

09 B: Graph Algorithms II

CS1102S: Data Structures and Algorithms

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Graph, Vertices, Edges

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Directed graph

If the pairs are ordered, then the graph is *directed*.

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Weight

Sometimes the edges have a third component, known as either a *weight* or a *cost*. Such graphs are called *weighted graphs*.

Paths

Path

A *path* in a graph is a sequence of vertices $w_1, w_2, w_3, \dots, w_N$ such that $(w_i, w_{i+1}) \in E$ for $1 \leq i < N$. It is said to lead from w_1 to w_N .

The Shortest Path Problem

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Weighted graph: associated with each edge (v_i, v_j) is a cost $c_{i,j}$ to traverse the edge.

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Weighted path length

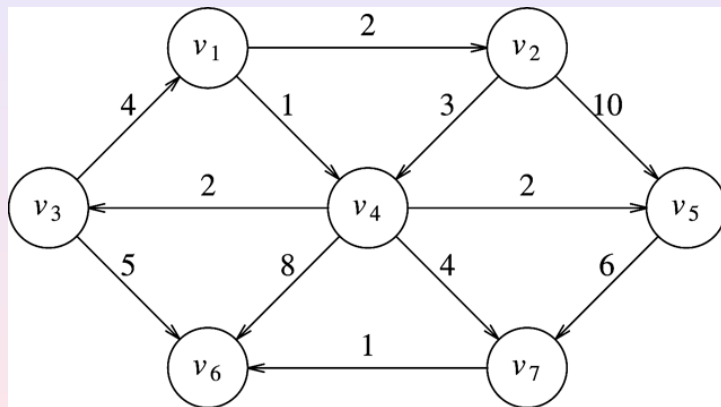
Cost of path $v_1 v_2 \cdots v_N$ is $\sum_{i=1}^{N-1} c_{i,i+1}$.

Single-Source Shortest-Path Problem

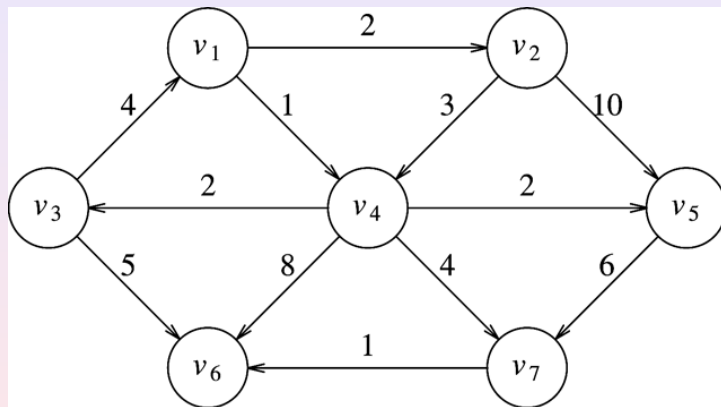
Problem

Given as input a weighted graph, $G = (V, E)$, and a distinguished vertex, s , find the shortest weighted path from s to every other vertex in G .

Example

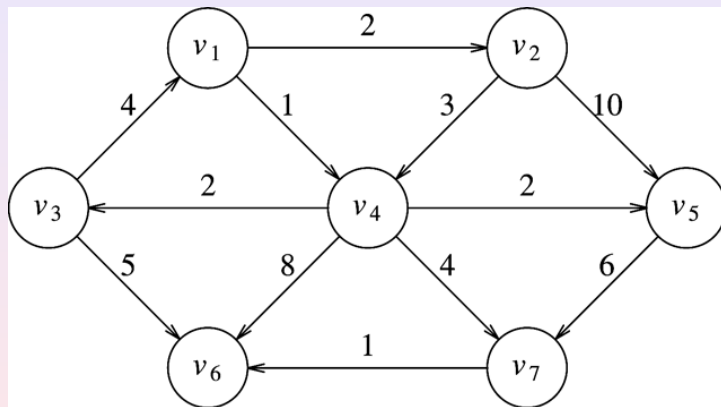


Example



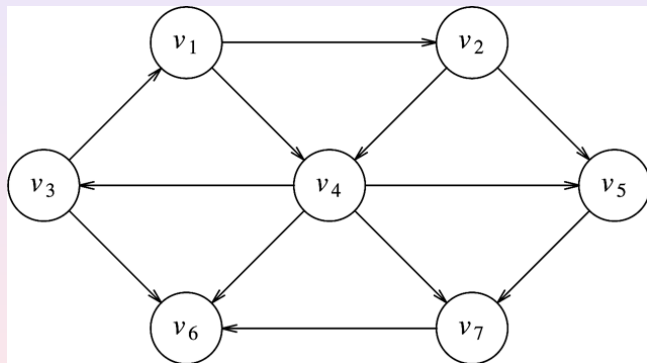
Shortest path from v_1 to v_6

Example



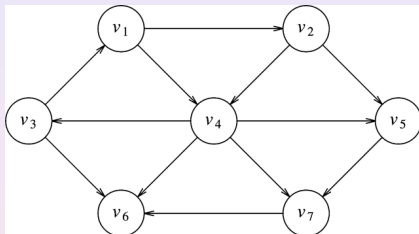
Shortest path from v_1 to v_6
has a cost of 6 and goes from v_1 to v_4 to v_7 to v_6 .

Unweighted Shortest Paths: Example



Find the shortest path from v_3 to all other vertices

Idea



Level-order traversal

Start with s (distance 0) and proceed in phases currDist , each time going through all vertices. If vertex is “known” and has distance currDist , set the distance of its neighbors to $\text{currDist} + 1$.

Implementation

```
void unweighted( Vertex s )
{
    for each Vertex v
    {
        v.dist = INFINITY;
        v.known = false;
    }

    s.dist = 0;

    for( int currDist = 0; currDist < NUM_VERTICES; currDist++ )
        for each Vertex v
            if( !v.known && v.dist == currDist )
            {
                v.known = true;
                for each Vertex w adjacent to v
                    if( w.dist == INFINITY )
                    {
                        w.dist = currDist + 1;
                        w.path = v;
                    }
            }
}
```

Inefficiency

Careless loop

In each phase, we go through all vertices. We can remember the “known” vertices in a data structure.

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Suitable data structure

Queue: will contain the vertices in order of increasing distance

Implementation

```
void unweighted( Vertex s )
{
    Queue<Vertex> q = new Queue<Vertex>( );

    for each Vertex v
        v.dist = INFINITY;

    s.dist = 0;
    q.enqueue( s );

    while( !q.isEmpty( ) )
    {
        Vertex v = q.dequeue( );

        for each Vertex w adjacent to v
            if( w.dist == INFINITY )
            {
                w.dist = v.dist + 1;
                w.path = v;
                q.enqueue( w );
            }
    }
}
```

Dijkstra's Algorithm: Idea

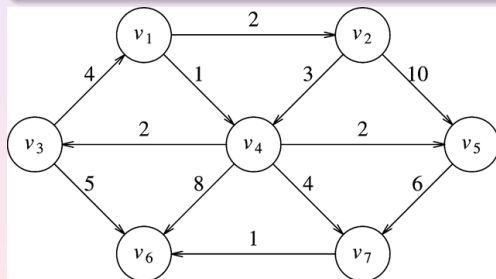
Idea

Treat nodes in the order of shortest distance

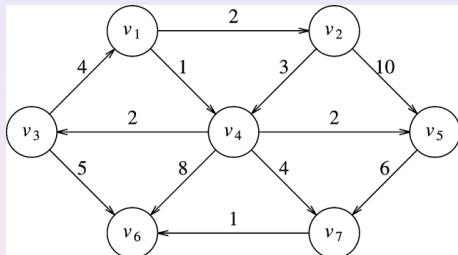
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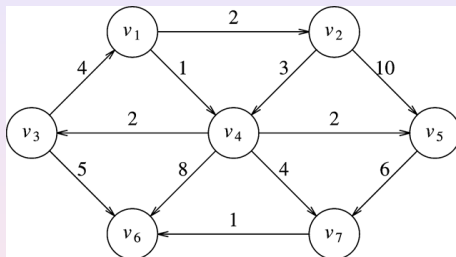
Example



Initial configuration:

v	$known$	d_v	p_v
v_1	F	0	0
v_2	F	∞	0
v_3	F	∞	0
v_4	F	∞	0
v_5	F	∞	0
v_6	F	∞	0
v_7	F	∞	0

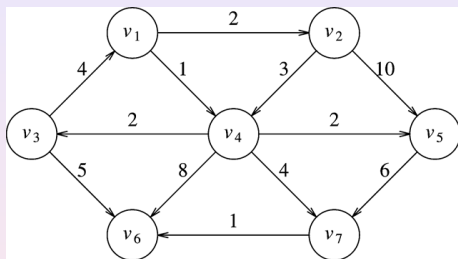
Example



After v_1 is declared known:

v	<i>known</i>	d_v	p_v
v_1	T	0	0
v_2	F	2	v_1
v_3	F	∞	0
v_4	F	1	v_1
v_5	F	∞	0
v_6	F	∞	0
v_7	F	∞	0

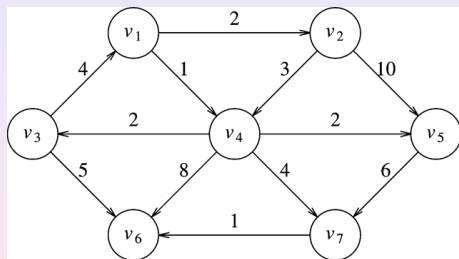
Example



After v_4 is declared known:

v	<i>known</i>	d_v	p_v
v_1	T	0	0
v_2	F	2	v_1
v_3	F	3	v_4
v_4	T	1	v_1
v_5	F	3	v_4
v_6	F	9	v_4
v_7	F	5	v_4

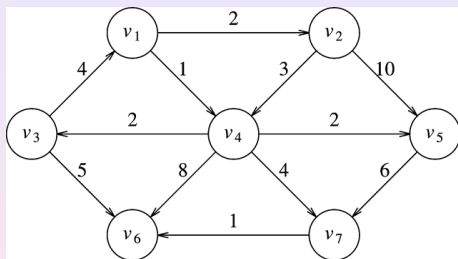
Example



After v_2 is declared known:

v	$known$	d_v	p_v
v_1	T	0	0
v_2	T	2	v_1
v_3	F	3	v_4
v_4	T	1	v_1
v_5	F	3	v_4
v_6	F	9	v_4
v_7	F	5	v_4

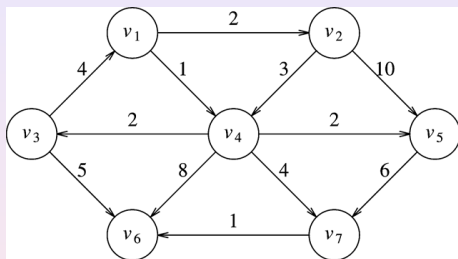
Example



After v_5 and then v_3 are declared known:

v	<i>known</i>	d_v	p_v
v_1	T	0	0
v_2	T	2	v_1
v_3	T	3	v_4
v_4	T	1	v_1
v_5	T	3	v_4
v_6	F	8	v_3
v_7	F	5	v_4

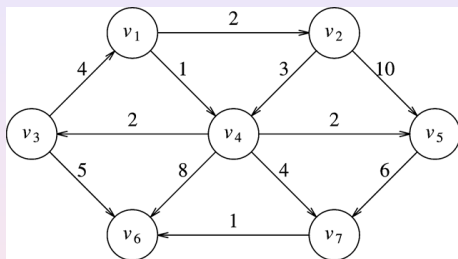
Example



After v_7 is declared known:

v	$known$	d_v	p_v
v_1	T	0	0
v_2	T	2	v_1
v_3	T	3	v_4
v_4	T	1	v_1
v_5	T	3	v_4
v_6	F	6	v_7
v_7	T	5	v_4

Example



After v_6 is declared known:

v	$known$	d_v	p_v
v_1	T	0	0
v_2	T	2	v_1
v_3	T	3	v_4
v_4	T	1	v_1
v_5	T	3	v_4
v_6	T	6	v_7
v_7	T	5	v_4

Pseudocode for Dijkstra's Algorithm

```
void dijkstra( Vertex s )
{
    for each Vertex v
    {
        v.dist = INFINITY;
        v.known = false;
    }

    s.dist = 0;

    for( ; ; )
    {
        Vertex v = smallest unknown distance vertex;
        if( v == NOT_A_VERTEX )
            break;
        v.known = true;

        for each Vertex w adjacent to v
            if( !w.known )
                if( v.dist + cvw < w.dist )
                {
                    // Update w
                    decrease( w.dist to v.dist + cvw );
                    w.path = v;
                }
    }
}
```

Shortest Subpath

Lemma

Any subpath of a shortest path must also be a shortest path.

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Proof

By contradiction: Assume that a subpath $p = v_i \cdots v_j$ of the shortest path $q = v_1 \cdots p \cdots v_k$ is not a shortest path. Then there is a shorter path p' from v_i to v_j . Plug p' into q to get $q' = v_1 \cdots p' \cdots v_k$ shorter than q !

Definition of Shortest Distance

Notation

We use the notation

$$\delta(v, w)$$

to denote the length of the shortest path from v to w .

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Distance

We call $\delta(v, w)$ the *distance* between v and w .

dist is a Relaxation

Lemma

At any point in time and for any vertex v , we have

$$v.\text{dist} \geq \delta(s, v)$$

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Lemma

At any point in time and for any vertex v , we have

$$v.\text{dist} \geq \delta(s, v)$$

Proof Idea

We show this by proving that whenever we set $v.\text{dist}$ to a finite value, there exists a path of that length.

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Hypothesis: claim holds for previous iterations.

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Proof

By induction over the number of iterations of outer loop.

Start: claim holds for s (distance 0) and all other vertices (distance ∞)

Hypothesis: claim holds for previous iterations.

Induction step: Updates are done such that an edge weight is added to a previously computed dist value

Order of Adding Vertices

Observation

Vertices are becoming known in order of increasing dist values.

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Observation

Once a vertex becomes known, its dist value does not change.

Correctness of Dijkstra's Algorithm

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    for each Vertex v
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        Vertex v = smallest unknown distance vertex;
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            if( !w.known )
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                    // Update w
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Main Correctness Lemma

Lemma

When we set $v.\text{known} = \text{true}$ then $v.\text{dist} = \delta(s, v)$.

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Correctness of Dijkstra's algorithm

Each iteration through the outer loop makes one vertex known. At the end every vertex v is known and thus, according to the lemma, its dist value is $\delta(s, v)$.

Proving the Main Lemma

Proof by contradiction

Assume that there is a vertex v for which the claim does not hold.

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Therefore, using relaxation property, when we set $v.\text{known} = \text{true}$ then $v.\text{dist} > \delta(s, v)$.

Proving the Main Lemma

Proof by contradiction

Assume that there is a vertex v for which the claim does not hold.

Therefore, using relaxation property, when we set $v.\text{known} = \text{true}$ then $v.\text{dist} > \delta(s, v)$.

Then, there must be a vertex u for which this is the case for the *first time* in the algorithm.

Situation

Analysis

Situation just before $u.\text{known} = \text{true}$: The value $u.\text{dist}$ reflects the length of the path given by $u.\text{path}$.

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The real shortest path

The real shortest path from s to u is shorter than $u.\text{dist}$.
Consider the real shortest path $s \cdots u$.

Situation

Jumping into “unknown”

Since $u.\text{known}$ still false, and $s.\text{known}=\text{true}$, there must be a pair of two neighboring vertices r and t in the real shortest path such that $r.\text{known}=\text{true}$ and $t.\text{known}=\text{false}$.

First pair

There must be a first pair x and y where this is the case.

Analysis of x and y

Processing of x

We have processed x , but not yet y . Since y 's dist value is decreased by $\text{decrease}(\dots)$, we know that

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Using hypothesis

Since x becomes known earlier than u , we have

$$x.\text{dist} = \delta(s, x)$$

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Using hypothesis

Since x becomes known earlier than u , we have

$$x.dist = \delta(s, x)$$

Shortest subpath

$s \cdots xy$ is subpath of shortest path, thus

$$\delta(s, y) = \delta(s, x) + c_{x,y} = x.dist + c_{x,y}$$

Recap

We have

$$y.\text{dist} \leq x.\text{dist} + c_{x,y}$$

and

$$\delta(s, y) = \delta(s, x) + c_{x,y} = x.\text{dist} + c_{x,y}$$

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Edge costs between y and u are non-negative, thus

$$\delta(s, y) \leq \delta(s, u)$$

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$$y.\text{dist} = \delta(s, y) \leq \delta(s, u) < u.\text{dist}$$

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and thus

$$y.\text{dist} = \delta(s, y) \leq \delta(s, u) < u.\text{dist}$$

If $y.\text{dist} < u.\text{dist}$, and since vertices become known in order of increasing distance, y would have become known before u , which contradicts the assumption that u is next vertex to become known!