GREEDY ALGORITHMS

Assume that you have the coins below.

States are
$$d_1 = 25$$
 (quarter), $d_2 = 10$ (dime), $d_3 = 5$ (nickel), and $d_4 = 1$ (penny).

 How would you give change with coins of these denominations of, say, 48 cents?

- If you came up with the answer 1 quarter, 2 dimes, and 3 pennies, you followed—consciously or not—a logical strategy of making a sequence of best choices among the currently available alternatives. Indeed, in the first step, you could have given one coin of any of the four denominations.
- "Greedy" thinking leads to giving one quarter because it reduces the remaining amount the most, namely, to 23 cents.
- In the second step, you had the same coins at your disposal, but you could not give a quarter, because it would have violated the problem's constraints.
- So your best selection in this step was one dime, reducing the remaining amount to 13 cents. Giving one more dime left you with 3 cents to be given with three pennies.

What is greedy?

- The approach applied in the opening paragraph to the change-making problem is called greedy.
- Computer scientists consider it a general design technique despite the fact that it is applicable to optimization problems only.
- The greedy approach suggests constructing a solution through a sequence of steps, each expanding a partially constructed solution obtained so far, until a complete solution to the problem is reached.
 On each step—and this is the central point of this technique—the choice made must be:

What is greedy?

- feasible, i.e., it has to satisfy the problem's constraints
- locally optimal, i.e., it has to be the best local choice among all feasible choices available on that step
- irrevocable, i.e., once made, it cannot be changed on subsequent steps of the algorithm

What is greedy?

- These requirements explain the technique's name: on each step, it suggests a "greedy" grab of the best alternative available in the hope that a sequence of locally optimal choices will yield a (globally) optimal solution to the entire problem.
- We refrain from a philosophical discussion of whether greed is good or bad. (If you have not seen the movie from which the chapter's epigraph is taken, its hero did not end up well.)
- From our algorithmic perspective, the question is whether such a greedy strategy works or not.
- As we shall see, there are problems for which a sequence of locally optimal choices does yield an optimal solution for every instance of the problem in question.
- However, there are others for which this is not the case; for such problems, a greedy algorithm can still be of value if we are interested in or have to be satisfied with an approximate solution.

- If you have 20, 19, 5, 1 coins and you try to have 24 how can you do that?
- Greedy algorithms always try to approach the answer mostly so it firstly will choose 20 coin.
- After that 4 remaining and the algorithms complete these 4 with 4 pennies.
- So it becomes totally 5 coins.
- \bullet 20 + 1 + 1 + 1 + 1

- But there is a better colution such as 19 + 5
- It becomes only two coins.

- If you have 10, 9 and 1 coins and you want to pay 37 cents, what are the solutions?
- Greedy solutions choose the most valuable coin first of all. It is 10 cent.
- We pay 1 10 cent and 37-10 = 27 is remaining
- Greedy algorithm choose again 10 for paying 27

- So we pay 2 10 cent and 37 10 10 = 17 is remaining
- Greedy algorithm choose again 10 for paying 17

- So we pay 3 10 cent and 37 10 10 10 = 7 is remaining
- Greedy algorithm choose 1 cent for paying 7
- We pay 7 1 cents for remaining 7.

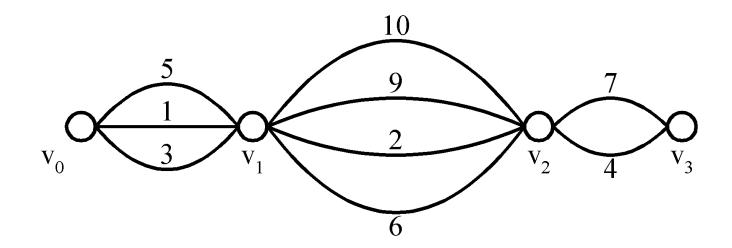
- So the greedy algorithm solution is such as:
- There are 10 coins.

- Is there a more efficient solution?
- Yes there is.
- 10 + 9 + 9 + 9 = 37
- You can pay 37 cents with only 4 coins.

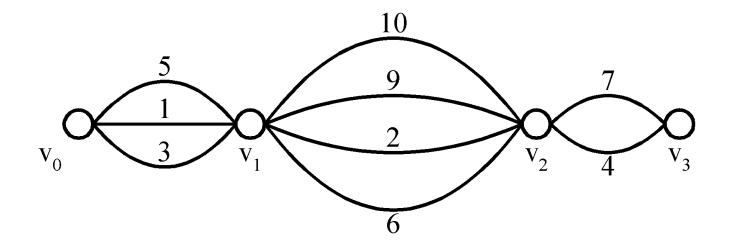
- Another question?
- If you have coins for 1, 4, 5 and 10 cent coins and you need to pay 8 cents, how can you pay?

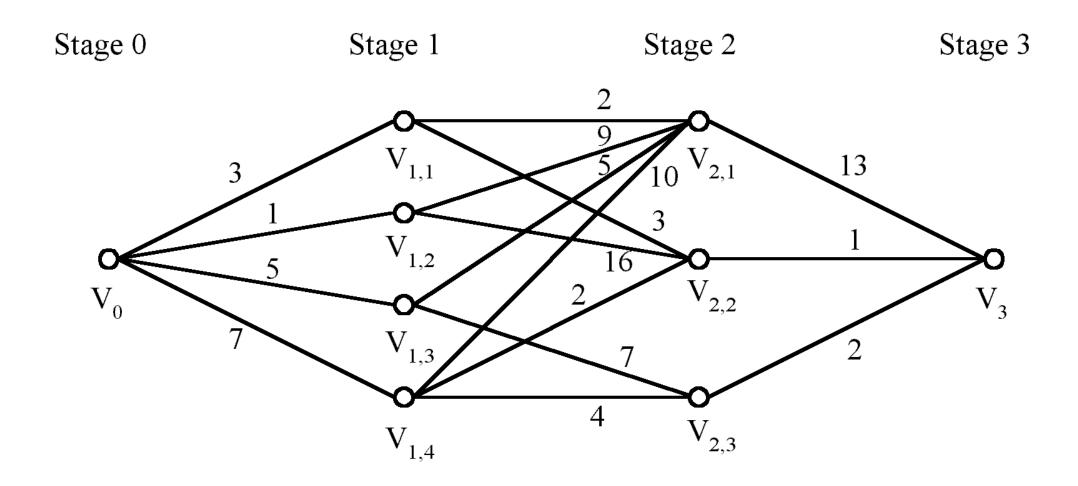
- First of all, Greedy algorithm chooses the mos valuable coin which is less then 8.
- It is 5.
- So 8 5 = 3 remaining.
- The algorithm makes 3 1 cents for 3 cents.

- So the solution is
- \bullet 5 + 1 + 1 + 1 = 8
- The solution has 4 coins.
- But if we write 4 + 4 = 8 cents, the problem is solved with only 2 coins.

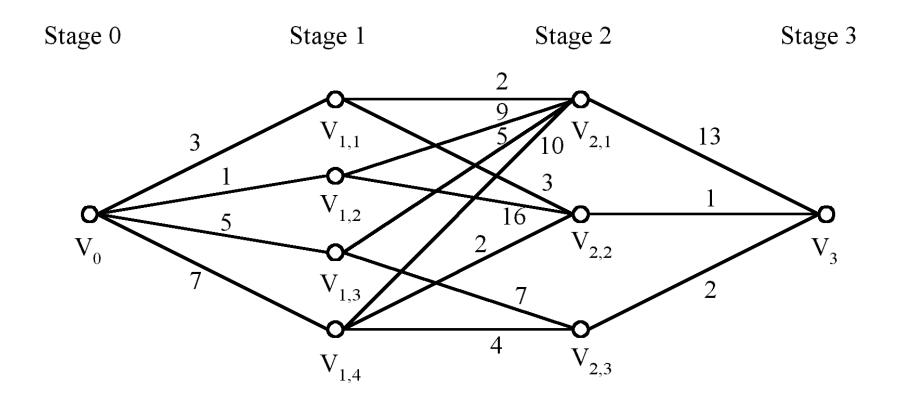


- What is the shortest path between v0 and v3 in this graph?
- Can you find it with greedy algorithms?





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Change-Making Problem - GENERAL

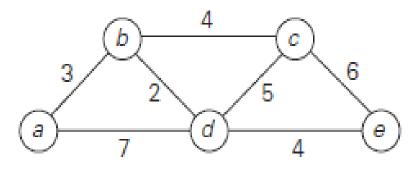
• Given a value N, if we want to make change for N cents, and we have infinite supply of each of S = { S1, S2, ..., Sm} valued coins, how many ways can we make the change? The order of coins doesn't matter.

• For example, for N = 4 and S = {1,2,3}, there are four solutions: {1,1,1,1},{1,1,2},{2,2},{1,3}. So output should be 4. For N = 10 and S = {2, 5, 3, 6}, there are five solutions: {2,2,2,2,2}, {2,2,3,3}, {2,2,6}, {2,3,5} and {5,5}. So the output should be 5.

Change-Making Problem - GENERAL

- To count total number solutions, we can divide all set solutions in two sets.
 - 1) Solutions that do not contain mth coin (or Sm).
 - 2) Solutions that contain at least one Sm. Let count(S[], m, n) be the function to count the number of solutions, then it can be written as sum of count(S[], m-1, n) and count(S[], m, n-Sm).
- Therefore, the problem has optimal substructure property as the problem can be solved using solutions to subproblems.

```
ALGORITHM Dijkstra(G, s)
    //Dijkstra's algorithm for single-source shortest paths
    //Input: A weighted connected graph G = \langle V, E \rangle with nonnegative weights
              and its vertex s
    //Output: The length d_v of a shortest path from s to v
                and its penultimate vertex p_v for every vertex v in V
    Initialize(Q) //initialize priority queue to empty
    for every vertex v in V
         d_v \leftarrow \infty; p_v \leftarrow \text{null}
         Insert(Q, v, d_v) //initialize vertex priority in the priority queue
    d_s \leftarrow 0; Decrease(Q, s, d_s) //update priority of s with d_s
    V_T \leftarrow \varnothing
    for i \leftarrow 0 to |V| - 1 do
         u^* \leftarrow DeleteMin(Q) //delete the minimum priority element
         V_T \leftarrow V_T \cup \{u^*\}
         for every vertex u in V - V_T that is adjacent to u^* do
              if d_{u*} + w(u^*, u) < d_u
                   d_u \leftarrow d_{u^*} + w(u^*, u); \quad p_u \leftarrow u^*
                   Decrease(O, u, d_u)
```



Tree vertices	Remaining vertices	Illustration
a(-, 0)	$b(a,3)\ c(-,\infty)\ d(a,7)\ e(-,\infty)$	3 2 d 5 6 e
b(a, 3)	$c(b, 3+4) d(b, 3+2) e(-, \infty)$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
d(b, 5)	c(b, 7) e(d, 5+4)	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
c(b, 7)	e(d, 9)	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$
e(d, 9)		

The shortest paths (identified by following nonnumeric labels backward from a destination vertex in the left column to the source) and their lengths (given by numeric labels of the tree vertices) are as follows:

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from a to b: a-b of length 3
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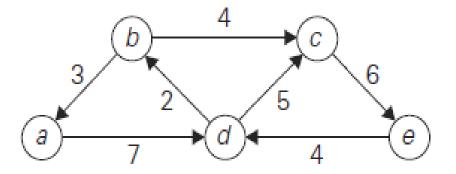
from a to d:
$$a - b - d$$
 of length 5

from
$$a$$
 to c : $a-b-c$ of length 7

from a to e: a-b-d-e of length 9

Exercises

• Find the shortest path in the figure starting from a and ending with another letter.



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