CSCI 232: Data Structures and Algorithms

Greedy Algorithms + Tractability

Reese Pearsall Spring 2024

https://www.cs.montana.edu/pearsall/classes/spring2024/232/main.html



Announcements

Lab 12 due **Sunday** at 11:59 PM



Greedy Algorithms

Technique to solve a problem that involves making the choice the **best helps some objective**

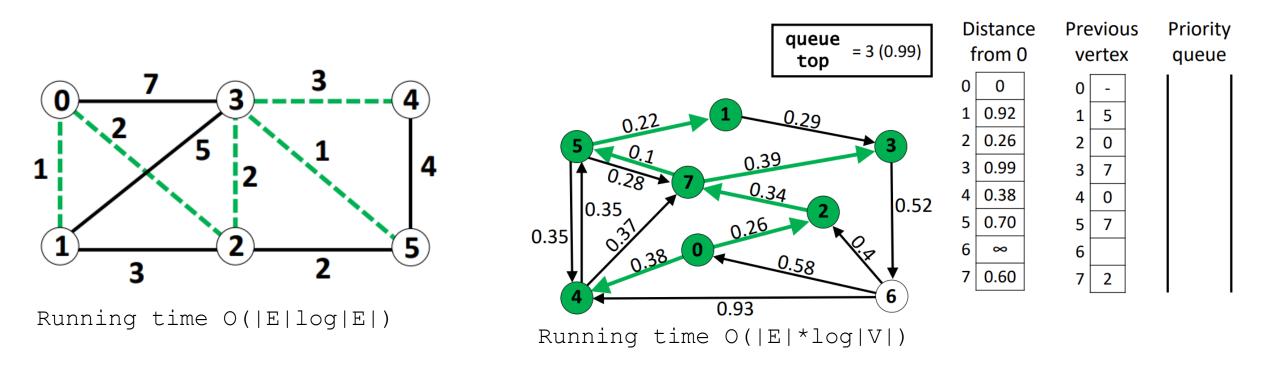
Objective = shortest cost, most profit, spend least money as possible

We (usually) do not look ahead, plan, or revisit past decisions

Hope that optimal local choices lead to optimal global solutions

Sometimes the greedy approach is not the best solution a problem

Greedy Algorithms



Kruskal's and Dijkstra's algorithm are both examples of greedy algorithms

At each step of the algorithm, they attempt to select the edge with the **minimum** cost

(The greedy approach works fine for these, because these algorithms always return the optimal result)

Suppose you pay D dollars to enter a buffer. You can eat only N items before you get full. You know the cost of every item in the buffet



Our goal is to get the most "bang for our buck",

aka. maximize C = (D - S) where S is the sum of the N items we ate at the buffet

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- C = (D S) where S is the sum of the N items
- we ate at the buffet

 Sort items by their value (greatest-to-least)

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Our goal is to get the most "bang for our buck",

aka. maximize

C = (**D** – **S**) where S is the sum of the N items we ate at the buffet

$$N = 3$$
, $S = 94 , $D = 40 , $C = 54

- Sort items by their value (greatest-to-least)
- 2. Select the first N items in the list

You are a thief with a knapsack that can hold up to **N** weight. You are robbing a store, where each item has a weight, and a value

Goal: Maximize value of items being stolen, and don't overfill knapsack



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Suppose our knapsack can only hold 10 pounds



Knapsack (10)

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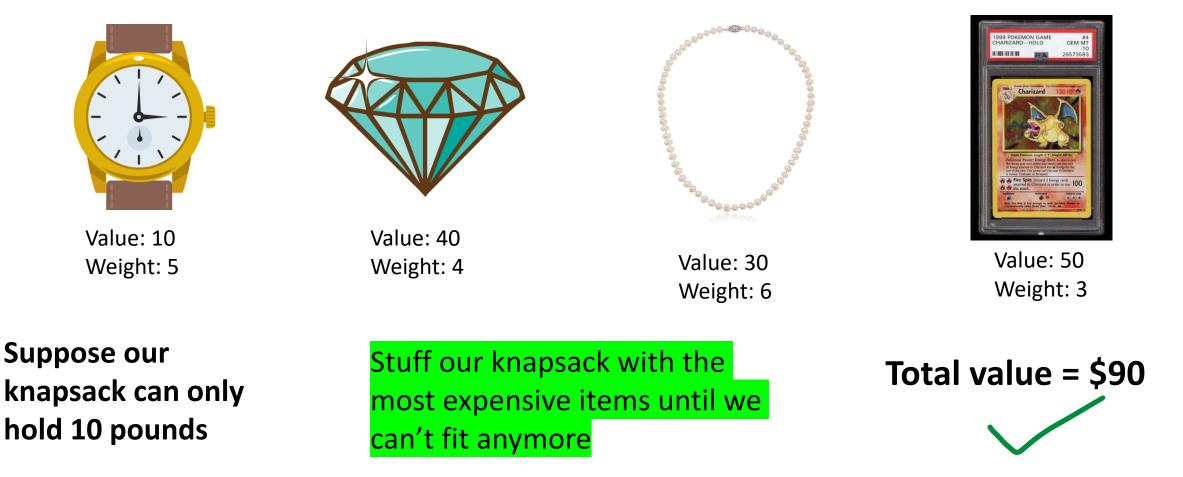
Knapsack (0)

Suppose our knapsack can only hold 10 pounds

Stuff our knapsack with the most expensive items until we can't fit anymore

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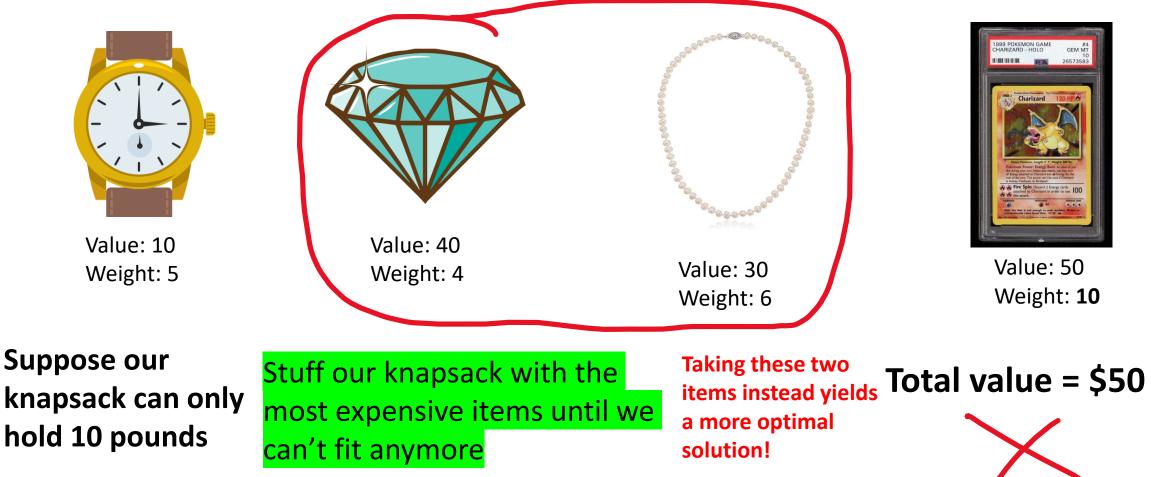
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Value: 50 Weight: **10**

Suppose our knapsack can only hold 10 pounds

Compute the **ratio** of value/weight, and select items based on that

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Value: 40 Weight: 4 Ratio: 10

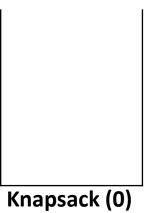


Ratio: 5



Ratio: 2

1. Sort items based on ratio



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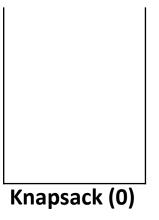
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Value: 10

1. Sort items based on ratio

- 2. Add items to knapsack if they will not exceed the knapsack
- 3. Repeat step 2 until we've checked every item



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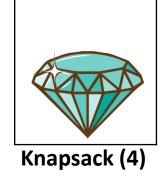
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We cannot select this item, because it will exceed the knapsack

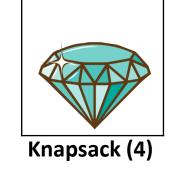
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Compute the **ratio** of value/weight, and select items based on that

Total profit of knapsack: \$40 + \$30 = **\$70**

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Value: 5.5 Weight: 4 **Ratio: 1.38**



Ratio: 1.33

Let N = 6



- 1. Sort items based on ratio
- Add items to knapsack if they will not exceed the knapsack
- Repeat step 2 until we've checked every item

Compute the **ratio** of value/weight, and select items based on that

Given these new prices, weights, and ratios, will our algorithm still work?

You are a thief with a knapsack that can hold up to **N** weight. You are robbing a store, where each item has a weight, and a value

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We can't add these items, because they would exceed our knapsack capacity

- 1. Sort items based on ratio
- 2. Add items to knapsack if they will not exceed the knapsack
- 3. Repeat step 2 until we've checked every item

Compute the **ratio** of value/weight, and select items based on that

Value: 5.5

Weight: 4

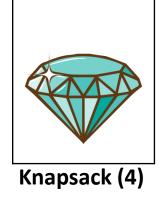
Ratio: 1.38

Let N = 6

Ratio: 1.33

Given these new prices, weights, and ratios, will our algorithm still work? Total profit = 5.5

27



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Compute the **ratio** of value/weight, and select items based on that

Given these new prices, weights, and ratios, will our algorithm still work? **Total profit = 5.5 Optimal solution= 8**

You a st

This is the 0/1 knapsack problem, which means that we either take the item, or we don't. We can't take "half" of a watch to fill the remaining empty space of our knapsack '

The greedy approach **does not** always yield the optimal solution 🛞



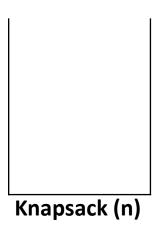
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Fractional variant = we can take a fraction of an item





Knapsack capacity: 35



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1. Sort items based on ratio



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- 2. Iterate through sorted list
 - 1. If adding item will not exceed the capacity, add it
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Knapsack (1)



- 1. Sort items based on ratio
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Value: 20Value: 40Weight: 1Weight: 5Ratio: 20Ratio: 8

Knapsack capacity: 35

: 40 Value: 88 ht: 5 Weight: 12 : 8 Ratio: **7.33**

Value: 52

Weight: 8

Ratio: 6.5



Value: 100 Weight: 20 **Ratio: 5** Value: 13 Weight: 8 **Ratio: 1.625**



Value: 13 Weight: 10: **Ratio 1.3**



- 1. Sort items based on ratio
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Knapsack (6)





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Value: 13 Weight: 10: **Ratio 1.3**



Knapsack (18)

- 1. Sort items based on ratio
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Knapsack capacity: 35

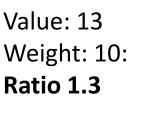


- 10000 10000
- Value: 100 Weight: 20 **Ratio: 5**



Value: 13 Weight: 8 **Ratio: 1.625**







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Knapsack (26)





Value: 20Value: 40Value: 88Weight: 1Weight: 5Weight: 12Ratio: 20Ratio: 8Ratio: 7.33

Knapsack capacity: 35



Value: 52

Weight: 8

Ratio: 6.5



Value: 100 Weight: 20 **Ratio: 5** Value: 13 Weight: 8 Ratio: 1.625



Value: 13 Weight: 10: **Ratio 1.3**



Knapsack (26)

- 1. Sort items based on ratio
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Value: 20Value: 40Value: 88Weight: 1Weight: 5Weight: 12Ratio: 20Ratio: 8Ratio: 7.33

Knapsack capacity: 35



Value: 52

Weight: 8

Ratio: 6.5



Value: 100 Weight: 20 **Ratio: 5**



Value: 13 Weight: 8 Ratio: 1.625



Value: 13 Weight: 10: **Ratio 1.3**



- 1. Sort items based on ratio
- 2. Iterate through sorted list
 - If adding item will not exceed the capacity, add it
 - If adding item will exceed the capacity, take a fraction of it to fill the remaining space of knapsack

We cannot take the full 20 pounds of the gold bar, but **we can** take a fraction of it

Cut off 9 pounds of the gold bar, and place it in out knapsack

Knapsack (26)



- 1. Sort items based on ratio
- 2. Iterate through sorted list
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Knapsack (35)





Value: 88

Weight: 12

Ratio: **7.33**

Value: 52

Weight: 8

Ratio: 6.5



Knapsack capacity: 35

- 1. Sort items based on ratio
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Pearl: 20 Emerald: 40 Silver: 88 Sapphire: 52 9/20 of a gold bar: ???

Value: 100

Weight: 20

Ratio: 5

Value: 13

Weight: 8

Ratio: 1.625



Value: 13 Weight: 10: **Ratio 1.3**



Knapsack (35)





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Value: 13 Weight: 10: **Ratio 1.3**



Knapsack (35)





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Value: 13 Weight: 10: **Ratio 1.3**





Knapsack (35)



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2.

In the **fractional** knapsack problem, the greedy approach **will** guarantee an optimal solution



Knapsack (35)

of it to fill the remaining space of knapsack

Total profit: \$245



Customer are constantly arriving to a check out line. Walmart is open 24/7

Instead of first-in-first-out, you want to serve **as many customers as possible** in your 2-hour shift





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Instead of first-in-first-out, you want to serve **as many customers as possible** in your 2-hour shift

You can assume that there will always be several waiting in line, and you know how many minutes it will take to check them out



Any ideas to achieve our goal?



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15 min

8 min



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11 min



Select the customer that would take the least time

And repeat this until your shift is over!



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11 min



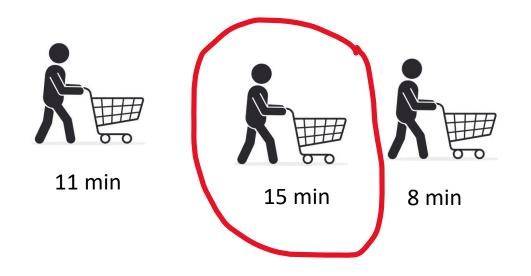
8 min

Select the customer that would take the least time

Is this algorithm good?

15 min





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You can assume that there will always be several waiting in line, and you know how many minutes it will take to check them out

Select the customer that would take the least time

Optimal, but not Fair!

This customer has a longer service time, and they may potentially wait **a very long** time until they are served

Being a Cashier at Walmart CPU Job Scheduling



Image: Non-SecondsImage: Non-Seconds

This problem is very relevant in the world of **operating systems**.

There are many processes/software running on your computer all at the same time

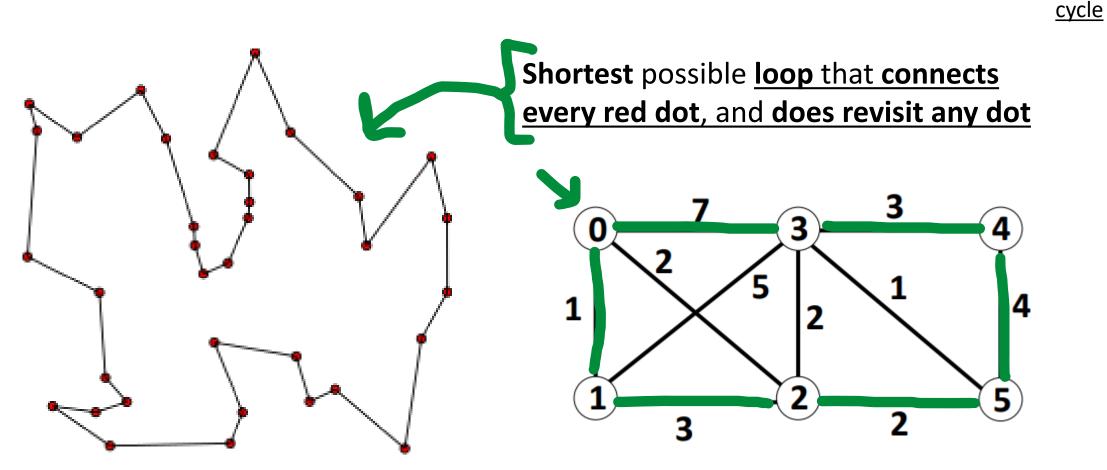
Each process needs to use the hardware on the computer to do its job.

OS oversees selecting which job will be processed by the CPU next

Ideally, we want a fair approach 🙂

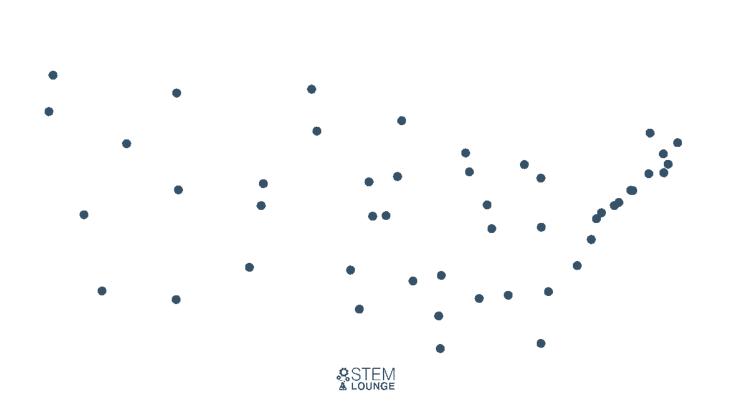
Traveling Salesman

Given a graph with edge weights and a starting node, what is the **shortest** path that will visit every node, and start and end at the starting node, without visiting the same node twice



Hamiltonian

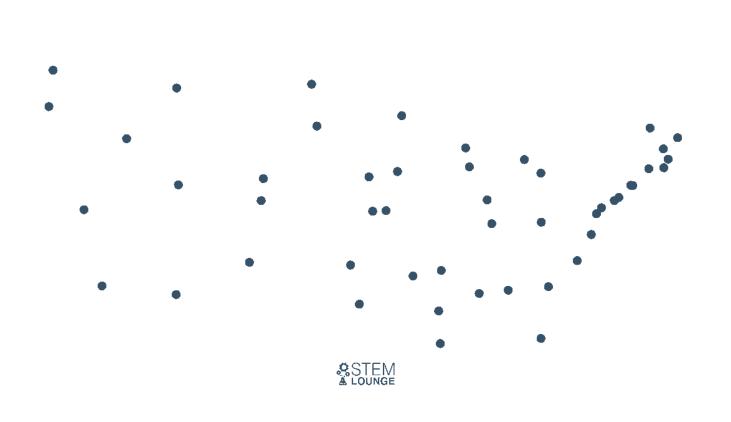
Given a graph with edge weights and a starting node, what is the **shortest** path that will visit every node, and start and end at the starting node, without visiting the same node twice



Given the nodes of major cities in the US, what would the greedy algorithm look like for the TSP problem?

You can assume every node as a direct path to every other node

Given a graph with edge weights and a starting node, what is the **shortest** path that will visit every node, and start and end at the starting node, without visiting the same node twice

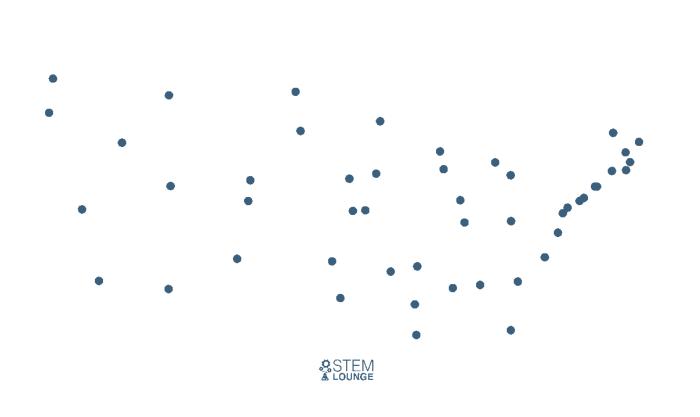


Given the nodes of major cities in the US, what would the greedy algorithm look like for the TSP problem?

Nearest neighbor: always travel to the nearest unvisited neighbor, and then travel to their nearest neighbor, and then travel to their nearest neighbor...

Once we have visited all nodes, travel back to starting node

Given a graph with edge weights and a starting node, what is the **shortest** path that will visit every node, and start and end at the starting node, without visiting the same node twice

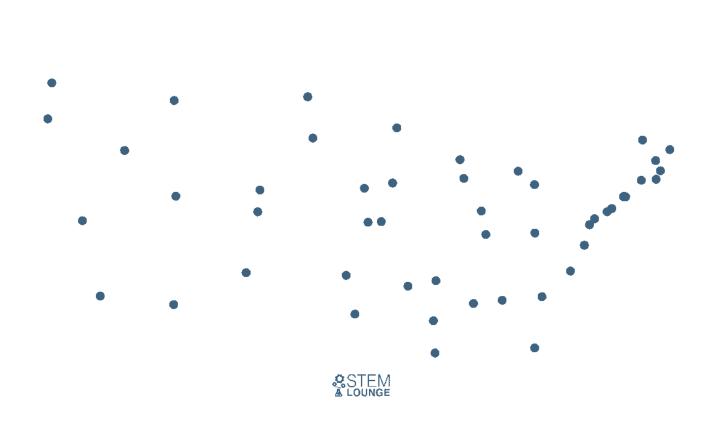


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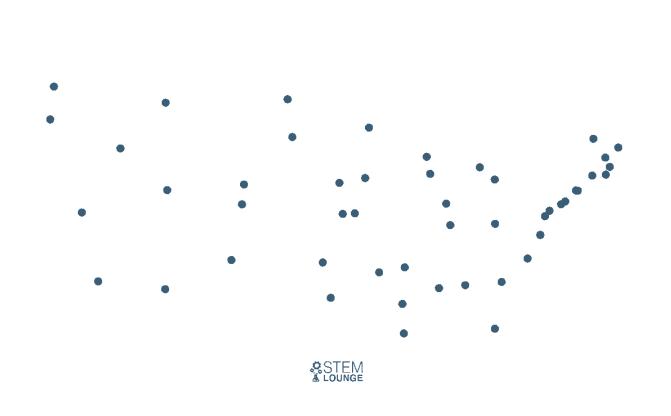
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Given the nodes of major cities in the US, what would the greedy algorithm look like for the TSP problem?

Lowest edge cost: add the shortest edge that will neither create a vertex with more than 2 edges, nor a cycle with less than the total number of cities until we have a cycle

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Given the nodes of major cities in the US, what would the greedy algorithm look like for the TSP problem?

Nearest Insertion

Start with a cycle, keep growing the cycle by adding the city nearest to the cycle

Given a graph with edge weights and a starting node, what is the **shortest** path that **will visit every node**, and **start and end at the starting node**, **without visiting the same node twice**

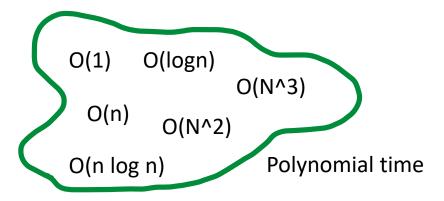
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The traveling salesman problem is a difficult problem... in fact, it is one of the **most difficult** problems in computer science

We do not know of an algorithm that can solve TSP in **polynomial time**

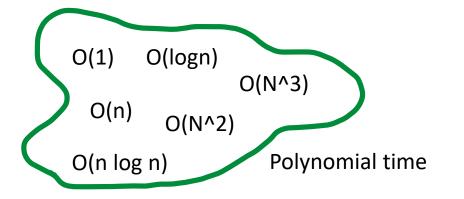


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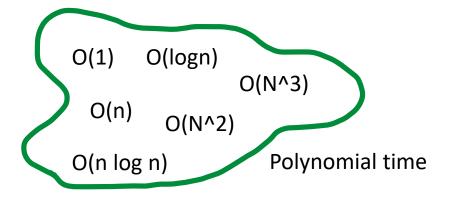
The algorithms we currently have for solving TSP run in **exponential** or **factorial** time, which are infeasible for large input sizes

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If you can solve TSP in polynomial time, you will become a millionaire (literally)

Tractability refers to the problems we can solve and not solve in polynomial time

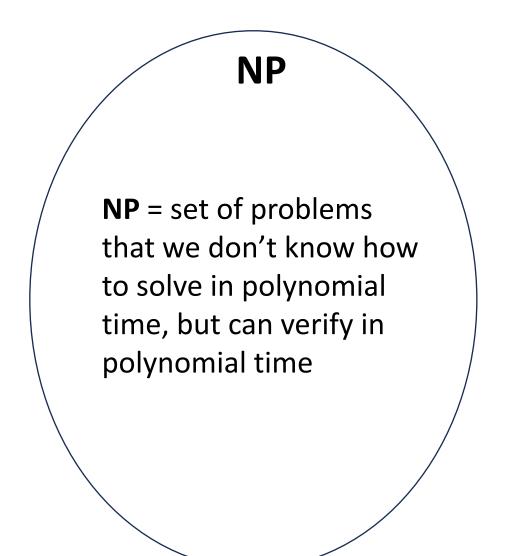
sorting Shortest path Fractional knapsack Closest pair of points Rod cutting Change making and much more

Ρ

P is the set of all problems that we can solve in polynomial time

Tractability refers to the problems we can solve and not solve in polynomial time

Ρ sorting Shortest path Fractional knapsack Closest pair of points Rod cutting Change making and **much** more



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Verify = given a solution to problem, verify if it is correct/incorrect

NP = set of problems that we don't know how to solve in polynomial time, but can verify in polynomial time

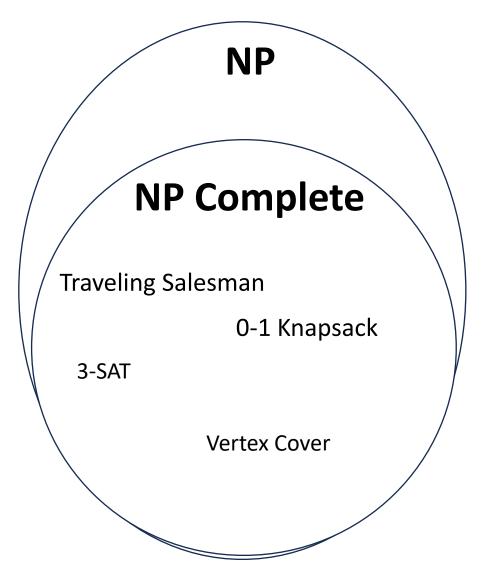
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NP-Complete-

The hardest problems of NP. If we can solve one NP-Complete problem, we can solve all other NP-Complete problems

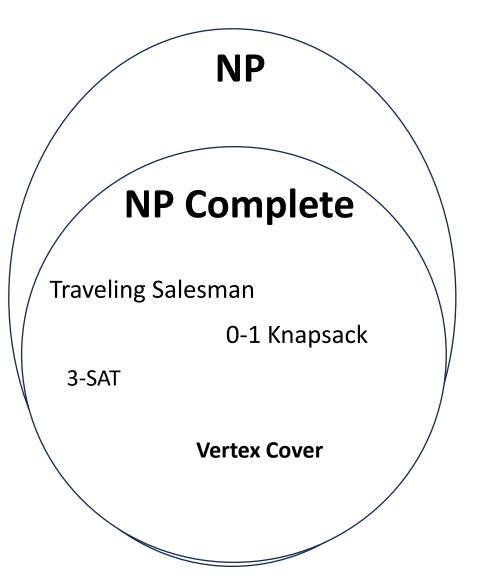


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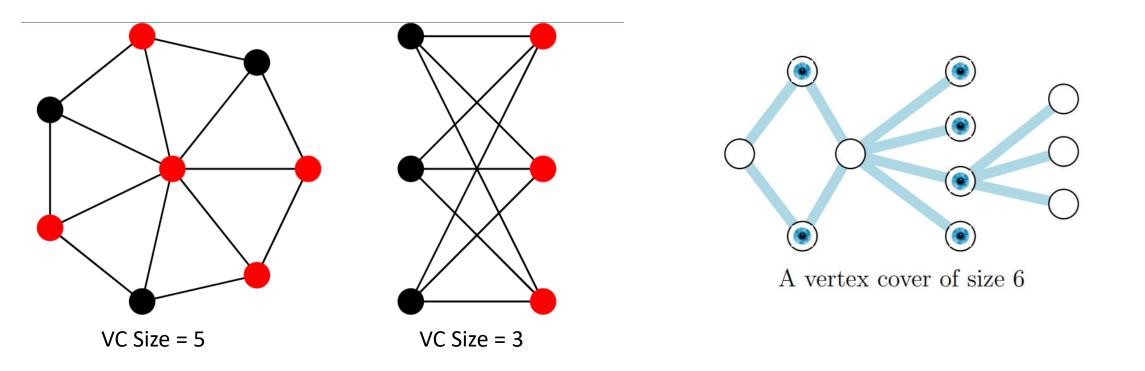
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Vertex Cover = Given a graph, compute a set of vertices S that include at least one endpoint of every edge.

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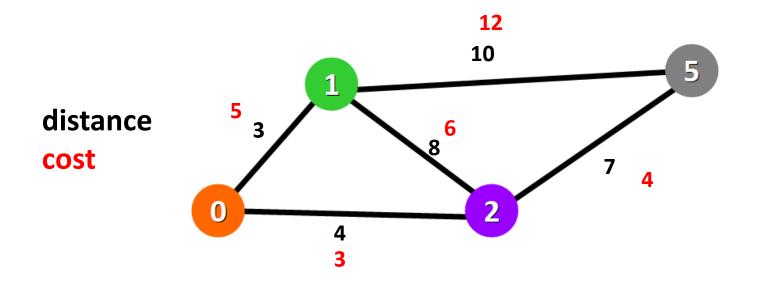


Typically, we are concerned about finding the *minimum* vertex cover

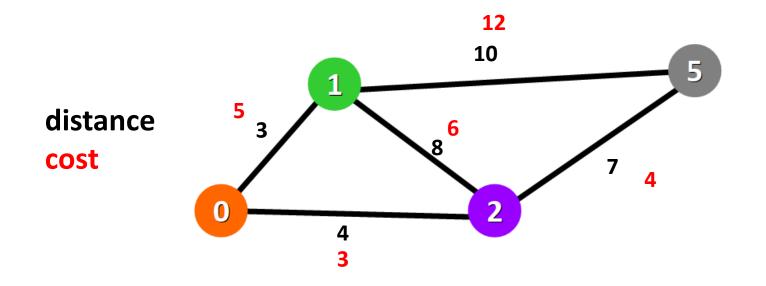
Finding shortest paths in a distance-weighted graph can be done in polynomial time.

Dijkstra's Algorithm

Finding shortest paths that cost at most 20 in a distance-and-cost-weighted graph can be done in polynomial time.



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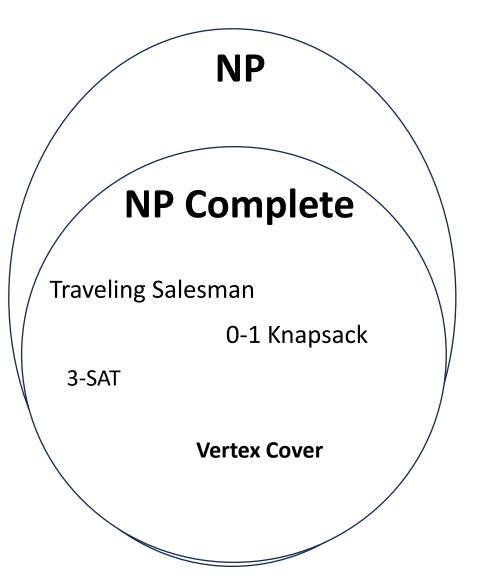
By making a small tweak to this problem, this problem actually becomes *much* more challenging (NP-C)

Tractability refers to the problems we can solve and not solve in polynomial time

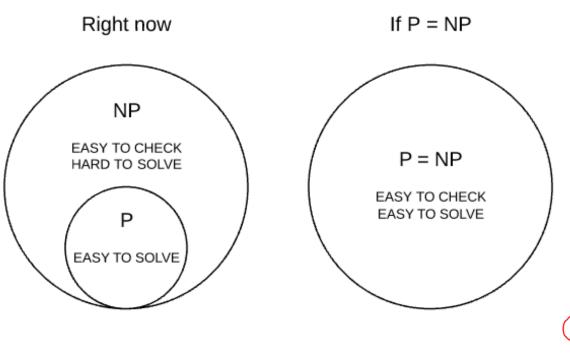
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On May 24, 2000, <u>Clay Mathematics Institute</u> came up with seven mathematical problems, for which, the solution for any of the problem will earn US \$1,000,000 reward for the solver. Famously known as the Millennium Problems, so far, only one of the seven problems is solved till date.

Wanna make a million dollar, try solving one from <u>this list</u>. These are the problems listed for a million dollar prize reward.

1. Yang–Mills and Mass Gap

2. Riemann Hypothesis

3. P vs NP Problem

4. Navier–Stokes Equation

If you can solve TSP in polynomial time, you can win a million dollars, probably become a tenured CS professor, and also break cybersecurity

