Shortest Paths, and Dijkstra's Algorithm: Overview

- Graphs with lengths/weights/costs on edges.
- Shortest paths in edge-weighted graphs
- Dijksta's classic algorithm for computing single-source shortest paths.

S Q A

3

< 回 > < 三 > < 三 > 、

Graphs with edge "length" (or "weight/cost")

An edge-weighted directed graph, G = (V, E, w), has a length/weight/cost function, $w : E \to \mathbb{N}$, which maps each edge $(u, v) \in E$ to a non-negative integer "length" (or "weight", or "cost"): $w(u, v) \in \mathbb{N}$.

We can **extend** the "length" function w to a function $w : V \times V \rightarrow \mathbb{N} \cup \{\infty\}$, by letting w(u, u) = 0, for all $u \in V$, and letting $w(u, v) = \infty$ for all $(u, v) \notin E$.

Consider a directed path:

 $X_0 e_1 X_1 e_2 \ldots e_n X_n$

from $u = x_0 \in V$ to $v = x_n \in V$, in graph G = (V, E, w). The **length** of this path is defined to be: $\sum_{i=1}^{n} w(x_{i-1}, x_i)$.

SQ A

Graphs with edge "length" (or "weight/cost")

An edge-weighted directed graph, G = (V, E, w), has a length/weight/cost function, $w : E \to \mathbb{N}$, which maps each edge $(u, v) \in E$ to a non-negative integer "length" (or "weight", or "cost"): $w(u, v) \in \mathbb{N}$.

We can **extend** the "length" function w to a function $w : V \times V \rightarrow \mathbb{N} \cup \{\infty\}$, by letting w(u, u) = 0, for all $u \in V$, and letting $w(u, v) = \infty$ for all $(u, v) \notin E$.

Consider a directed path:

 $X_0 e_1 X_1 e_2 \ldots e_n X_n$

from $u = x_0 \in V$ to $v = x_n \in V$, in graph G = (V, E, w). The **length** of this path is defined to be: $\sum_{i=1}^{n} w(x_{i-1}, x_i)$.

Question: Given G and a pair of vertices $u, v \in V$, how do we compute the length of the **shortest path** from u to v?

Dijkstra's single-sources shortest-path algorithm

Input: Edge-weighted graph, G = (V, E, w), with (*extended*) weight function $w : V \times V \rightarrow \mathbb{N}$, and a source vertex $s \in V$.

Output: Function $L : V \to \mathbb{N} \cup \{\infty\}$, such that for all $v \in V$, L(v) is the length of the shortest path from *s* to *v* in *G*.

Algorithm:

Initialize: $S := \{s\}; L(s) := 0;$ Initialize: L(v) := w(s, v), for all $v \in V - \{s\}$; while $(S \neq V)$ do $u := \arg \min_{z \in V-S} \{L(z)\}$ $S := S \cup \{u\}$ for all $v \in V - S$ such that $(u, v) \in E$ do $L(v) := \min\{L(v), L(u) + w(u, v)\}$ end for end while Output function $L(\cdot)$. <ロ > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < 回 > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ >

SQ A

Why does Dijkstra's algorithm work?

Claim: The While loop of Dijkstra's algorithm maintains the following **invariant** properties of the function *L* and the set *S*:

- **1.** $\forall v \in S, L(v)$ is the shortest path length from *s* to *v* in *G*.
- 2. $\forall v \in V S$, L(v) is the length of the shortest path from *s* to *v* which uses only vertices in $S \cup \{v\}$.
- **3**. For all $u \in S$ and $v \in V S$, $L(u) \leq L(v)$.

Note that the three invariants hold after initialization, just prior to the first iteration of the while loop.

The claim follows once we prove (on board) that **if** the invariants hold just prior to a while loop iteration **then** they hold just after.

Since each iteration adds one vertex to *S*, it follows that the algorithm halts, at which point S = V, and thus, by invariant (1.), the function $L : V \to \mathbb{N} \cup \{\infty\}$ is the correct answer.

SQ Q

3

. ₹ 📑 🕨 ...

Remarks on Dijkstra's Algorithm

- If Dijkstra's algorithm is implemented naively, it has running time $O(n^2)$, where n = |V|.
- With clever data structures (e.g., so called "Fibbonacci Heaps") Dijkstra's algorithm can be implemented much more efficiently: essentially in time $O(m + n \log n)$ where, n = |V| and m = |E|.

This increased efficiency can make a **big difference** on huge "sparse" graphs where *n* is much smaller than *m*.

 Dijkstra's algorithm can be augmented to also output a description of a shortest path from the source vertex s to every other vertex v.

We will not describe these extensions, and we will certainly not assume that you know them.

Kousha Etessami (U. of Edinburgh, UK)

Graph Colouring

SQ (V

E

Graph Colouring

Suppose we have *k* distinct colours with which to colour the vertices of a graph. Let $[k] = \{1, ..., k\}$. For an undirected graph, G = (V, E), an admissible vertex *k*-colouring of *G* is a function $c : V \to [k]$, such that for all $u, v \in V$, if $\{u, v\} \in E$ then $c(u) \neq c(v)$.

For an integer $k \ge 1$, we say an undirected graph G = (V, E) is *k*-colourable if there exists a *k*-colouring of *G*.

The chromatic number of G, denoted $\chi(G)$, is the smallest positive integer k, such that G is k-colourable.

500

Some observations about Graph colouring

- Note that any graph G with n vertices in n-colourable.
- The *n*-Clique, *K_n*, i.e., the complete graph on *n* vertices, has chromatic number $\chi(K_n) = n$. All its vertices must get assigned different colours in any admissible colouring.
- The clique number, $\omega(G)$, of a graph G is the maximum positive integer r > 1, such that K_r is a subgraph of G.
- Note that for all graphs G, $\omega(G) \leq \chi(G)$: if G has an *r*-clique then it is not (r - 1)-colorable.
- However, in general, $\omega(G) \neq \chi(G)$. For instance, The 5-cycle, C_5 , has $\omega(C_5) = 2 < \chi(C_5) = 3$.

(日)

More observations about colouring

- As already mentioned, any bipartite graph is 2-colourable. Indeed, that is an equivalent definition of being bipartite.
- More generally, a graph G is k-colourable precisely if it is k-partite, meaning its vertices can be partitioned into k disjoint sets such that all edges of the graph are between nodes in different parts.

S Q A

▲□ ▶ ▲ □ ▶ ▲ □ ▶ □

Algorithms/complexity of colouring graphs

To determine whether a *n*-vertex graph G = (V, E) is *k*-colourable by "*brute force*", we could try all possible colourings of *n* nodes with *k* colours.

Difficulty: There are k^n such k-colouring functions $c: V \rightarrow [k]$.

Question: Is there an efficient (polynomial time) algorithm for determining whether a given graph *G* is *k*-colourable?

SQ Q

(口)

Algorithms/complexity of colouring graphs

To determine whether a *n*-vertex graph G = (V, E) is *k*-colourable by "*brute force*", we could try all possible colourings of *n* nodes with *k* colours.

Difficulty: There are k^n such k-colouring functions $c: V \rightarrow [k]$.

Question: Is there an efficient (polynomial time) algorithm for determining whether a given graph *G* is *k*-colourable?

Answer: No, no generally efficient (polynomial time) algorithm is known, and even the problem of determining whether a given graph is 3-colourable is **NP-complete**. (Even approximating the chromatic number of a given graph is NP-hard.)

In practice, there are hueristic algorithms that do obtain good colourings for many classes of graphs.

SQ Q

(日)

Applications of Graph Colouring (many) Final Exam Scheduling

- There are *n* courses, $\{1, \ldots, n\}$.
- Some courses have the same students registered for both, so their exams can't be scheduled at the same time.
- Let $G = (\{1, ..., n\}, E)$ be a graph such that $\{i, j\} \in E$ if and only if $i \neq j$ and courses *i* and *j* have a student in common.
- Question: What is the minimum number of exam time slots needed to schedule all n exams?
- Answer: This is precisely the chromatic number $\chi(G)$ of G.

Furthermore, a k-colouring of G yields an *admissible* schedule of exams into k time slots, allowing all students to attend all their exams, as long as different "colors" are scheduled in disjoint time slots.