.20 | E 0.30 | F 0.35 |
|--------|--------|--------|--------|

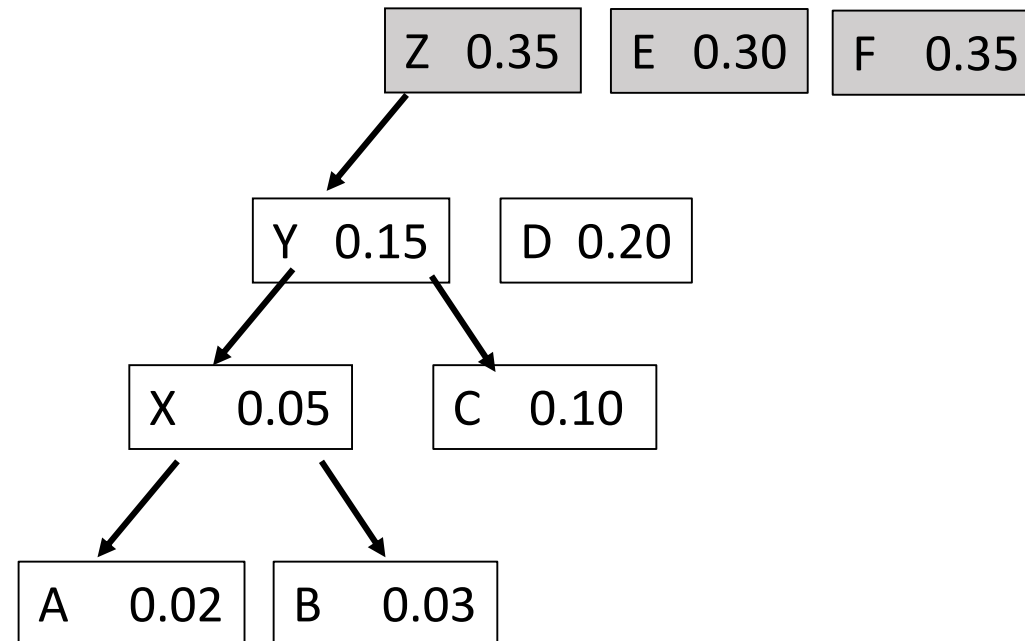
# Huffman coding

## ■ Example

➤ Step 2



Step 3

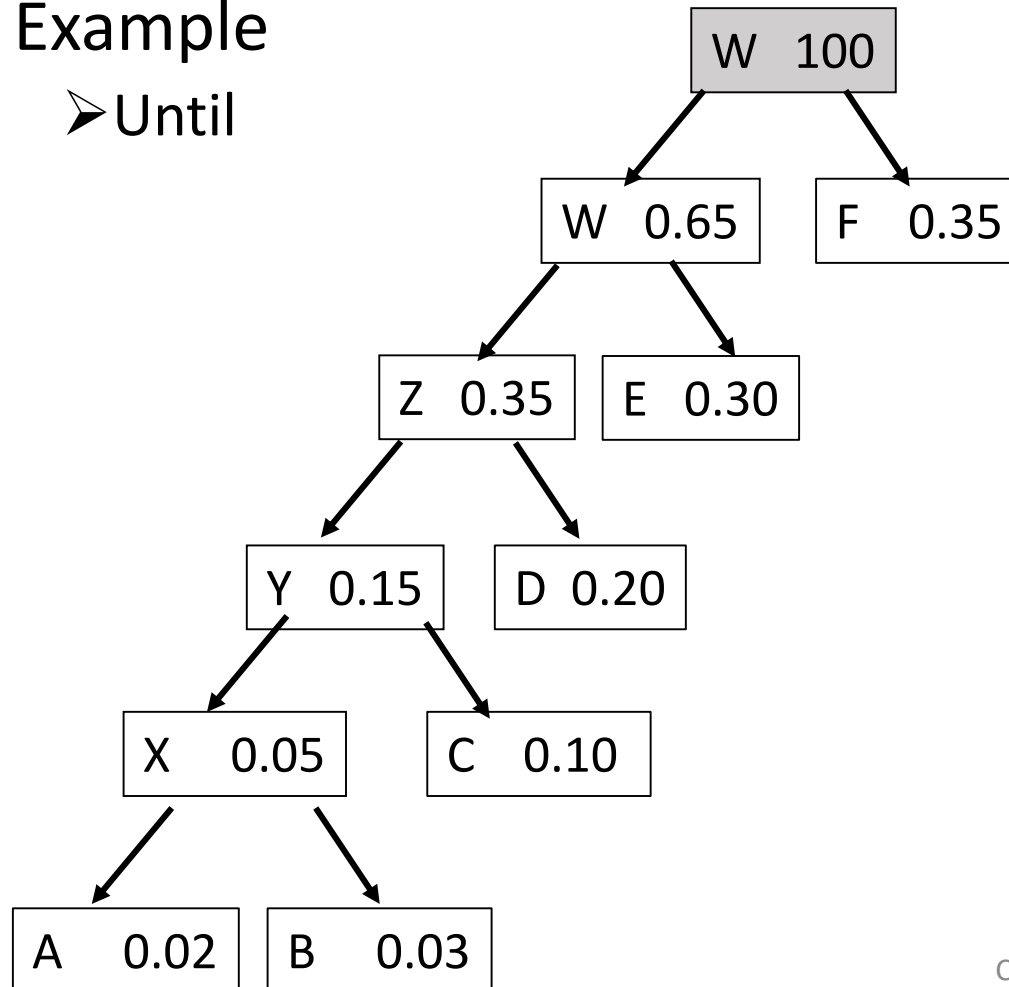




# Huffman coding

- Example

➤ Until



Each priority queue operation (e.g. heap):  $O(\log n)$

In each iteration: 1 less subtree.

Initially:  $n$  subtrees.

Total:  $O(n \log n)$  time.

# Huffman coding

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## ■ Correctness

### ➤ Greedy Choice Property:

- There exists a minimum cost prefix tree where the 2 smallest frequency characters are indeed siblings with the longest path from root.
- → the greedy choice does not “hurt” finding the optimum

### ➤ Optimal Substructure Property:

- An optimal solution to the problem once we choose the 2 least frequent elements
- and combine them to produce a smaller problem,
  - is a solution to the problem when the 2 elements are added.

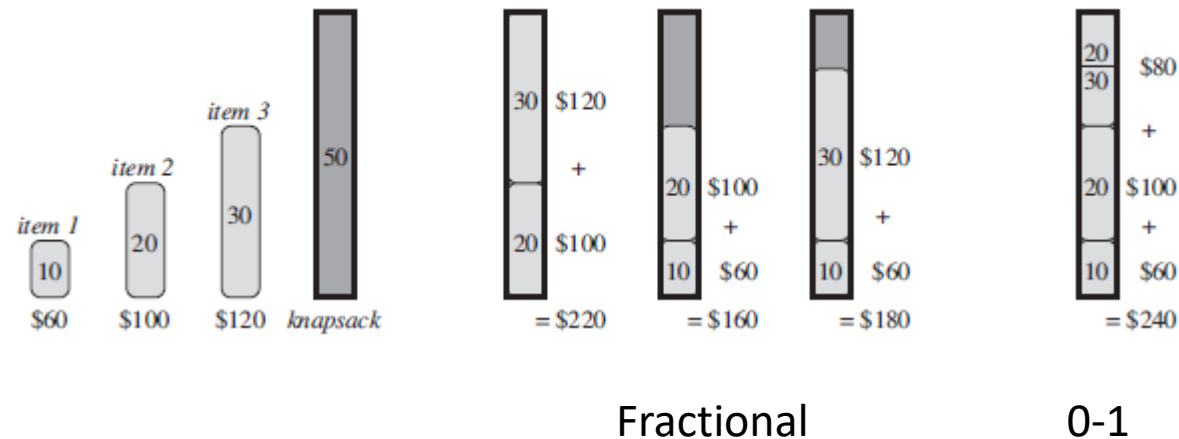
# Applications

- **Kruskal's Minimum Spanning Tree (MST)**
  - Create a MST by picking edges one by one.
  - **Greedy Choice:** pick the **smallest weight edge that doesn't cause a cycle in the MST constructed so far.**
- **Prim's Minimum Spanning Tree**
  - Create a MST by picking edges one by one.
  - Maintain 2 sets: set of the vertices already included in MST and the set of the vertices not yet included.
  - **Greedy Choice:** pick the **smallest weight edge that connects the 2 sets.**
- **Dijkstra's Shortest Path** (similar to Prim's algorithm)
  - The shortest path tree is built up, edge by edge.
  - Maintain 2 sets: set of the vertices already included in the tree and the set of the vertices not yet included.
  - **Greedy Choice:** pick the edge that connects the 2 sets, and is on the smallest weight path from source to the set that contains not yet included vertices.
- **Huffman Coding**
  - lossless compression technique.
  - It assigns variable length bit codes to different characters.
  - **Greedy Choice:** assign **least bit length code** to the **most frequent character.**

# Conclusion

## ■ Greedy

- Short term policy: makes the choice that looks best at the moment.
  - Once it is done, it is done and we are close to the final solution
- They do not always yield optimal solutions,
  - **but** for many problems they do



# Questions ?

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- Reading
  - Canvas Csci 115 book – Chapter 9
  - Introduction to Algorithms, Chapter 16.

