

# CS254 Algorithmic Graph Theory - Lecture Notes

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# Chapter 1

## Basics

### 1.1 Graphs

Lecture 1

**Definition.** A graph  $G$  is an ordered pair  $(V(G), E(G))$  consisting of a set  $V(G)$  of vertices and a set  $E(G)$  of edges. Each edge  $e \in E(G)$  is associated with a pair of vertices  $\{u, v\}$ . We will say that  $u, v$  are endpoints of  $e$ . We also say  $u, v$  are incident to  $e$  and vice versa.

**Definition.** Two vertices which are incident with a common edge are adjacent (or neighbours). Two or more edges with the same pair of endpoints are said to be parallel.

**Definition.** A simple graph is one without parallel edges.

**Notation.** By convention,  $G$  will denote a graph,  $n$  and  $m$  will be the number of vertices  $|V(G)|$  and the number of edges  $|E(G)|$  respectively.

### 1.2 Special graphs

**Definition.** A complete graph is a simple graph in which every two vertices are adjacent.

Such a graph on  $n$  vertices is denoted  $K_n$ .

We can see that  $|E(K_n)| = \frac{n(n-1)}{2}$ .

**Definition.** A graph  $G = (V, E)$  is called empty if  $E = \emptyset$ .

**Definition.** A graph is called bipartite if its vertex set can be partitioned into two subsets  $X$  and  $Y$  so that every edge has one endpoint in  $X$  and one in  $Y$ .

**Example.** The set of all (monogamous) people can be represented by a graph with edges joining married couples, this graph is bipartite.

We will call  $X, Y$  the parts of the graph. So a bipartite graph is often denoted as  $G = (X, Y, E)$ .

**Notation.** We write  $K_{m,n}$  for the complete bipartite graph with  $|X| = m$ ,  $|Y| = n$  and  $E$  containing all edges between  $X$  and  $Y$ .

**Definition.** The graph  $K_{1,n}$  is often called a star.

**Definition.** A path is a simple graph whose vertices can be rearranged into a linear sequence in such a way that two adjacent vertices are always consecutive in the sequence and vice versa, non-adjacent vertices are non-consecutive, such a graph on  $n$  vertices is often denoted  $P_n$ .

**Definition.** A cycle on  $n$  vertices is a graph defined in a similar way but a cyclic sequence of vertices.

**Notation.** The length of a path or cycle is the number of its edges. A path or cycle of length  $k$  is often called a  $k$ -path or  $k$ -cycle.

### 1.2.1 Adjacency and incidence matrices

**Definition.** The incidence matrix of  $G = (V, E)$  when  $|V| = m$  and  $|E| = n$  is the  $m \times n$  matrix  $M_G = (m_{ve})$  where  $m_{ve}$  is the number of times that vertex  $v$  and  $e$  are incident ( $m_{ve} \in \{0, 1\}$ ).

**Example.**

**Definition.** The adjacency matrix of  $G = (V, E)$  when  $|V| = m$  is the  $m \times m$  matrix  $A_G = (a_{uv})$  where  $a_{uv}$  is the number of edges joining  $u$  and  $v$ .

**Example.**

$$= A_{K_3}$$

**Definition.** An adjacency list is a list  $(N(v), v \in V)$  used to represent matrices,  $N(v)$  is the set of neighbours of  $v$  in arbitrary order.

### 1.2.2 Adjacency and incidence matrices

**Definition.** The degree of a vertex  $v$  in a graph  $G$  is denoted  $d_G(v)$  or  $d_v$  and is the number of edges of  $G$  incident with  $v$ .

**Theorem** (Handshake lemma). *For any graph*

$$\sum_{v \in V} d_v = 2|E| \quad (1.1)$$

**Theorem.** *In any graph, the number of vertices of odd degree is even.*

*Proof.* Consider the equation 1.1 module 2. We have degree of each vertex  $d_v \equiv 1$  if  $d_v$  is odd, or 0 if  $d_v$  is even. Therefore the left hand side of 1.1 is congruent to the number of vertices of odd degree and the RHS is 0. Therefore the number of vertices of odd degree is congruent to 0 mod 2.  $\square$

**Definition.** A graph  $G$  is called  $k$ -regular if  $d_G(v) = k$  for all  $v \in V(G)$ . A graph is regular if it is  $k$ -regular for some  $k$ .

**Example.**  $K_n$  is  $(n-1)$ -regular.  
 $C_n$  is 2-regular.  
 $P_n$  is not regular.

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**Theorem.** Let  $G = (X, Y, E)$  be a bipartite graph without isolated vertices s.t.  $d_x \geq d_y$  for all  $x \in X$  and  $y \in Y$  s.t.  $(x, y) \in E$ . Then  $|X| \leq |Y|$  with equality iff  $d_x = d_y \forall (x, y) \in E$ .

*Proof.* Take the adjacency matrix  $B$ . Define the matrix  $B'$  by dividing each row of  $B$  corresponding to vertex  $u$  by  $d_u$ . consider one column sum corresponding to  $y \in Y$ .

$$\sum_{x \in X} b'_{xy} = \sum_{(x,y) \in E} \frac{1}{d_x} = \frac{d_y}{d_x} \leq 1$$

Therefore  $|X| = \sum_{x \in X} \sum_{y \in Y, (x,y) \in E} \frac{1}{d_x}$

$$\begin{aligned} &= \sum_{x \in X, y \in Y} \sum_{(x,y) \in E} \frac{1}{d_x} \leq \sum_{x \in X, y \in Y} \sum_{(x,y) \in E} \frac{1}{d_y} \\ &= \sum_{y \in Y} \sum_{x \in X, (x,y) \in E} \frac{1}{d_y} = |Y| \end{aligned}$$

If  $|X| = |Y|$ , then the middle inequality must be an equality implying that  $d_x = d_y$  for all  $(x, y) \in E$ .  $\square$

### 1.2.3 Isomorphisms

**Definition.** Two graphs  $G$  and  $H$  are isomorphic, written  $G \cong H$ , if there is a bijection  $\Theta: V(G) \rightarrow V(H)$  such that  $(u, v) \in E$  iff  $(\Theta(u), \Theta(v)) \in E(H)$ .

**Example.**

**Remark.** • The mapping  $\Theta$  is not unique.

- Isomorphic graphs have the same number of vertices and edges.
- Up to isomorphism there is only one complete graph on  $n$  vertices denoted  $K_n$ . So  $K_n$  is a class of graphs that are all isomorphic to each other.
- The same as above goes for each of  $K_{m,n}$ ,  $P_n$ ,  $C_n$ .

### 1.2.4 Directed graphs

**Definitions.** In a graph  $G = (V, E)$  a *walk* from a starting vertex  $v_0$  to a destination vertex  $v_n$  is a sequence

$$W = \langle v_0, e_1, v_1, e_2, \dots, e_n, v_n \rangle \text{ s.t. } e_i = (v_{i-1}, v_i)$$

The *length* of the walk is the number of edges in the sequence  $W$ . A *closed walk* has the same starting and ending vertices. A *path* is a walk where every vertex appears at most once.

The *distance*  $d(s, t)$  between a vertex  $s$  and a vertex  $t$  in a graph  $G$  is the length of the shortest path between  $s$  and  $t$  (if one exists) and  $d(s, t) = +\infty$  otherwise. A *weighted graph*  $G = (V, E, \omega)$  is a graph and a function  $\omega: E \rightarrow \mathbb{R}$  assigning weights to edges.

We now define some problems that drive our study of algorithms in graph theory.

**Problem 1** Given an edge weighted graph  $G = (V, E, \omega)$ . Find a shortest path between  $s$  and  $t$  (length is the sum of the edge weights).

**Problem 2** Minimum-weight spanning tree: Given an undirected graph  $G = (V, E, \omega)$  find an acyclic connected subgraph of  $G$  of smallest weight.

**Problem 3** Travelling salesman problem: We are given a complete weighted graph  $G$ . Find a cycle that visits every vertex exactly once and minimises its length.

**Problem 4** Maximum flow problem: You have a directed graph  $G$  representing a system of pipes, each pipe has a capacity. We are given also the source  $S$  and the sink  $T$ ,  $S, T \in V(G)$ . How much water can we pump through the network per time unit?

**Definition.** A *leaf* is a vertex of degree one.

**Proposition.** Every tree  $T$  has at least two leaves.

*Proof.* Consider  $P = \langle v_1, \dots, v_k \rangle$  a path of maximum length in  $T$ . Consider  $v_1$  if  $v_1$  has degree more than one then we can either extend the path or  $T$  is not acyclic.  $\square$

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**Corollary.** If the degree of each node in  $G$  is  $\geq 2$  then  $G$  has a cycle.

**Proposition.** Each tree has exactly  $|V| - 1$  edges.

*Proof.* [By induction]  $\square$

**Definitions.** An acyclic graph is called a *forest*.

A *connected component* of  $G$  is a maximal connected subgraph of  $G$ . Let  $c(G)$  be the number of connected components of  $G$ .

**Corollary.** Any graph on  $n$  vertices has at least  $n - c(G)$  edges.

**Proposition.** If  $G$  is a simple graph with  $n$  vertices then

$$|E(G)| \leq \frac{(n-k)(n-k+1)}{2} \text{ where } k = c(G)$$

*Proof.* Let  $c_1, \dots, c_k$  be the connected components of  $G$ , i.e.  $c_i = (V_i, E_i)$  and  $E(G) = \bigcup_{i=1}^k E_i$ ,  $V = \bigcup_{i=1}^k V_i$ ,  $n = |V|$ . Then  $n = \sum_{i=1}^k n_i$  and  $n_i \geq 1$ . Obviously  $|E(G_i)| \leq \frac{n_i^2 - n_i}{2}$ . Thus  $|E(G)| \leq \sum_{i=1}^k \frac{n_i^2 - n_i}{2}$ . The maximum of RHS occurs when  $|C_1| = |C_2| = \dots = |C_{k-1}| = 1$  and  $|C_k| = n - k + 1$ .

Assume the contrary that the maximum occurs when  $c_i = k$ , and  $c_j = k_s$  where  $r \geq s \geq 2$ . Then  $|E(c_i)| + |E(c_j)| = \frac{r^2 + s^2 - r - s}{2}$ . But is  $c_i = k_{r+1}$  and  $c_j = k_{s-1}$ , then we have  $\frac{(r+1)^2 + (s-1)^2 - (r+1) - (s-1)}{2}$  edges. We get a contradiction since  $(r+1)^2 + (s-1)^2 > r^2 + s^2$ .  $\square$

**Definition.** A *cut-edge* is an edge whose removal increases the number of connected components.

**Theorem.** Let  $T$  be a graph with  $n$  vertices, then the following statements are equivalent:

1.  $T$  is a tree.
2.  $T$  contains no cycles and has  $(n - 1)$  edges.
3.  $T$  is connected and has  $(n - 1)$  edges.
4.  $T$  is connected and every edge is a cut-edge.
5. Any two vertices of  $T$  are connected by exactly one path.
6.  $T$  contains no cycles and for any new edges the graph  $T + e$  has exactly one cycle.

*Proof.* (1  $\implies$  2) [Previous proposition] and definition of tree.

(2  $\implies$  3) Suppose  $T$  has  $k$  connected components. Then each component must be a tree (because they have no cycles) therefore  $T$  must contain  $n - k$  edges, exactly  $\implies k = 1$ .

(3  $\implies$  4) Let  $e$  be an edges of  $T$ . Since  $T \setminus e$  has  $n - 2$  edges by one of the corollaries, we must have  $c(T) \geq 2$ .

(4  $\implies$  5) (By contradiction) Suppose  $x$  and  $y$  are two vertices of tree  $T$  that have two different paths, say  $p_1$  and  $p_2$  connecting them. Let  $u$  be the first vertex where the paths diverge and let  $v$  be the first vertex where the paths meet again. Any edge of the cycle containing  $u$  and  $v$  is not a cut edge which contradicts (4).

(5  $\implies$  6)  $T$  contains no cycles otherwise we would have two paths between  $x, y \in \text{cycle}$ . If we add an edge  $e = (u, v)$  we will create a cycle since  $T$  contains a path between  $u$  and  $v$ . If  $T + e$  contains two cycles then both cycles must contain  $e$ .  $u, v$  have two different paths containing them in  $T$  is a contradiction.  
 (6  $\implies$  1) [By definition ] □

### 1.2.5 Huffman Trees and Optimal Prefix Codes

**Definitions.** A *binary code* is an assignment of symbols or other meanings to a set of bitstrings.

A *prefix code* is a binary code with the property that no codeword is an initial substring of any other codeword.

This allows for transmission without confusion.

**Example.** A prefix code for an alphabet: construct a binary tree with a leaf corresponding to a letter of your alphabet, every left-hand edge is labelled by 0 and right-hand edge is labelled by 1.

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**Definitions.** Assume we have a code. Each codeword has a length  $l_c$  and each symbol has its frequency  $p_c$ . The *code efficiency* is then  $\sum p_c l_c$ . Let  $T$  be a binary tree with leaves  $s_1, \dots, s_l$  such that each leaf  $s_i$  is assigned weight  $w_i$ .

The *average weighted depth* of the binary tree  $T$  is  $\text{wt}(T) = \sum_{i=1}^l \text{depth}(s_i) w_i$  where depth is the number of edges to the root.

**Algorithm** (Constructing Huffman Prefix Codes).

**Input**  $s_1, \dots, s_l$  of symbols and a list of  $w_1, \dots, w_l$  of weights  $w_i$  corresponding to  $s_i$ .

**Output** A binary tree representing a prefix code for a set of symbols whose codewords have minimum average weighted length.

Initialise  $F$  to be a forest of isolated vertices, labelled  $s_1, \dots, s_l$  with respective weights  $w_1, \dots, w_l$ .

For  $i = 1$  to  $l - 1$  Choose two trees  $T$  and  $T'$  from the forest of smallest weights. Create a binary tree by adding a new vertex with  $T$  as a child on the left and  $T'$  a child on the right. Label the edges to these new child trees 0 and 1 respectively. Assign to the new tree the weight  $\text{wt}(T) + \text{wt}(T')$ .

Replace  $T$  and  $T'$  in  $F$  by the new tree.

When finished return  $F$ .

**Example.**

**Lemma.** If the leaves of a binary tree are assigned weights and if each internal vertex (node) is assigned a weight equal to the sum of its children's weights, then the tree's average weighted depth equals the sum of the weights of its internal (non-leaf) vertices.



*Proof.* In assignments.  $\square$

**Theorem.** For a given list of weights  $w_1, \dots, w_l$  a Huffman tree has the smallest average weighted depth among all binary trees whose leaves are assigned these weights.

*Proof.* Induction on  $l$ . Let  $T$  be the Huffman tree. **Base case**  $l = 2$ . Then  $\text{wt}(T) = w_1 + w_2$ , because  $T$  is the only tree on 3 vertices. **Inductive step** Assume that for some  $l \geq 2$  the Huffman algorithm produces a Huffman tree of minimum average weighted depth for any list of  $l$  weights.

We are given  $w_1 \leq w_2 \leq \dots \leq w_{l+1}$ . The Huffman tree  $H$  that is constructed by the algorithm consists of the Huffman tree  $\bar{H}$  for the weights  $w_1 + w_2, w_3, \dots, w_{l+1}$  where the leaf with weight  $w_1 + w_2$  is replaced by a tree with two children of weights  $w_1$  and  $w_2$ . By the previous lemma, we know  $\text{wt}(H) = \text{wt}(\bar{H}) + w_1 + w_2$ . By the inductive hypothesis,  $\bar{H}$  is optimal among all binary trees whose leaves are assigned weights  $w_1 + w_2, w_3, \dots, w_{l+1}$ .

Suppose  $T^*$  is an optimal binary tree for the weights  $w_1, \dots, w_l$ . Let  $x$  be an internal vertex of  $T^*$  of greatest depth and suppose  $y$  and  $z$  are its left child and right child respectively. Without loss of generality,  $y$  and  $z$  have weights  $w_1$  and  $w_2$  otherwise we can swap their weights with  $w_1$  and  $w_2$  and produce a tree with the same or better weights. Delete  $y$  and  $z$  from  $T^*$  and call this tree  $\bar{T}$ , we have  $\text{wt}(T^*) = \text{wt}(\bar{T}) + w_1 + w_2$ . We also know that  $\text{wt}(\bar{T}) \geq \text{wt}(\bar{H}) \implies \text{wt}(T^*) \geq \text{wt}(H)$ .  $\square$

Lecture 5

**Definition.** Given trees  $T = (V, E)$  and  $T' = (V, E')$  we say  $T = T'$  if  $E = E'$ .

**Example.**

**Theorem** (Cayley's Formula). The number of  $n$  vertex labelled trees is  $n^{n-2}$ .

**Definition.** A Prüfer sequence of length  $n - 2$  for  $n \geq 2$  is any sequence of integers between 1 and  $n$  with repetitions allowed. There are  $n^{n-2}$  of these.

**Algorithm.**

**Input** An  $n$ -vertex tree.

**Output** A Prüfer sequence of length  $n - 2$ .

Initialise  $T$  to be a given tree

For  $i = 1$  to  $n - 2$

    Let  $v$  be the vertex of degree 1 with smallest label.

    Let  $s_i$  be the only neighbour of  $v$ .

    Define  $T = T \setminus v$

Return the sequence  $\langle s_1, \dots, s_{n-2} \rangle$

**Example.**

**Proposition.** Let  $d_k$  be the number of occurrences of the number  $k$  in a Prüfer encoding sequence for a labelled tree  $T$ . Then the degree of  $k$  in  $T$  is  $d_k + 1$ .

*Proof.* By induction:

The assertion is true for any tree on 3 vertices since the Prüfer sequence consists of a single label of the vertex of degree 2.

Assume the assertion is true for all labelled trees on  $n$  vertices for some  $n \geq 3$ . Let  $T$  be a labelled tree on  $n + 1$  vertices. Let  $v$  be a leaf with smallest label  $l(v)$ . Let  $w$  be a neighbour of  $v$ . Then the Prüfer sequence  $S$  for  $T$  consists of  $l(w)$  followed by  $S^*$ , the Prüfer sequence of  $T^* = T \setminus v$ .

Now by the inductive hypothesis, for each  $u \in T^*$  the number of occurrences of  $l(u)$  in  $S^*$  is  $d_{T^*}(u) - 1$ . But for all  $u \neq w$  the number of occurrences of the label  $l(u)$  in  $S^*$  is the same as in  $S$ , and  $d_T(u) = d_{T^*}(u)$ . Moreover,  $d_T(w) = d_{T^*}(w) + 1$  and  $l(w)$  has one more occurrence in  $S$  than in  $S^*$ .  $\square$

**Algorithm** (for Prüfer decoding).

**Input** A Prüfer sequence of length  $n - 2$ .

**Output** An  $n$ -vertex labelled tree.