

SHORTEST PATH ALGORITHM

CHAPTER 28

Lecture 21

CS2110 -Fall 2017

A7. Implement shortest-path algorithm

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One semester: Average time was 3.3 hours.

We give you complete set of test cases and a GUI to play with.

Efficiency and simplicity of code will be graded.

Read pinned A7 FAQs note carefully:

2. Important! Grading guidelines.

We demo it.

Dijkstra's shortest-path algorithm

Edsger Dijkstra, in an interview in 2010 (*CACM*):

... the algorithm for the shortest path, which I designed in about 20 minutes. One morning I was shopping in Amsterdam with my young fiance, and tired, we sat down on the cafe terrace to drink a cup of coffee, and I was just thinking about whether I could do this, and I then designed the algorithm for the shortest path. As I said, it was a 20-minute invention. [Took place in 1956]

Dijkstra, E.W. A note on two problems in Connexion with graphs. *Numerische Mathematik* 1, 269–271 (1959).

Visit <http://www.dijkstrascry.com> for all sorts of information on Dijkstra and his contributions. As a historical record, this is a gold mine.

Dijkstra's shortest-path algorithm

Dijkstra describes the algorithm in English:

- When he designed it in 1956 (he was 26 years old), most people were programming in assembly language.
- Only *one* high-level language: Fortran, developed by John Backus at IBM and not quite finished.

No theory of order-of-execution time —topic yet to be developed. In paper, Dijkstra says, “my solution is preferred to another one ... “the amount of work to be done seems considerably less.”

Dijkstra, E.W. A note on two problems in Connexion with graphs. *Numerische Mathematik* 1, 269–271 (1959).

1968 NATO Conference on Software Engineering

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- In Garmisch, Germany
- Academicians and industry people attended
- For first time, people admitted they did not know what they were doing when developing/testing software. Concepts, methodologies, tools were inadequate, missing
- The term *software engineering* was born at this conference.
- The NATO Software Engineering Conferences:
<http://homepages.cs.ncl.ac.uk/brian.randell/NATO/index.html>
Get a good sense of the times by reading these reports!

1968 NATO Conference on Software Engineering, Garmisch, Germany



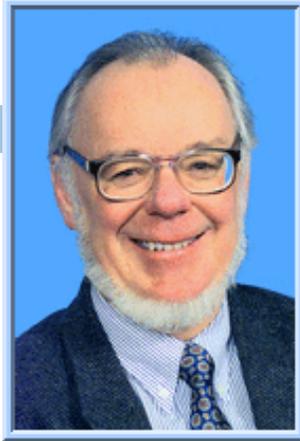
Term “software engineering” coined for this conference

1968 NATO Conference on Software Engineering, Garmisch, Germany



1968/69 NATO Conferences on Software Engineering

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Editors of the proceedings

Beards

The reason why some people grow aggressive tufts of facial hair is that they do not like to show the chin that isn't there.

a grook by Piet Hein



Edsger Dijkstra



Niklaus Wirth



Tony Hoare



David Gries

Dijkstra's shortest path algorithm

The n (> 0) nodes of a graph numbered $0..n-1$.

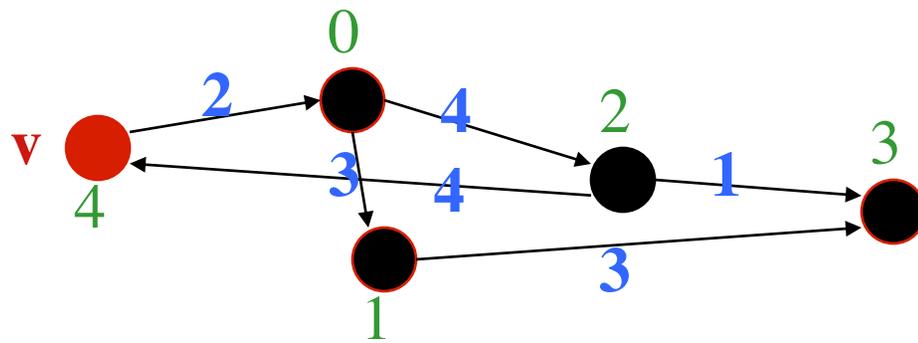
Each edge has a positive weight.

$wgt(v1, v2)$ is the weight of the edge from node $v1$ to $v2$.

Some node v be selected as the *start* node.

Calculate length of shortest path from v to each node.

Use an array $d[0..n-1]$: for **each** node w , store in $d[w]$ the length of the shortest path from v to w .



$$d[0] = 0$$

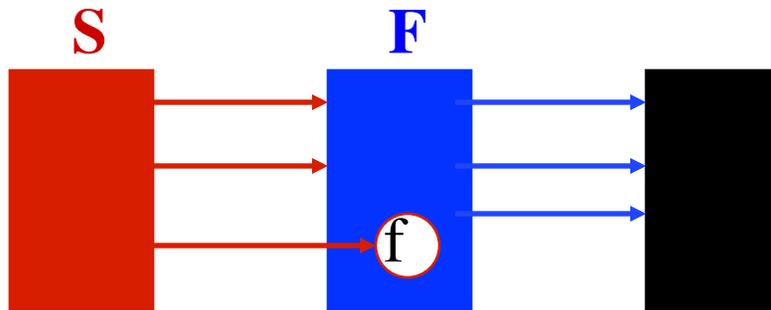
$$d[1] = 3$$

$$d[2] = 2$$

$$d[3] = 4$$

$$d[4] = 4$$

Settled Frontier Far off



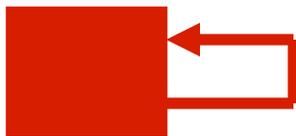
The loop invariant

(edges leaving the Far off set and edges from the Frontier to the Settled set are not shown)

1. For a Settled node s , a shortest path from v to s contains only settled nodes and $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For a Frontier node f , at least one $v \rightarrow f$ path contains only settled nodes (except perhaps for f) and $d[f]$ is the length of the shortest such path
3. All edges leaving S go to F .



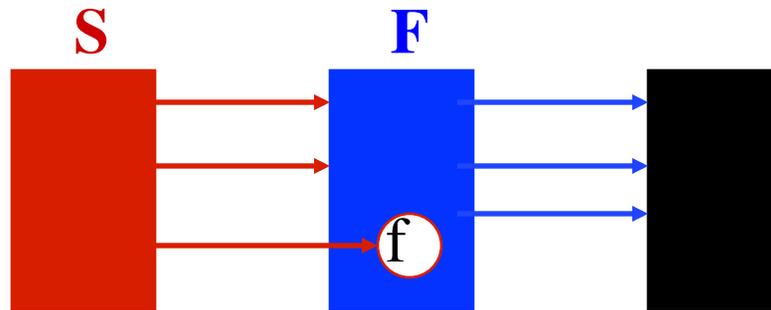
Settled S



This edge does not leave S !

Another way of saying 3:
There are no edges from S to the far-off set.

Settled Frontier Far off



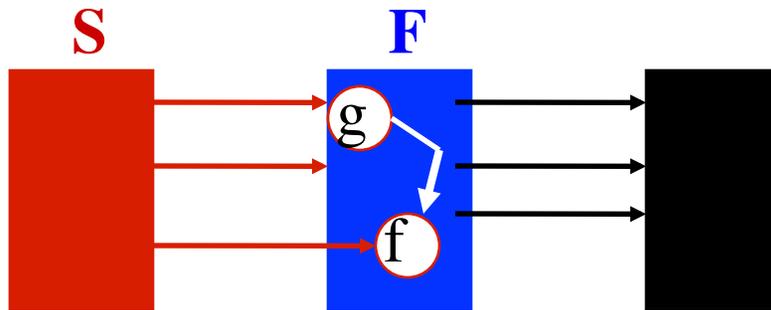
The loop invariant

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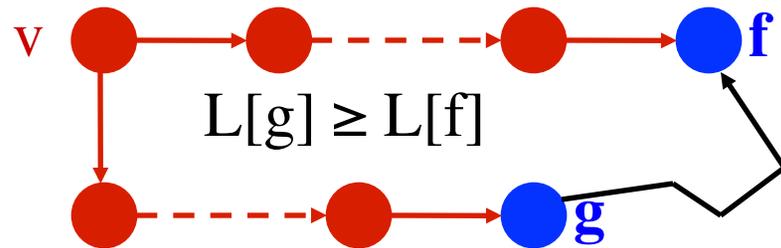
1. For a Settled node s , a shortest path from v to s contains only settled nodes and $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For a Frontier node f , at least one $v \rightarrow f$ path contains only settled nodes (except perhaps for f) and $d[f]$ is the length of the shortest such path
3. All edges leaving S go to F .



Settled **Frontier** **Far off**



Theorem about the invariant

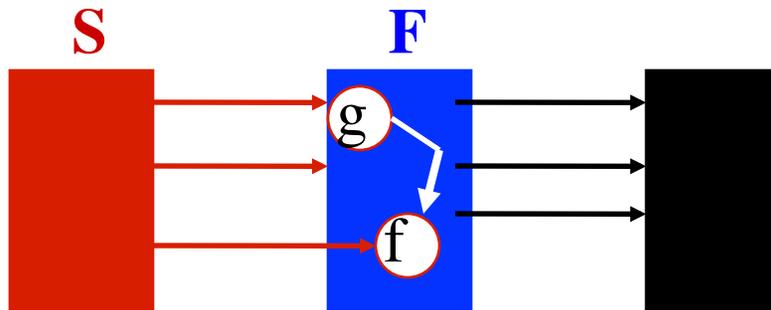


1. For a Settled node s , $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For a Frontier node f , $d[f]$ is length of shortest $v \rightarrow f$ path using only Settled nodes (except for f).
3. All edges leaving S go to F .

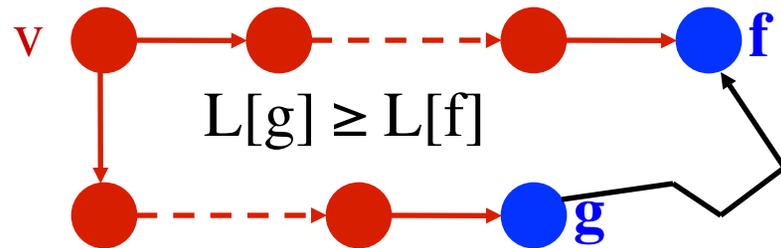
Theorem. For a node f in F with minimum d value (over nodes in F), $d[f]$ is the length of a shortest path from v to f .

Proof. Show that any other $v \rightarrow f$ path has a length $\geq d[f]$. Look only at case that v is in S .

Settled **Frontier** **Far off**



Theorem about the invariant



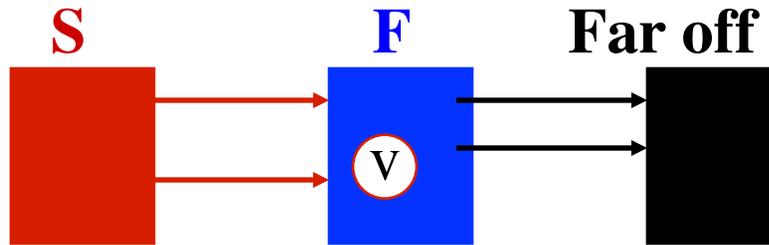
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2. For a Frontier node f , $d[f]$ is length of shortest $v \rightarrow f$ path using only Settled nodes (except for f).
3. All edges leaving S go to F .

Theorem. For a node f in F with minimum d value (over nodes in F), $d[f]$ is the length of a shortest path from v to f .

Case 1: v is in S .

Case 2: v is in F . Note that $d[v]$ is 0; it has minimum d value

The algorithm



$$S = \{ \}; F = \{ v \}; d[v] = 0;$$

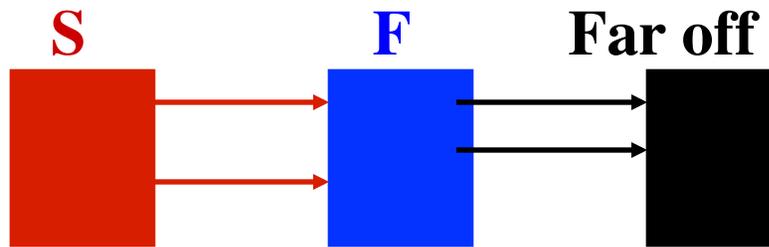
1. For s , $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For f , $d[f]$ is length of shortest $v \rightarrow f$ path using red nodes (except for f).
3. Edges leaving S go to F .

Theorem: For a node f in F with min d value, $d[f]$ is shortest path length

Loopy question 1:

How does the loop start? What is done to truthify the invariant?

The algorithm



1. For s , $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For f , $d[f]$ is length of shortest $v \rightarrow f$ path using red nodes (except for f).
3. Edges leaving S go to F .

Theorem: For a node f in F with min d value, $d[f]$ is shortest path length

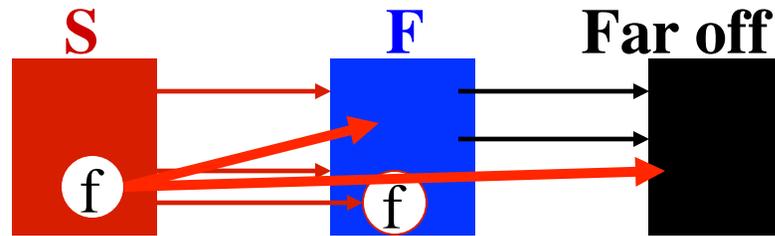
```
S = { }; F = { v }; d[v] = 0;  
while ( F ≠ { } ) {
```

```
}
```

Loopy question 2:

When does loop stop? When is array d completely calculated?

The algorithm



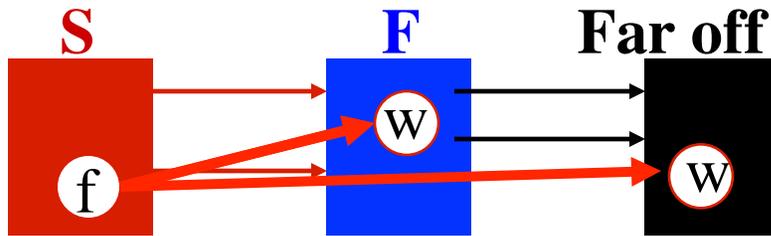
```
S = { }; F = { v }; d[v] = 0;
while ( F ≠ { } ) {
    f = node in F with min d value;
    Remove f from F, add it to S;
```

1. For s , $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For f , $d[f]$ is length of shortest $v \rightarrow f$ path using red nodes (except for f).
3. Edges leaving S go to F .

Theorem: For a node f in F with min d value, $d[f]$ is shortest path length

Loopy question 3: Progress toward termination?

The algorithm



1. **For s**, $d[s]$ is length of shortest $v \rightarrow s$ path.
2. **For f**, $d[f]$ is length of shortest $v \rightarrow f$ path using **red nodes** (except for **f**).
3. **Edges leaving S go to F.**

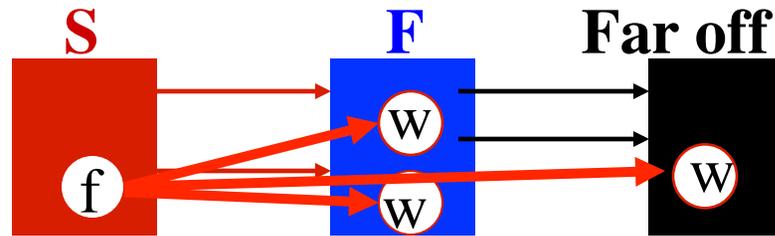
Theorem: For a node **f** in **F** with min d value, $d[f]$ is shortest path length

```

S= { }; F= { v }; d[v]= 0;
while ( F ≠ { } ) {
    f= node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            } else {
            }
        }
    }
}
    
```

Loopy question 4: Maintain invariant?

The algorithm



- 1. For s, $d[s]$** is length of shortest $v \rightarrow s$ path.
- 2. For f, $d[f]$** is length of shortest $v \rightarrow f$ path using **red nodes** (except for **f**).
- 3. Edges leaving S go to F.**

Theorem: For a node **f** in **F** with min d value, $d[f]$ is shortest path length

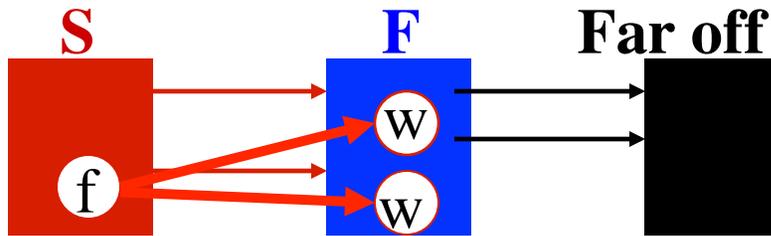
```

S= { }; F= { v }; d[v]= 0;
while ( F ≠ { } ) {
    f= node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w]= d[f] + wgt(f, w);
            add w to F;
        } else {
        }
    }
}

```

Loopy question 4: Maintain invariant?

The algorithm



1. For s , $d[s]$ is length of shortest $v \rightarrow s$ path.
2. For f , $d[f]$ is length of shortest $v \rightarrow f$ path of form
3. Edges leaving S go to F .

Theorem: For a node f in F with min d value, $d[f]$ is its shortest path length

```

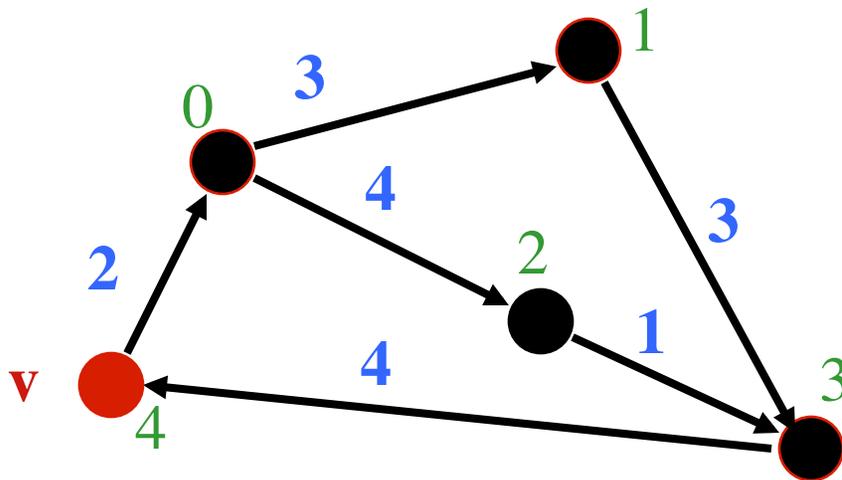
S= { }; F= { v }; d[v]= 0;
while ( F ≠ { } ) {
    f= node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w]= d[f] + wgt(f, w);
            add w to F;
        } else
            if (d[f] + wgt (f,w) < d[w]) {
                d[w]= d[f] + wgt(f, w);
            }
        }
    }
}

```

Algorithm is finished!

Extend algorithm to include the shortest path

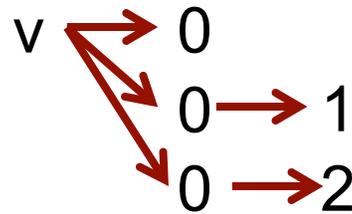
Let's extend the algorithm to calculate not only the length of the shortest path but the path itself.



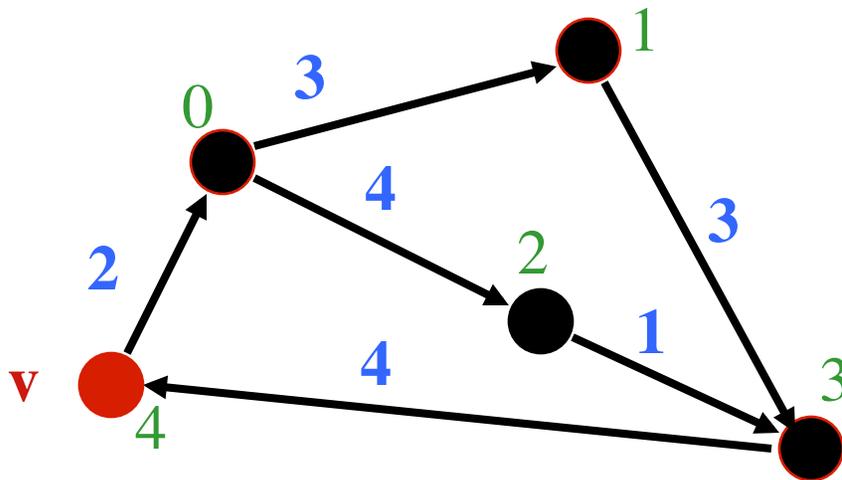
$$\begin{aligned}d[0] &= 2 \\d[1] &= 5 \\d[2] &= 6 \\d[3] &= 7 \\d[4] &= 0\end{aligned}$$

Extend algorithm to include the shortest path

Question: should we store in v itself the shortest path from v to every node? Or do we need another data structure to record these paths?



Not finished!
And how do we maintain it?



$d[0] = 2$
 $d[1] = 5$
 $d[2] = 6$
 $d[3] = 7$
 $d[4] = 0$

Extend algorithm to include the shortest path

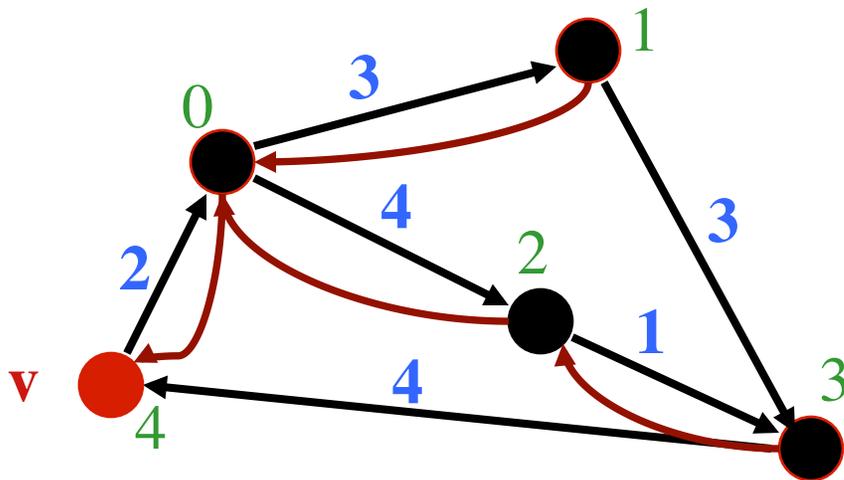
For each node, maintain the *backpointer* on the shortest path to that node.

Shortest path to 0 is $v \rightarrow 0$. Node 0 backpointer is 4.

Shortest path to 1 is $v \rightarrow 0 \rightarrow 1$. Node 1 backpointer is 0.

Shortest path to 2 is $v \rightarrow 0 \rightarrow 2$. Node 2 backpointer is 0.

Shortest path to 3 is $v \rightarrow 0 \rightarrow 2 \rightarrow 1$. Node 3 backpointer is 2.



bk[w] is w's backpointer

$$d[0] = 2$$

$$bk[0] = 4$$

$$d[1] = 5$$

$$bk[1] = 0$$

$$d[2] = 6$$

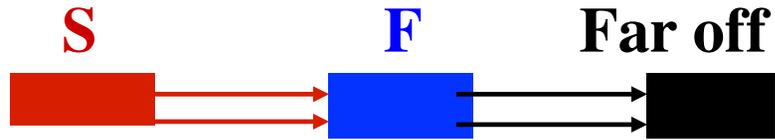
$$bk[2] = 0$$

$$d[3] = 7$$

$$bk[3] = 2$$

$$d[4] = 0$$

$$bk[4] \text{ (none)}$$



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

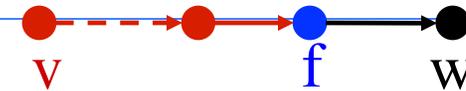
while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
            add w to F; bk[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}

```

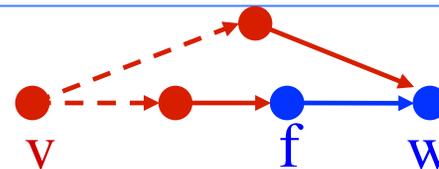
Maintain backpointers

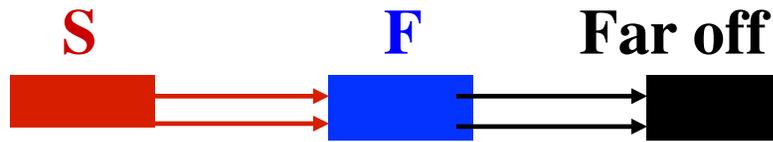
Wow! It's so easy to maintain backpointers!

When w not in S or F:
Getting first shortest path so far:



When w in S or F and have shorter path to w:





$S = \{ \}; F = \{v\}; d[v] = 0;$

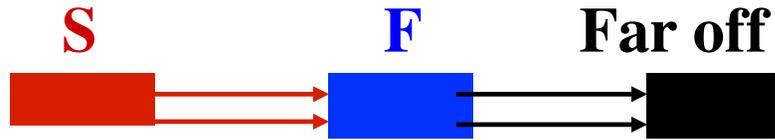
```

while (F  $\neq$  { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
            add w to F; bk[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}

```

This is our final high-level algorithm. These issues and questions remain:

1. How do we implement F?
2. The nodes of the graph will be objects of class Node, not ints. How will we maintain the data in arrays d and bk?
3. How do we tell quickly whether w is in S or F?
4. How do we analyze execution time of the algorithm?

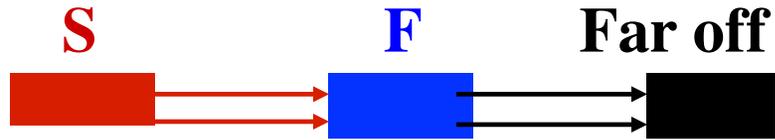


1. How do we implement F?

```

S= { }; F= {v}; d[v]= 0;
while (F ≠ { }) {
    f= node in F with min d value;
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    for each neighbor w of f {
        if (w not in S or F) {
            d[w]= d[f] + wgt(f, w);
            add w to F; bk[w]= f;
        } else if (d[f]+wgt (f,w) < d[w]) {
            d[w]= d[f] + wgt(f, w);
            bk[w]= f;
        }
    }
}
}

```



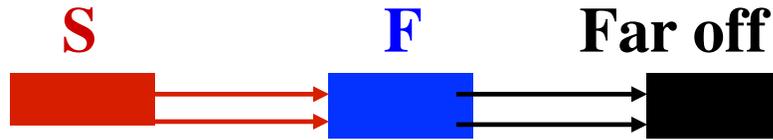
$S = \{ \}; F = \{v\}; d[v] = 0;$

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while (F  $\neq$  { }) {
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            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}

```

For what nodes do we need a distance and a backpointer?



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

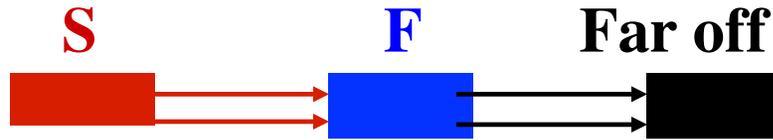
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        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}

```

For what nodes do we need a distance and a backpointer?

For every node in S or F we need both its d-value and its backpointer (null for v)

Don't want to use arrays d and bk! Instead, keep information associated with a node. What data structure to use for the two values?



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

while (F  $\neq$  { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
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        } else if (d[f] + wgt(f, w) < d[w]) {
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        }
    }
}

```

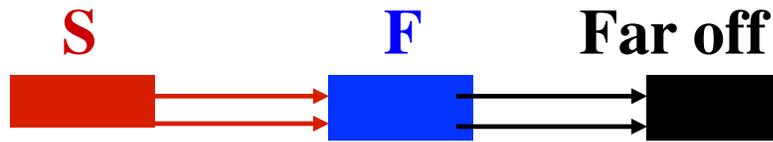
For what nodes do we need a distance and a backpointer?

For every node in S or F we need both its d-value and its backpointer (null for v)

```

public class SFinfo {
    private node bckPntr;
    private int distance;
    ...
}

```



$S = \{ \}$; $F = \{v\}$; $d[v] = 0$;

```

while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
            add w to F; bk[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}

```

F implemented as a heap of Nodes.

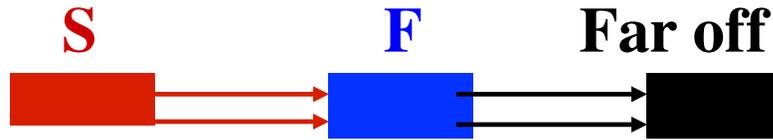
What data structure do we use to maintain an SFinfo object for each node in S and F?

For every node in S or F we need both its d-value and its backpointer (null for v):

```

public class SFinfo {
    private node bckPtr;
    private int distance;
    ...
}

```



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
            add w to F; bk[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}

```

Algorithm to implement

For every node in S or F, we need an object of class SFdata. What data structure to use?

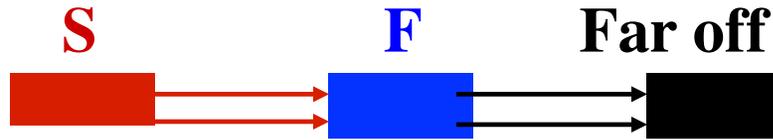
HashMap<Node, SFinfo> map

You will implement the algorithm on this slide. S, d, and b are replaced by map. F is implemented as a min-heap.

```

public class SFinfo {
    private node bckPtr;
    private int distance;
    ...
}

```



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
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}
HashMap<Node, SFinfo> map

```

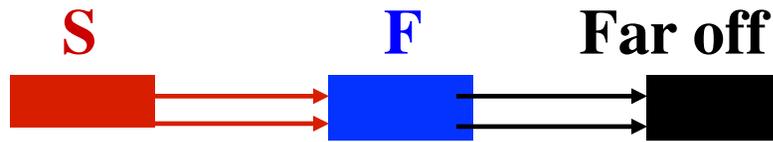
Investigate execution time.
 Important: understand algorithm well enough to easily determine the total number of times each part is executed/evaluated

Assume:
 n nodes reachable from v
 e edges leaving those n nodes

```

public class SFinfo {
    private node backPtr;
    private int distance; ... }

```



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
            add w to F; bk[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}
HashMap<Node, SFinfo> map

```

Assume:

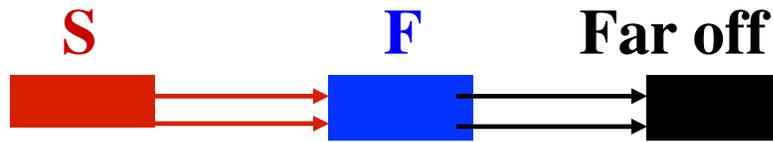
n nodes reachable from v
e edges leaving those n nodes

Example. How many times
does $F \neq \{ \}$ evaluate to true?

```

public class SFinfo {
    private node bckptr;
    private int distance; ... }

```



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
            d[w] = d[f] + wgt(f, w);
            add w to F; bck[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bck[w] = f;
        }
    }
}
HashMap<Node, SFinfo> map

```

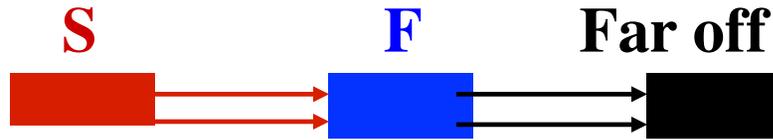
Directed graph
 n nodes reachable from v
 e edges leaving those n nodes
 $F \neq \{ \}$ is true n times

Harder: In total, how many times does the loop
 for each neighbor w of f
 find a neighbor and execute
 repetend?

```

public class SFinfo {
    private node bckPtr;
    private int distance; ... }

```



$S = \{ \}; F = \{v\}; d[v] = 0;$

```

while (F ≠ { }) {
    f = node in F with min d value;
    Remove f from F, add it to S;
    for each neighbor w of f {
        if (w not in S or F) {
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            add w to F; bk[w] = f;
        } else if (d[f] + wgt(f, w) < d[w]) {
            d[w] = d[f] + wgt(f, w);
            bk[w] = f;
        }
    }
}
HashMap<Node, SFinfo> map

```

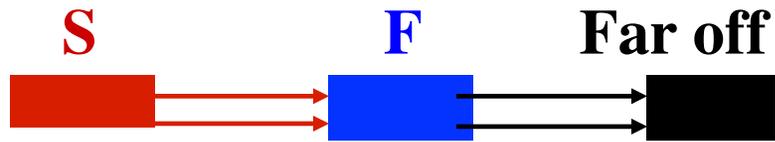
Directed graph
 n nodes reachable from v
 e edges leaving those n nodes
 $F \neq \{ \}$ is true n times
 First if-statement: done e times

How many times does
 w not in S or F
 evaluate to true?

```

public class SFinfo {
    private node bckPtr;
    private int distance; ... }

```



Number of times (x) each part is executed/evaluated

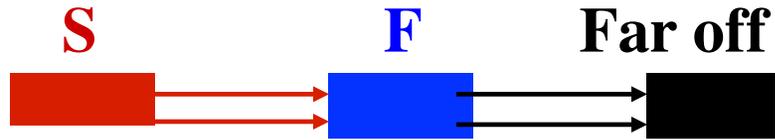
```

S= { }; F= {v}; d[v]= 0;
while (F ≠ { }) {
  f= node in F with min d value;
  Remove f from F, add it to S;
  for each neighbor w of f {
    if (w not in S or F) {
      d[w]= d[f] + wgt(f, w);
      add w to F; bk[w]= f;
    } else if (d[f]+wgt (f,w) < d[w]) {
      d[w]= d[f] + wgt(f, w);
      bk[w]= f;
    }
  }
}
}

```

1 x
 true n x, false 1 x
 n x
 n x
 true e x, false n x
 done e x, true n-1 x, false e-(n-1) x
 n-1 x
 n-1 x
 done e-(n-1) x, true ?
 done at most e-(n-1) x
 done at most e-(n-1) x

Assume: directed graph, using adjacency list
 n nodes reachable from v
 e edges leaving those n nodes



```

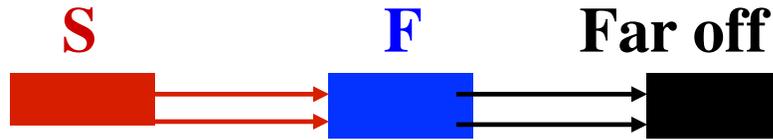
S= { }; F= {v}; d[v]= 0;
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  f= node in F with min d value;
  Remove f from F, add it to S;
  for each neighbor w of f {
    if (w not in S or F) {
      d[w]= d[f] + wgt(f, w);
      add w to F; bk[w]= f;
    } else if (d[f]+wgt (f,w) < d[w]) {
      d[w]= d[f] + wgt(f, w);
      bk[w]= f;
    }
  }
}

```

To find an upper bound on time complexity, multiply complexity of each part by the number of times its executed.

Then add them up.

Assume: directed graph, using adjacency list
 n nodes reachable from v
 e edges leaving those n nodes



Expected time

```

S= { }; F= {v}; d[v]= 0;
while (F ≠ { }) {
  f= node in F with min d value;
  Remove f from F, add it to S;
  for each neighbor w of f {
    if (w not in S or F) {
      d[w]= d[f] + wgt(f, w);
      add w to F; bk[w]= f;
    } else if (d[f]+wgt (f,w) < d[w]) {
      d[w]= d[f] + wgt(f, w);
      bk[w]= f;
    }
  }
}
}

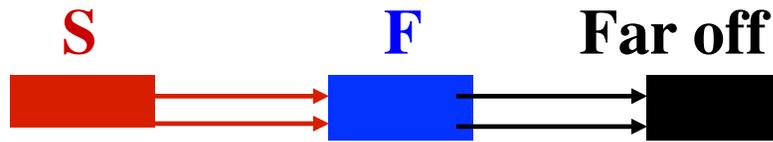
```

```

1 * O(1)
(n+1) * O(1)
n * O(1)
n * (O(log n) + O(1))
(e+n) * O(1)
e * O(1)
(n-1) * O(1)
(n-1) * O(log n)
(e-(n-1)) * O(1)
(e-(n-1)) * O(log n)
(e-(n-1)) * O(1)

```

Assume: directed graph, using adjacency list
 n nodes reachable from v
 e edges leaving those n nodes



Expected time

```

S= { }; F= {v}; d[v]= 0;
while (F ≠ { }) {
  f= node in F with min d value;
  Remove f from F, add it to S;
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    } else if (d[f]+wgt (f,w) < d[w]) {
      d[w]= d[f] + wgt(f, w);
      bk[w]= f;
    }
  }
}

```

| | |
|-----------------------|----|
| 1 * O(1) | 1 |
| (n+1) * O(1) | 2 |
| n * O(1) | 3 |
| n * (O(log n) + O(1)) | 4 |
| (e+n) * O(1) | 5 |
| e * O(1) | 6 |
| (n-1) * O(1) | 7 |
| (n-1) * O(log n) | 8 |
| (e-(n-1)) * O(1) | 9 |
| (e-(n-1)) * O(log n) | 10 |
| (e-(n-1)) * O(1) | 11 |

Dense graph, so e close to n*n: Line 10 gives $O(n^2 \log n)$

Sparse graph, so e close to n: Line 4 gives $O(n \log n)$