Single-Source Shortest Path

Data Structures and Algorithms

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The Shortest Path Problem

• Given a map, what's the shortest path from s to t?

• Consider a graph G=(V,E) and a weight function w that associates a real-valued weight w(u,v) to each edge (u,v). Given s and t in V, what's the \min weight path

from s to t?

Weights are not always lengths.

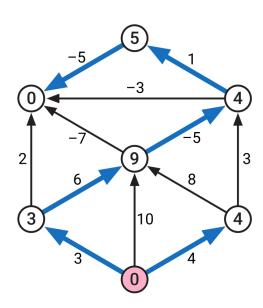
• E.g., time/cost to walk the edge.

- The graph can be directed.
 - Thus $w(u, v) \neq w(v, u)$ possible.
- Negative edge weight allowed.
- Negative cycle not allowed.
 - Problem not well-defined then.



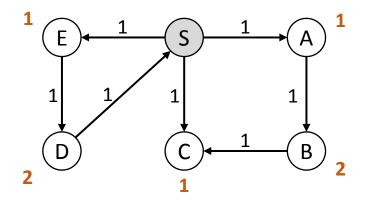
Single-Source Shortest Path (SSSP)

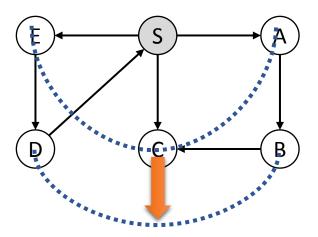
- The SSSP Problem: Given a graph G = (V, E) and a weight function W, given a source node S, find a shortest path from S to every node $U \in V$.
- Consider <u>directed</u> graphs <u>without</u> negative cycle.
- Case 1: Unit weight.
- Case 2: Arbitrary positive weight.
- Case 3: Arbitrary weight without cycle.
- Case 4: Arbitrary weight.



SSSP in unit weight graphs

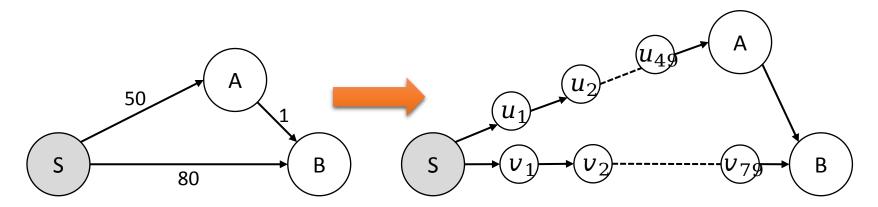
- How to solve SSSP in an unit weight graph?
 - That is, a graph in which each edge is of weight one.
- How to "traverse by layer" in an unweighted graph?
 - Visit all distance d node before visiting any distance d+1 node.
- Simple, just use BFS!





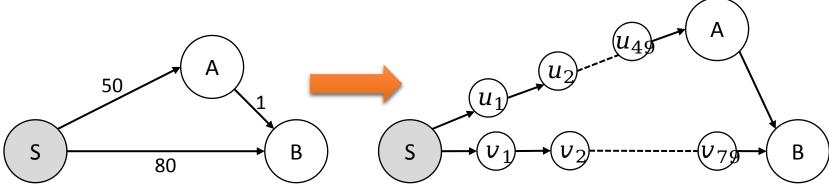
SSSP in positive weight graphs

- Solve SSSP in a graph with arbitrary positive weights?
- Extension of unit graph SSSP algorithm:
 - Add dummy nodes on edges so graph becomes unit weight graph.
 - Run BFS on the resulting graph.
- Problem with this approach?
- Too slow when edge weights differs a lot!



SSSP in positive weight graphs

- Simple BFS extension for SSSP in positive weight graphs:
 - Add dummy nodes on edges so graph becomes unit weight graph.
 - Run BFS on the resulting graph.
- The algorithm is too slow when edge weights differ a lot!
- To save time, bypass the events that process dummy nodes!
 - ullet Imagine we have an alarm clock T_u for each node u
 - Alarm for source node s goes off at time 0
 - If T_u goes off, for each edge (u,v), update $T_v = \min\{T_v, T_u + w(u,v)\}$

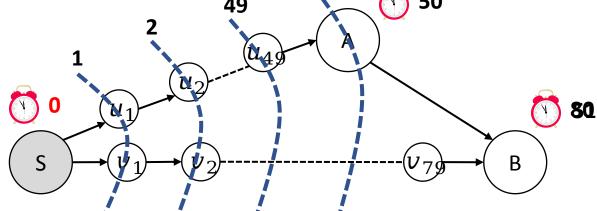


SSSP in positive weight graphs

Extension of the BFS algorithm

- Extension of BFS for SSSP in positive weight graphs:
 - ullet Imagine we have an alarm clock $T_{\,u}$ for each node u
 - Alarm for source node S goes off at time 0
 - If T_u goes off, for each edge (u,v), update $T_v = \min\{T_v, T_u + w(u,v)\}$
- This process is just mimicking the BFS process!
- At any time, value of T_u is an estimate of dist(s, u).

• At any time, $T_u \ge dist(s, u)$, with a quality holds when T_u goes off.



SSSP in positive weight graphs via extension of BFS Dijkstra's algorithm

- Extension of BFS for SSSP in positive weight graphs:
 - ullet Imagine we have an alarm clock $T_{\,u}$ for each node u
 - Alarm for source node s goes off at time 0
 - If T_u goes off, for each edge (u, v), update $T_v = \min\{T_v, T_u + w(u, v)\}$
- How to implement the "alarm clock"?
- Use priority queue (such as binary heap).



Edsger W. Dijkstra (1930-2002) ACM Turing Award Recipient

SSSP in positive weight graphs via extension of BFS

Dijkstra's algorithm

- Extension of BFS for SSSP in positive weight graphs:
 - Imagine we have an alarm clock T_u for each node u
 - Alarm for source node S goes off at time 0
 - If T_u goes off, for each edge (u, v), update $T_v = \min\{T_v, T_u + w(u, v)\}$

How to implement the "alarm clock"?



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SSSP in positive weight graphs via extension of BFS

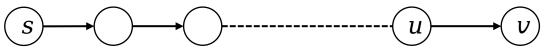
Dijkstra's algorithm

- Correctness of Dijkstra's algorithm?
- Similar to the correctness proof of BFS.
- Efficiency of Dijkstra's algorithm?
- $O((n+m)\log n)$ when using a binary heap.

```
DijkstraSSSP(G,s):
                        O(n) in total
for (each u in V)
  u.dist=INF, u.parent=NIL
                              O(n) in total
s.dist = 0
Build priority queue Q based on dist
while (!Q.empty())
                        O(n\log n) in total
  u = O.ExtractMin()
  for (each edge (u, v) in E)
    if (v.dist > u.dist + w(u,v))
      v.dist = u.dist + w(u,v)
      v.parent = u
                         O(m\log n) in total
      Q.DecreaseKey(v)
```

Alternative derivation of Dijkstra's alg.

- What's BFS doing: <u>expand</u> outward from S, growing the <u>region</u> to which distances and shortest paths are known.
- Growth should be orderly: closest nodes first.
- Q: But how to identify the node to expend to?
- Consider a *shortest path* from source S to V via U.

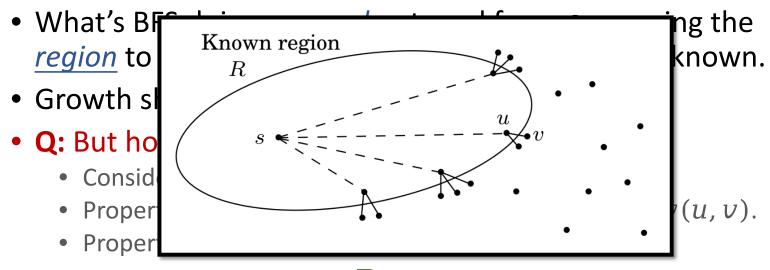


- It must be dist(s, v) = dist(s, u) + w(u, v).
 - Thus shortest path exhibits **optimal substructure** property.
- It must be dist(s, v) > dist(s, u).
 - Since we are considering positive edge weight graphs.

Alternative derivation of Dijkstra's alg.

- What's BFS doing: <u>expand</u> outward from *S*, growing the <u>region</u> to which distances and shortest paths are known.
- Growth should be *orderly*: closest nodes first.
- **Q:** But how to identify the node to expend to?
 - Consider a shortest path from source S to V via U.
 - Property 1: It must be dist(s, v) = dist(s, u) + w(u, v).
 - Property 2: It must be dist(s, v) > dist(s, u).
- A: Given "known region R", find $\min_{u' \in R, v' \in V-R} \{dist(s, u') + w(u', v')\}.$
 - Assume v is the node to expend to. (A shortest path is $s \to u \to v$.)
 - Property 2 ensures $u \in R$.
 - Property 1 then ensures we correctly identify ν to expend to.

Alternative derivation of Dijkstra's alg.

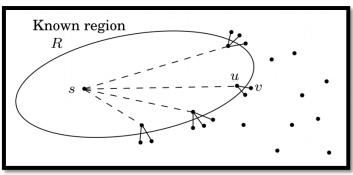


- A: Given "known region R", find $\min_{u' \in R, v' \in V-R} \{dist(s, u') + w(u', v')\}.$
 - Assume v is the node to expend to. (A shortest path is $s \to u \to v$.)
 - Property 2 ensures $u \in R$.
 - Property 1 then ensures we correctly identify ν to expend to.

- What's BFS doing: <u>expand</u> outward from S, growing the <u>region</u> to which distances and shortest paths are known.
- How to expend: Given "known region R", expend to node with $\min_{u' \in R, v' \in V-R} \{dist(s, u') + w(u', v')\}.$

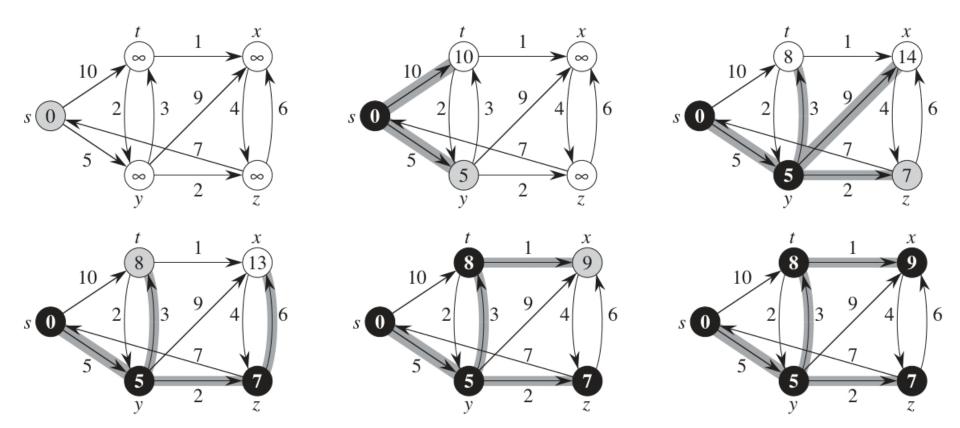
```
DijkstraSSSP(G,s):
for (each u in V)
   u.dist=INF, u.parent=NIL
s.dist = 0
Build priority queue Q based on dist
while (!Q.empty())
   u = Q.ExtractMin()
   for (each edge (u,v) in E)
      if (v.dist > u.dist + w(u,v))
      v.dist = u.dist + w(u,v)
      v.parent = u
      Q.DecreaseKey(v)
```





```
DijkstraSSSPAbs(G,s):
for (each u in V)
    u.dist = INF
s.dist = 0
R = Ø
while (R != V)
    Find node v in V-R with min v.dist
    Add v to R
    for (each edge (v,z) in E)
        if (z.dist > v.dist + w(v,z))
        z.dist = v.dist + w(v,z)
```

DijkstraSSSP(G,s): for (each u in V) u.dist=INF, u.parent=NIL s.dist = 0 Build priority queue Q based on dist while (!Q.empty()) u = Q.ExtractMin() for (each edge (u,v) in E) if (v.dist > u.dist + w(u,v)) v.dist = u.dist + w(u,v) v.parent = u Q.DecreaseKey(v)



DFS, BFS, Prim, Dijkstra, and others...

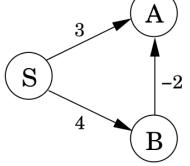
DFSIterSkeleton(G,s): Stack Q Q.push(s) while (!Q.empty()) u = Q.pop() if (!u.visited) u.visited = true for (each edge (u,v) in E) Q.push(v)

```
BFSSkeletonAlt(G,s):
FIFOQueue Q
Q.enque(s)
while (!Q.empty())
  u = Q.dequeue()
  if (!u.visited)
    u.visited = true
    for (each edge (u,v) in E)
       Q.enque(v)
```

```
PrimMSTSkeleton(G,x):
PriorityQueue Q
Q.add(x)
while (!Q.empty())
  u = Q.remove()
  if (!u.visited)
    u.visited = true
    for (each edge (u,v) in E)
      if (!v.visited and ...)
      Q.update(v,...)
```

```
DijkstraSSSPSkeleton(G,x):
PriorityQueue Q
Q.add(x)
while (!Q.empty())
   u = Q.remove()
   if (!u.visited)
       u.visited = true
      for (each edge (u,v) in E)
       if (!v.visited and ...)
            Q.update(v,...)
```

- Dijkstra's algorithm no longer works!
- Why would this happen?
- Dijkstra's algorithm for finding next closest node to expend to: Given "known region R", find $\min_{u' \in R, v' \in V R} \{dist(s, u') + w(u', v')\}$.
 - Assume v is the node to expend to. (A shortest path is $s \to u \to v$.)
 - Positive edge weights ensures $u \in R$.
 - Optimal substructure then ensures we correctly identify ν to expend to.
- "Shortest path from s to any node v must pass exclusively through nodes that are closer than v" no longer holds!



- But how dist values are maintained in Dijkstra is helpful:
 - Each node $u \neq s$ initially set $u.dist = \infty$, and s.dist = 0
 - When processing edge (u, v), execute procedure Update(u, v): $v. dist = min \{v. dist, u. dist + w(u, v)\}$
- This way two properties are maintained:
 - For any v, at any time, v. dist is either an overestimate, or correct.
 - Assume u is the last node on a shortest path from s to v. If u. dist is correct and we run Update(u, v), then v. dist becomes correct.
- Update(u, v) is <u>safe</u> and helpful!
 - [Safe] Regardless of the sequence of Update operations we execute, for any node v, value v. dist is either an overestimate or correct.
 - [Helpful] With correct sequence of Update, we get correct $v.\ dist.$

- Update(u, v): $v. dist = min\{v. dist, u. dist + w(u, v)\}$
- Update(u, v) is safe and helpful!
 - [Safe] Regardless of the sequence of Update operations we execute, for any node ν , value ν . dist is either overestimate or correct.
 - [Helpful] Assume u is the last node on a shortest path from s to v. If u. dist is correct and we run Update(u, v), then v. dist becomes correct.
- Observation 1: if Update(s, u_1), Update(u_1, u_2), ..., Update(u_{k-1}, u_k), Update(u_k, v) are executed, then we correctly obtain the shortest path.
- Observation 2: in above sequence, before and after each pdate, we can add arbitrary pdate sequence, and still get shortest path from s to t.
- Algorithm: simply Update <u>all</u> edges, for k+1 times!

- Update(u, v): $v. dist = min\{v. dist, u. dist + w(u, v)\}$
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 - [Safe] Regardless of the sequence of Update operations we execute, for any node ν , value ν . dist is either overestimate or correct.
 - [Helpful] Assume u is the last node on a shortest path from s to v. If u. dist is correct and we run Update(u, v), then v. dist becomes correct.
- Consider a shortest path from S to V.



- Observation 1: if $Update(s, u_1)$, $Update(u_1, u_2)$, ..., $Update(u_{k-1}, u_k)$, $Update(u_k, v)$ are executed, then we correctly obtain the shortest path.
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Consider a shortest path from S to V.



- Observation 1: if $Update(s, u_1)$, $Update(u_1, u_2)$, ..., $Update(u_{k-1}, u_k)$, $Update(u_k, v)$ are executed, then we correctly obtain the shortest path.
- Observation 2: in above sequence, before and after each pdate, we can add arbitrary pdate sequence, and still get shortest path from s to t.
- Algorithm: simply Update <u>all</u> edges, for k+1 times!
- But how large is k+1?
- Observation 3: any shortest path cannot contain a cycle. (WHY?)
- Algorithm: simply Update <u>all</u> edges, for n-1 times!

SSSP in directed graphs with negative weights The Bellman-Ford Algorithm

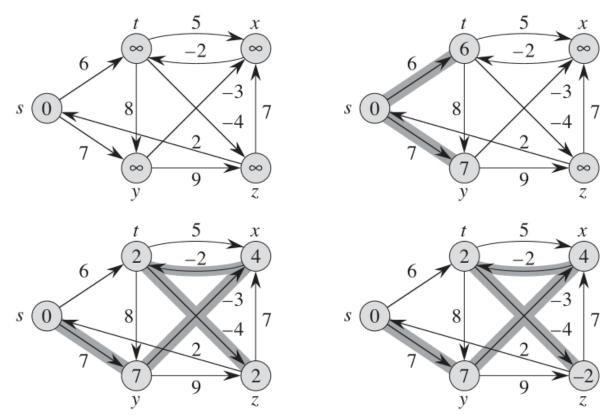
- Bellman-Ford Algorithm:
 - Update all edges;
 - Repeat above step for n-1 times.
- Time complexity of Bellman-Ford: $\Theta(nm)$

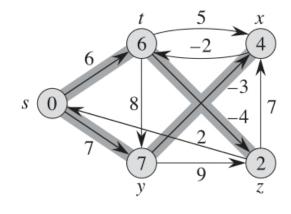
```
BellmanFordSSSP(G,s):
for (each u in V)
   u.dist=INF, u.parent=NIL
s.dist = 0
repeat n-1 times:
   for (each edge (u,v) in E)
    if (v.dist > u.dist + w(u,v))
       v.dist = u.dist + w(u,v)
      v.parent = u
```

BellmanFordSSSP(G,s):

```
for (each u in V)
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  for (each edge (u,v) in E)
   if (v.dist > u.dist + w(u,v))
     v.dist = u.dist + w(u,v)
     v.parent = u
```

Edge order: (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t)





SSSP in directed graphs with negative weights The Bellman-Ford Algorithm

- What if the graph contains a negative cycle?
- After n-1 repetitions of "Update all edges", some node v still has v.dist > u.dist + w(u,v).
- Bellman-Ford can also detect negative cycle!

```
BellmanFordSSSP(G,s):
for (each u in V)
   u.dist=INF, u.parent=NIL
s.dist = 0
repeat n-1 times:
   for (each edge (u,v) in E)
      if (v.dist > u.dist + w(u,v))
        v.dist = u.dist + w(u,v)
        v.parent = u
for (each edge (u,v) in E)
   if (v.dist > u.dist + w(u,v))
      return "Negative Cycle"
```

SSSP in DAG (with negative weights)

- Bellman-Ford still works, but we can be more efficient!
- Core idea of Bellman-Ford: perform a sequence of Update that includes every shortest path as a subsequence.
- Observation: in DAG, every path, thus every shortest path, is a subsequence in the topological order.

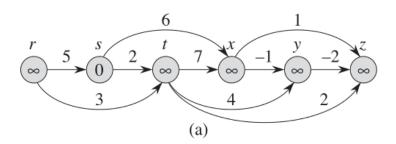
```
DAGSSSP(G,s):
    O(n + m) time

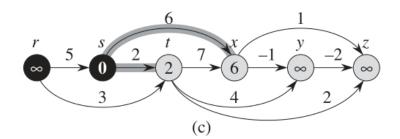
for (each u in V)
    u.dist=INF, u.parent=NIL

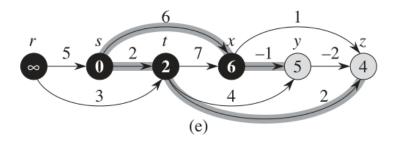
s.dist = 0

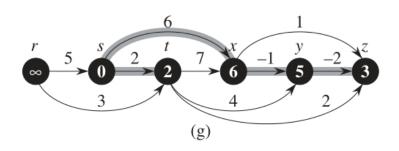
Run DFS to obtain topological order

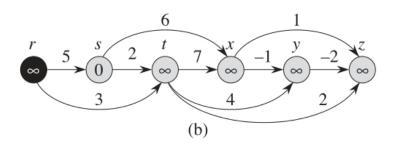
for (each node u in topological order)
    for (each edge (u,v) in E)
    if (v.dist > u.dist + w(u,v))
        v.dist = u.dist + w(u,v)
        v.parent = u
```

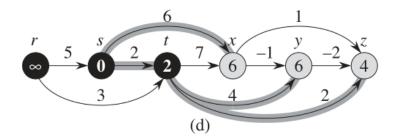


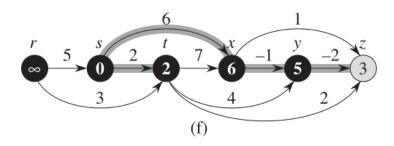












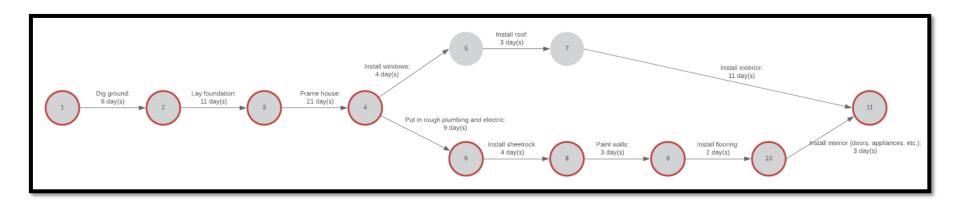
DAGSSSP(G,s):

```
for (each u in V)
  u.dist=INF, u.parent=NIL
s.dist = 0
Run DFS to obtain topological order
for (each node u in topological order)
  for (each edge (u,v) in E)
   if (v.dist > u.dist + w(u,v))
     v.dist = u.dist + w(u,v)
     v.parent = u
```

Application of SSSP in DAG

Computing Critical Path

- Assume you want to finish a task that involves multiple steps.
 Each step takes some time.
 For some step(s), it can only begin after certain steps are done.
- These dependency can be modeled as a DAG. (PERT Chart)
- How fast can you finish this task?
- Equivalently, longest path, a.k.a. critical path, in the DAG?
- Negate edge weights and compute a shortest path.



Summary

- The SSSP Problem: Given a graph G = (V, E) and a weight function W, given a source node S, find a shortest path from S to every node $V \in V$.
- Case 1: Unit weight graphs (directed or undirected).
 - Simply use BFS. O(n + m) runtime.
- Case 2: Arbitrary positive weight graphs (directed or undirected).
 - Dijkstra's algorithm. A greedy algorithm. $O((n+m)\log n)$ runtime.
- Case 3: Arbitrary weight without cycle in directed graphs.
 - Update in topological order. O(n + m) runtime.
- Case 4: Arbitrary weight without negative cycle in directed graphs.
 - Bellman-Ford algorithm. $\Theta(nm)$ runtime, can detect negative cycle.
- The shortest path problem has *optimal substructure* property.
- Update is a <u>safe</u> and <u>helpful</u> operation.

Reading

- [DPV] Ch.4 (More intuitive presentation.)
- [CLRS] Ch.24 (excluding 24.4) (Formal and rigorous.)
- Optional reading: [Erickson v1] Ch.8

