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7.2.1 Single Source Shortest Paths Problem: Dijkstra's Algorithm

REF.

E.W. Dijkstra. A note on two problems in connection with graphs. *Numerische Mathematik*, Volume 1, pp. 269-271, 1959.

- Greedy algorithm
- It works by maintaining a set S of "special" vertices whose shortest distance from the source is already known. At each step, a "non-special" vertex is absorbed into S .
- The absorption of an element of $V - S$ into S is done by a greedy strategy.
- The following provides the steps of the algorithm.

Let

$V = \{1, 2, \dots, n\}$ and source = 1 (7.1)

$C[i, j] =$ Cost of the arc (i, j) if the arc (i, j) exists; otherwise ∞

```
{
    S = { 1 };
    for (i = 2; i < n; i++)
        D[i] = C[1, i];
    for (i = 1; i <= n-1; i++)
    {
        choose a vertex w ∈ V-S such that D[w] is a minimum;

        S = S ∪ {w };

        for each vertex v ∈ V-S

            D[v] = min (D[v], D[w] + C[w, v])
    }
}
```

- The above algorithm gives the costs of the shortest paths from source vertex to every other vertex.
- The actual shortest paths can also be constructed by modifying the above algorithm.

Theorem: Dijkstra's algorithm finds the shortest paths from a single source to all other nodes of a weighted digraph with positive weights.

Proof: Let $V = 1, 2, \dots, n$ and 1 be the source vertex. We use mathematical induction to show that

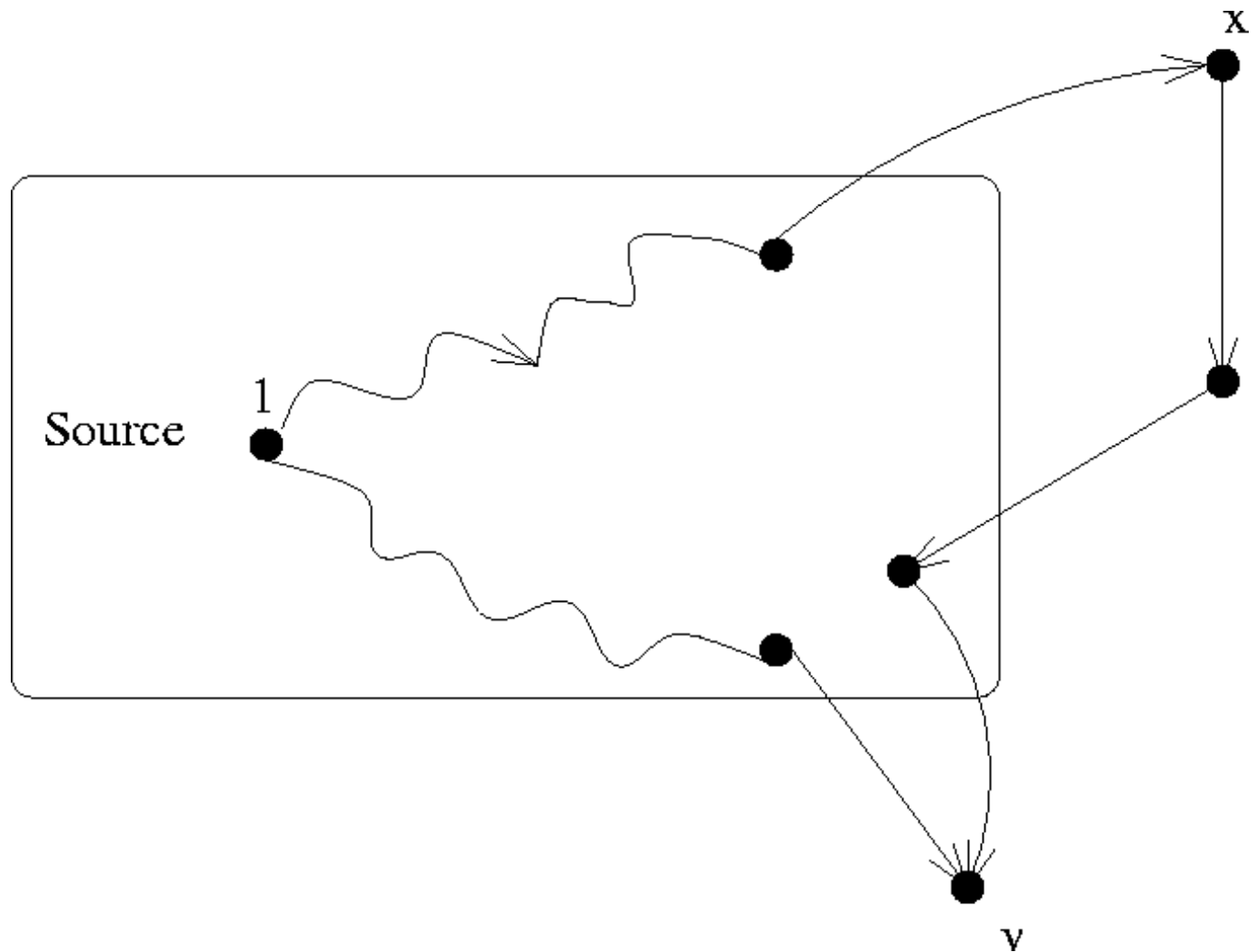
- (a) If a node $i (\neq 1) \in S$, then $D[i]$ gives the length of the shortest path from the source to i .
- (b) if a node $i \notin S$, then $D[i]$ gives the length of the shortest special path from the source to i .

Basis: Initially $S = 1$ and hence (a) is vacuously true. For $i \in S$, the only special path from the source is the direct edge if present from source to i and D is initialized accordingly. Hence (b) is also true.

Induction for condition (a)

- The induction hypothesis is that both (a) and (b) hold just before we add a new vertex v to S .
- For every node already in S before adding v , nothing changes, so condition (a) is still true.
- We have to only show (a) for v which is just added to S .

Figure 7.4: The shortest path to v cannot visit x



- Before adding it to S , we must check that $D[v]$ gives the length of the shortest path from source to v . By the induction hypothesis, $D[v]$ certainly gives the length of the shortest special path. We therefore have to verify that the shortest path from the source to v does not pass through any nodes that do not belong to S .
- Suppose to the contrary. That is, suppose that when we follow the shortest path from source to v , we encounter nodes not belonging to S . Let x be the first such node encountered (see Figure 7.4). The initial segment of the path from source to x is a special path and by part (b) of the induction hypothesis, the length of this path is $D[x]$. Since edge weights are non-negative, the total distance to v via x is greater than or equal to $D[x]$. However since the algorithm has chosen v ahead of x , $D[x] \geq D[v]$. Thus the path via x cannot be shorter than the shortest special path leading to v .

Induction step for condition (b): Let $w \neq v$ and $w \in S$. When v is added to S , these are two possibilities for the shortest special path from source to w :

1. It remains as before
2. It now passes through v (and possibly other nodes in S)

In the first case, there is nothing to prove. In the second case, let y be the last node of S visited before arriving at w . The length of such a path is $D[y] + C[y, w]$.

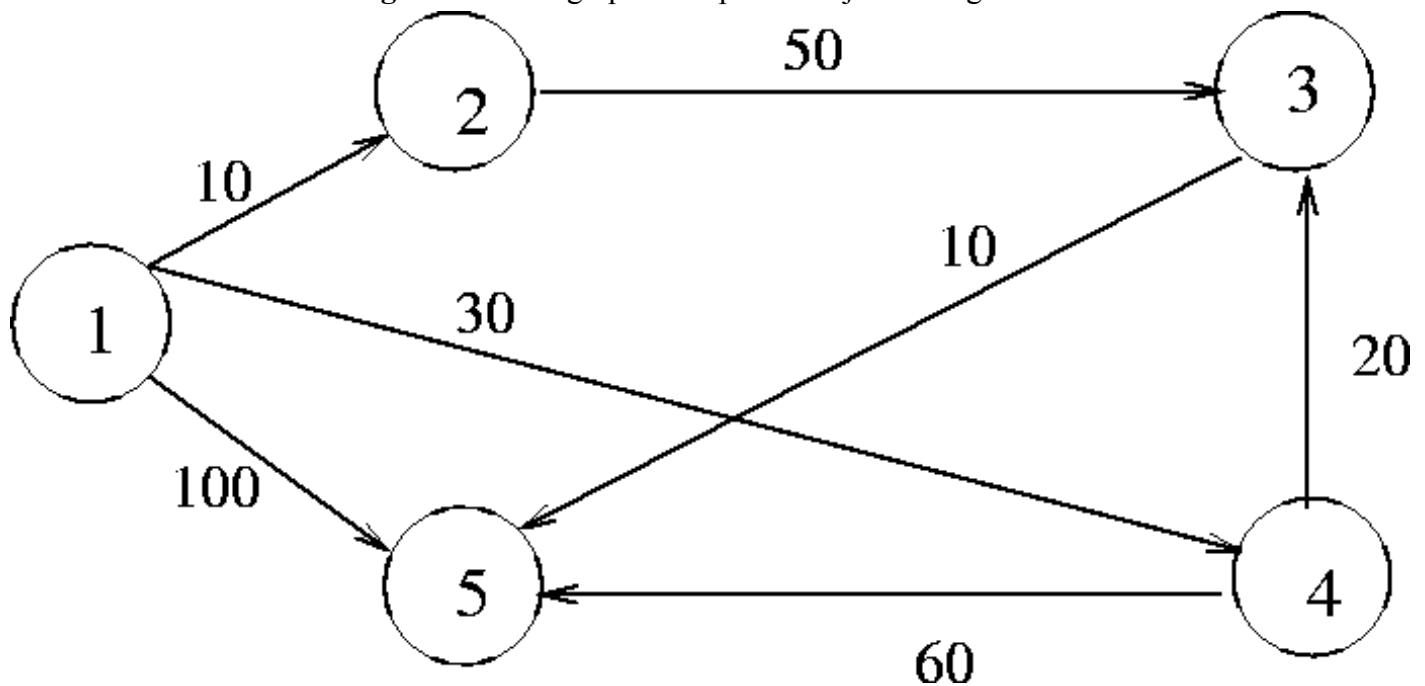
- At first glance, to compute the new value of $d[w]$, it looks as if we should compare the old value of $D[w]$ with $D[y] + C[y,w]$ for every $y \in S$ (including v)
- This comparison was however made for all $y \in S$ except v , when y was added to S in the algorithm.

Thus the new value of $D[w]$ can be computed simply by comparing the old value with $D[v] + C[v,w]$. This the algorithm does.

When the algorithm stops, all the nodes but one are in S and it is clear that the vector $D[1], D[2], \dots, D[n]$ contains the lengths of the shortest paths from source to respective vertices.

Example: Consider the digraph in Figure 7.5.

Figure 7.5: A digraph example for Dijkstra's algorithm



Initially:

$$S = \{1\} \quad D[2] = 10 \quad D[3] = \infty \quad D[4] = 30 \quad D[5] = 100$$

Iteration 1

Select $w = 2$, so that $S = \{1, 2\}$

$$D[3] = \min(\infty, D[2] + C[2, 3]) = 60 \quad (7.2)$$

$$D[4] = \min(30, D[2] + C[2, 4]) = 30 \quad (7.3)$$

$$D[5] = \min(100, D[2] + C[2, 5]) = 100$$

Iteration 2

Select $w = 4$, so that $S = \{1, 2, 4\}$

$$D[3] = \min(60, D[4] + C[4, 3]) = 50 \quad (7.4)$$

$$D[5] = \min(100, D[4] + C[4, 5]) = 90$$

Iteration 3

Select $w = 3$, so that $S = \{1, 2, 4, 3\}$

$$D[5] = \min(90, D[3] + C[3, 5]) = 60$$

Iteration 4

Select $w = 5$, so that $S = \{1, 2, 4, 3, 5\}$

$$D[2] = 10 \quad (7.5)$$

$$D[3] = 50 \quad (7.6)$$

$$D[4] = 30 \quad (7.7)$$

$$D[5] = 60$$

Complexity of Dijkstra's Algorithm

With adjacency matrix representation, the running time is $O(n^2)$. By using an adjacency list representation and a partially ordered tree data structure for organizing the set $V - S$, the complexity can be shown to be

$$O(e \log n)$$

where e is the number of edges and n is the number of vertices in the digraph.



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