

Unweighted Graphs

Wishful Thinking – you decide to go to work on your sun-tan in 'Hatzuk' beach in Tel-Aviv. Therefore, you take your swimming suit and Tel-Aviv's bus trails map , and go on bus 405 to the central station of Tel-Aviv. The bus ride inside the city costs 1 nis per station.
How will you find the cheapest way from the central station to the beach?

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Unweighted Graphs - cont.

- **BFS** finds shortest paths from a single source (i.e the central station) in an unweighted graph.
- How much will you pay?
- n nis, when n is the number of stations you
- passed in your way.
- •Can you think of an algorithm that finds a single shortest **path**, and always works better
- then BFS?
- •No such algorithm is known.

Weighted Graphs

- Now assume that you pay for the bus ride between stations according to the distance between the stations. That is every 'edge' is the bus trails map has a different price (= weight).
- Total payment = sum over the costs between the stations on the way.
- Will BFS work?
- No BFS counts the number of edges on the path, but does not refer to the edges weights.

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Other Versions of Shortest Paths

- If we know how to find shortest paths from a single source, we can also find:
- A shortest path between a pair of vertices
- The shortest paths from all vertices to a single source. How?
- By reversing the direction of the edges in the graph
- Shortest paths between every pair of vertices. How?
- By running the sssp from every vertex (there are more efficient solutions)

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Observations

Observation 1: When can negative weights become a problem?

- When there is a circle with negative weight, reduce the weight of the path by repeating the circle over and over.
- Solution: either require non-negative weights, or identify and report circles with negative weights

Observations – cont.

Observation 2: Let $u \sim y \sim z \sim v$ be the shortest path between u and v. Is $y \sim z$ optimal, too?

- Yes! If there was a shorter path between y and z, we could use it to shorten the path between u and v → contradiction
- This property suggests that we can use greedy algorithms to find the shortest path

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Relaxation

The Idea: Build a shortest paths tree rooted in S. For every vertex, v, keep a value, d[v], of the shortest path from s that is currently known.

The general algorithm scheme:

- Initialization: $d[v] = \infty$, d[s] = 0

- In every iteration of the algorithm we check if we can do **relaxation** that is, find a shorter path from s to a vertex v then the path currently known.
 We will learn two algorithms:
- Dijkstra all weights are non-negative
- Belman-Ford identifies circles with negative weight

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- Dijkstra
- The idea: Maintain a set of vertices whose final shortest path weights from *s* have already been determined
- How will we choose the next vertex to add to the set?
- We will take the vertex *u* with the minimal *d* value. This is also the 'real' value of the shortest path.
- What relaxation can be done?
- We can check all the edges leaving *u*.



Bellman-Ford's Algorithm

- Now we want to identify negative circles.
- Assume that every iteration we do relaxation from all the edges. What edges might be relaxed on iteration i?
- The edges that have a path with I edges from s (shorter paths – updated in previous iterations)
- What is the maximum number of edges in any shortest path from s?

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Bellman-Ford's Algorithm

- So how many iterations are needed if there are no negative circles?
- |V| 1. After the |V| 1 iteration, all *d* values will be the lengths of the shortest paths.
- And how can we identify negative circles?
- By running the |V| iteration if we can find a relaxation, then there is a negative circle in the graph

Bellman-Ford's Algorithm

Initialize(G, s) for $i \leftarrow 1$ to |V| - 1for each edge $(u, v) \in E[G]$ do Relax(u, v, w)for each edge $(u, v) \in E[G]$ if d[v] > d[u] + w(u, v) return false return true

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Bellman-Ford Run Time

- The algorithm run time is:
 - Goes over v-1 vertexes, O(V)
 - For each vertex relaxation over E, O(E)
 - Altogether O(VE)

9482 Application of Bellman-Ford • Linear programming problems – find an optimal value (min / max) of a linear function of n variables, with linear contraints: $-x_2 \leq 0$ • A special case: $\underset{x_2 - x_5 \leq 1}{\underset{x_3 - x_1 \leq 5}{\underset{x_4 - x_1 \leq 4}{\underset{x_5 - x_3 \leq -1}{\underset{x_5 - x_3 \leq -3}{\underset{x_5 - x_4 < -3}{\underset{x_5 - x_4 }}}}}}}}}}$





Application of Bellman-Ford

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- What is the connection between the Bellman-Ford algorithm and a set of linear inequalities?
- We can interpret the problem as a directed graph.
- The graph is called the *constraint graph* of the problem
- After constructing the graph, we could use the Bellman-Ford algorithm.
- The result of the Bellman-Ford algorithm is the vector *x* that solves the set of inequalities.





Application of Bellman-Ford

- How can we extend the *constraint graph to a* single-source-shortest-paths problem?
- By adding vertex v_0 that directs at all the other vertices.
- Weight all edges from v_0 as zero.
- The weight of the other edges is determined by the inequality constraints.









• Formally:
$$x_i := \delta(v_0, v_i)$$

• Why is this correct?





Application of Bellman-Ford

• The left side sums to 0.

$$\sum_{i=1}^{k-1} (x_{i+1} - x_i) = x_k - x_1 = 0$$

- The right side sums to the cycle's weight w(c)
- We get $0 \le w(c)$
- But we assumed the cycle was negative...