# Algorithms II Cheat-Sheet

### Notation

```
A \in [10] \equiv A \in [1..10]
{a, b c} is a set of vertices
G\{a, b, c\} is a graph
\overrightarrow{x} just represents a vector x
```

# Big O

Notation	Intuitive meaning	Analogue
$f(n) \in O(g(n))$	f grows at most as fast as $g$	<u> </u>
$f(n) \in \Omega(g(n))$	f grows at least as fast as $g$	≥
$f(n) \in \Theta(g(n))$	f at the same rate as $g$	=
$f(n) \in o(g(n))$	f grows strictly less fast than $g$	<
$f(n) \in \omega(g(n))$	f grows strictly faster than $g$	>

Notation	Formal definition
$f(n) \in O(g(n))$	$\exists C, n_0 \colon \forall n \geq n_0 \colon f(n) \leq C \cdot g(n)$
$f(n) \in \Omega(g(n))$	$\exists c, n_0 \colon \forall n \geq n_0 \colon f(n) \geq c \cdot g(n)$
$f(n) \in \Theta(g(n))$	$\exists c, C, n_0 : \forall n \geq n_0 : c \cdot g(n) \leq f(n) \leq C \cdot g(n)$
$f(n) \in o(g(n))$	$\forall C \colon \exists n_0 \colon \forall n \geq n_0 \colon f(n) \leq C \cdot g(n)$
$f(n) \in \omega(g(n))$	$\forall c \colon \exists n_0 \colon \forall n \geq n_0 \colon f(n) \geq c \cdot g(n)$

## Interval Scheduling

A **request** is a pair of integers (s, f) with  $0 \le s \le f$ . We call s the start time and f the finish time.

A set A of requests is **compatible** if for all distinct (s, f),  $(s', f') \in A$ , either  $s' \ge f$  or  $s \ge f'$  — that is, the requests' time intervals don't overlap.

#### **Interval Scheduling Problem**

**Input:** An array  $\mathcal{R}$  of n requests  $(s_1, f_1), \ldots, (s_n, f_n)$ .

**Desired Output:** A compatible subset of  $\mathcal{R}$  of maximum possible size.

#### Algorithm: GREEDYSCHEDULE **Input**: An array $\mathcal{R}$ of n requests. **Output**: A maximum compatible subset of $\mathcal{R}$ . 1 begin Sort $\mathcal{R}$ 's entries so that $\mathcal{R} \leftarrow [(s_1, f_1), \dots, (s_n, f_n)]$ where $f_1 \leq \dots \leq f_n$ . Initialise $A \leftarrow []$ , lastf $\leftarrow 0$ . foreach $i \in \{1, \ldots, n\}$ do if $s_i \ge last f$ then Append $(s_i, f_i)$ to A and update lastf $\leftarrow f_i$ . Return A.

# Complexity:

Step 2 takes O(n log n)

Steps 3–6 all take O(1) time and are executed at most n times.

 $\therefore$  total running time = O(nlog n) + O(n)O(1) =O(nlogn).

# Interval Scheduling

### Formal GreedySchedule:

 $A^{+} := \operatorname{argmin} \{f : (s, f) \in R, A \cup \{(s, f)\} \text{ is }$ compatible for all  $A \subseteq R$ ,  $A_{i+1} := A_i \cup \{A_i^+\}$  $A_0 := \emptyset$ ,

 $t := \max\{i: A_i \text{ is defined}\}\$ 

## **Interval Scheduling Proofs**

**Lemma**: Greedy Schedule always outupts  $A_t$ 

**Proof**: By induction form the following loop invariant. At the start of the i'th iteration of 4-7:

- A is equal to  $A_t \cap \{(s_1, f_1), ..., (s_{i-1}, f_{i-1})\}$
- last is equal to the latest finish time of any request in A (or 0 if A = [])

**Lemma**:  $A_t$  is a compatible set

**Proof**: Instant by induction;  $A_0$  is compatible, and if  $A_i$  is compatible then so is  $A_{i+1} = A_i \cup A^+$  by the definition of  $A_i^+$ 

**Lemma**:  $A_t$  is a maximum compatible subset of the Array R (look in pseudocode)

#### Proof:

Base case for i = 1:  $A_0^+$  is the fastest finishing request in R by definition

Inductive step: Suppose  $A_i$  finishes faster than  $B_i$ . Let  $B_i^+$  be the (i+1)'st fastetst-finishing element of

B. Since  $A_i$  finishes faster than  $B_i$ ,  $A_i \cup \{B_i^+\}$  is compatible. Hence by definition,  $A_i^+$  exists and finishes no later than  $B^+$ 

**Theorem:** GreedySchedule outputs  $A_t$ , which is a maximum compatible set.

**Proof**: putting all of the above proofs together, we prove the theorem.

## Graph Theory

**Graph**: G = (V, E)

**Edge**: E = E(G) is a set of edges contained in

 $\{\{u,v\}: u,v \in V, u \neq v\}$ **Vertex**: V = V(G) is a set of vertices

**Subgraph**:  $H = (V_H, E_H)$  of G is a graph with  $V_H \subseteq V$  and  $E_H \subseteq E$ 

Induced Subgraph: is a subgraph if  $E_H = \{e \in E : e \subseteq V_H\}$ 

Component: H of G is a maximal connected induced subgraph of G.

**Degree**: d(v) = |N(v)|

**Neighbourhood**:  $N(v) = \{w \in V : \{v, w\} \in E\}$ 

Walk: sequence of vertices  $v_0...v_k$  such that  $\{v_i, v_{i+1}\} \in E \text{ for all } i \leq k-1$ 

**Length**: the value of k (see above walk definition)

Euler Walk: a walk that contains every edge in G exactly once.

**Isomorphism**: two graphs are isomorphic if there is a bijection f:  $V_1 \to V_2$  such that  $\{f(u), f(v)\} \in E_2$  if and only if  $\{u, v\} \in E_1$ 

**Path**: is a walk in which no vertices repeat

**Connected**: A graph is connected if any two vertices are joined by a path

**Digraph**: is a pair G = (V, E), V is a set of vertices and E is a set of edges contained in  $\{(u,v): u,v \in V, u \neq v\}$ 

Strongly connected: G is .. if for all  $u, v \in V$ , there is a path from u to v and a path from v to u. Weakly connected:

In-Neighbourhood:  $N^-(v) = \{u \in V(G) : (u,v) \in$ E(G)

Out-Neighbourhood:  $N^+(v) = \{w \in V(G) : v \in V(G) : v$  $(v,w) \in E(G)$ 

Cycle: is a walk  $W = w_0...w_k$  with  $w_0 = w_k$  and  $k \geq 3$ , in which every vertex appears at most once except for  $w_0$  and  $w_k$  (which appear twice)

**Hamilton cycle**: is a cycle containing every vertex in the graph

**k-regular**: a graph is .. if every vertex has degree k Bijection:

### Graph Theory

**Theorem**: If G has an Euler walk, then either:

- every vertex of G has even degree; or
- all but two vertices  $v_0$  and  $v_k$  have even degree, and any euler walk must have  $v_0$  and  $v_k$  as endpoints

**Theorem**: let G = (V, E) be a digraph with no isolated vertices, and let  $U, v \in V$ . Then G has an Euler walk from u to v if and only if G is weakly connected and either:

- u = v and every vertex of G has equal in- and out-degrees; order
- $u \neq v, d^+(u) = d^-(u) + 1, d^-(v) = d^+(v) + 1$ and every other vertex of G has equal in- and out-degrees

**Dirac's Theorem:** Let  $n \geq 3$ . Then any n-vertex graph G with minimum degree at least  $\frac{n}{2}$  has a Hamilton cycle.

Handshake lemma: For any graph

$$G = (V, E), \sum_{v \in V} d(v) = 2|E|$$

**Proof**: All edges contain two vertices, and each vertex v is in d(v) edges. Count the number of verted-edge pairs: Let  $X = \{(v,e) \in V \times E : v \in E\}$ . Then |X| = 2|E| and  $|X| = \sum_{v \in V} d(v)$ , so we're done. **Directed Handshake lemma**: For any graph

Directed Handshake lemma: For any graph  $G = (V, E), \sum_{v \in V} d^+(v) = \sum_{v \in V} d^-(v) = 2|E|$ 

**Proof**: TODO. Instead of counting vertex-edge pairs, we count tail-edge pairs. Each edge has one tail so |X| = |E|

#### Trees

Forest: a graph with no cycles

Tree: a forest that is connected

**Root**: for T = (V, E). Root  $r \in V$  as follows. For all vertices  $v \neq r$ , let  $P_v$  be the unique path from r to v. Then direct each  $P_v$  from r to v.

 $\mathbf{Leaf}\colon$  is a degree-1 vertex. Root cannot be a leaf.

**Ancestor**: u is an .. of v if u i on  $P_v$ **Parent**: u is the .. of v if  $u \in N^-(v)$ 

**level**: first ..  $L_0$  of T is r, and  $L_{i+1} = N^+(L_i)$ .

**depth**: of T is max{i:  $L_i \neq \emptyset$ }. Root doesn't count

**Lemma 1**: If T = (V, E) is a tree, then any pair of vertices  $u, v \in V$  is joined by a unique path uTv in T.

Lemma 2: Any n-vertex tree has n-1 edges

**Lemma 3**: Any n-vertex tree T = (V, E) with  $n \ge 2$  has at least 2 leaves

# Tree Properties:

**A**: T is connected and has no cycles

B: T has n-1 edges and is connected

C: T has n-1 edges and has no cycles

D: T has a unique path between any pair of vertices

 $A \Longrightarrow B, C, D$   $A \Longleftarrow B, C, D$ .

# Search and Dijkstra

# Adjacency Matrix:

Storing:  $\Theta(|V|^2)$  space

Adjacency query:  $\Theta(1)$  time

Neighbourhood query:  $\Theta(|V|)$  time

# Adjacency List:

$$\boxed{s} \rightarrow b, a$$

$$a \rightarrow s, c$$

Storing:  $\Theta(|V| + |E|)$  space Adjacency query:  $\Theta(d^+(u))$  time Neighbourhood query:  $\Theta(d^+(u))$  time

## Linear Programming

**Feasible**: We say  $\overrightarrow{x} \in \mathbb{R}^n$  is a feasible solution if  $\overrightarrow{x} > \overrightarrow{0}$  and  $A\overrightarrow{x} < \overrightarrow{b}$ .

**Optimal:** We say  $\overrightarrow{x}$  is an optimal solution if  $f(\overrightarrow{y}) \leq f(\overrightarrow{x})$  for all feasible  $y \in \mathbb{R}^n$ 

 ${\bf Polytope}:$  is a geometric object with flat sides

Corollary: There will always be an optimal solution

### Non-Standard Form:

$$-4x + 5y - z \rightarrow \text{max subject to}$$

$$x + y + z \le 5;$$

$$x + y + z \ge 5;$$

$$x + 2y \ge 2;$$

$$x, z \geq 0$$
.

### **Standard Form:**

$$-4x + 5(y_1 - y_2) - z \rightarrow \text{max subject to}$$

$$x + (y_1 - y_2) + z \le 5;$$

$$-x - (y_1 - y_2) - z \le -5;$$

$$-x - 2(y_1 - y_2) \le -2;$$

$$x, y_1, y_2, z \ge 0.$$

### Matrix Form:

$$-4x + 5y_1 - 5y_2 - z \rightarrow \text{max subject to}$$

$$\begin{pmatrix} 1 & 1 & -1 & 1 \\ -1 & -1 & 1 & -1 \\ -1 & -2 & 2 & 0 \end{pmatrix} \begin{pmatrix} x \\ y_1 \\ y_2 \\ z \end{pmatrix} \le \begin{pmatrix} 5 \\ -5 \\ -2 \end{pmatrix};$$

$$x, y_1, y_2, x \geq 0.$$

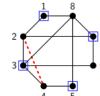
**Simplex Method**: Search greedily for a vertex of the feasible polytope which maximises the objective function

Worst case: hypercube which has  $\Omega(2^n)$  vertices.

In practice it only need  $\Theta(n)$  steps

**Vertex Cover:** in a graph G, is a set  $X \subseteq V$  such that every edge in E has at least one vertex in X

We can express finding a minimum vertex cover as solving a linear program in which the solutions must be integers: an integer linear program.



$$\sum_{\nu} x_{\nu} \rightarrow \text{min subject to}$$

$$x_{u} + x_{\nu} \geq 1 \text{ for all } \{u, v\} \in E;$$

$$x_{\nu} \leq 1 \text{ for all } v \in V;$$

$$x_{\nu} \geq 0 \text{ for all } v \in V;$$

$$x_{\nu} \in \mathbb{N} \text{ for all } v \in V.$$

 $X = \{1, 3, 5, 7\}$  is **not** a vertex cover.

Here we have  $x_1 = x_3 = x_5 = x_7 = 1$  and  $x_0 = x_2 = x_4 = x_6 = 0$ .

The uncovered edge  $\{2,4\}$  corresponds to the constraint  $x_2+x_4\geq 1$ , which is violated.

#### Flow Networks

the flow, v(f)

Flow Network: consists of a directed graph G = (V, E), a capacity function  $c : E \to \mathbb{N}$ , a source vertex  $s \in V$  with  $N^-(s) = \emptyset$ , and a sink vertex  $t \in V$  with  $N^+(t) = \emptyset$ 

**Flow**: is a function in (G, c, s, t)  $f: E \to \mathbb{R}$  with properties:

- No edge has more flow than capacity; formally, for all  $e \in E, 0 \le f(e) \le c(e)$
- Flow is conserved at vertices; flow in = flow out **Maximum Flow**: a flow f maximising the value of

**Cut**: is any pair of disjoint edges  $A, B \subseteq V$  with  $A \cup B = V$ ,  $s \in A$  and  $t \in B$ .

**Lemma 1:** For all sets  $X \subseteq V$   $\{s,t\}$ , we have  $f^+(X) = f^-(X)$ . So flow is conserved in sets/cuts as well as vertices

**Proof**: By summing conservation of flow over all  $v \in X$ :

 $\sum_{v \in X} \sum_{u \in N^-(v)} f(u, v) = \sum_{v \in X} \sum_{w \in N^+(v)} f(v, w)$ . For all  $e \subseteq X$ , f(e) appears once on each side; after cancelling those terms we're left with  $f^+(X) = f^-(X)$ .

**Lemma 2:** For all cuts (A, B),  $f^+(A) - f^-(A) = f^-(B) - f^+(B)$ .

**Proof**: We have shown that  $v(f) = f^+(A) - f^-(A)$  because A and B are disjoint and  $A \cup B = V$ .

**Lemma 3**: Push(G, c,s,t, f, P) returns a new flow f', with value v(f') = v(f) + C in O(|V(G)|) time

#### Ford-Fulkerson:

**Input**: A (weakly connected) flow network (G, c, s, t).

**Output:** A flow f with no augmenting paths.

1 begin

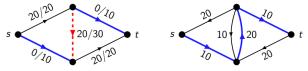
- Construct the flow f with f(e) = 0 for all  $e \in E(G)$ .
- Construct the residual graph  $G_f$ .
  - **while**  $G_f$  contains a path P from s to t **do**
- Find *P* using depth-first (or breadth-first) search.
- 6 Update  $f \leftarrow \text{Push}(G, c, s, t, f, P)$ .
- The Update  $G_f$  on the edges of P.
- Return f.

**Complexity**: Every step takes O(|E|) time or O(|V|) time, and since G is weakly connected we have |V| = O(|E|). So the running time is O(v(f\*)|E|).

#### Flow Networks

**Residual graph**:  $G_f$  of (G, c, s, t) on V(G) as follows:

- if flow i capacity: then forward edge with value capacity-flow
- if flow ; 0: add backward edge with value flow



Residual capacity of edge: max{capacity - flow, backward edge flow}

Residual capacity of network: minimum residual capacity of it's edges

## Augmenting Path:

**Max-flow min-cut theorem:** The value of a maximum flow is equal to the minimum capacity of a cut, i.e. the minimum value of  $c^+(A)$  over all cuts (A, B).

**Proof**: Let f be a maximum flow, and let (A, B) be a cut minimising  $c^+(A)$ . We already proved  $v(f) \leq c^+(A)$ . Moreover, there is no augmenting path for f, so exactly as before, there is a cut (A'.B') with  $c^+(A') = v(f)$ ; thus  $v(f) \geq c^+(A)$ . The result follows.