B8.5 Graph Theory Michaelmas Term 2022, 16 lectures Paul Balister Last updated: September 1, 2022

These notes are to accompany the lectures in MT 2022 on graph theory for Part B Mathematics, and are adapted from notes by Oliver Riordan, which in turn were adapted from notes by Alex Scott and Colin McDiarmid. They also owe much to the book *Modern Graph Theory*, Springer-Verlag, 1998 by Béla Bollobás, and are *not for distribution*. The notes may possibly be updated as the course proceeds, although no major changes are planned. If you spot any errors, email: Paul.Balister@maths.ox.ac.uk.

Relationship to Part A Graph Theory

Part A Graph Theory is recommended but not required as a prerequisite. The course as lectured should be self-contained, though a few key results covered in Part A will be stated as exercises to complete yourself if you did not do Part A.

1 Introduction

We need some preliminary definitions and notation (see the separate handout for a summary). We write [n] for the set $\{1, 2, ..., n\}$. For any set S, we write $\binom{S}{k}$ for the set of subsets of S of size k, that is, $\binom{S}{k} = \{A \subseteq S : |A| = k\}$ (and has cardinality $\binom{|S|}{k}$). Some authors write $S^{(k)}$ instead of $\binom{S}{k}$.

A (simple) graph G is an ordered pair (V, E), where V is a set and $E \subseteq \binom{V}{2}$. In this course V will almost always be finite and non-empty – this is assumed unless stated otherwise. The elements of V are called the *vertices* of G and the elements of E the *edges* of G.

We often write uv for an edge $\{u, v\}$ (so uv means the same as vu). We say that u and v are *adjacent* in G if uv is an edge of G. A vertex v and an edge e are *incident* if v is one of the *endvertices* of e, i.e., one of the two vertices in e. Two edges *meet* if they intersect, i.e., share a vertex. Graphically, we represent vertices as points (or more often blobs) and edges as lines or curves joining pairs of points (blobs); how a graph is drawn is irrelevant as far as the structure of the graph itself is concerned. The reason for using blobs is that it makes clear in the drawing where the vertices are: we may have to draw the lines/curves for two edges so that they cross even though the edges do not share a vertex.



Figure 1: Two ways of drawing the same graph $G = (\{a, b, c, d\}, \{ab, ac, ad, bd, cd\})$.



Figure 2: Some special graphs.

If G = (V, E), we write V(G) for V and E(G) for E. The *order* of a graph G, denoted by |G| or v(G), is the number of vertices, so |G| = v(G) = |V(G)|. The *size* of G is the number of edges, e(G) = |E(G)|; however, sometimes 'size' is used to mean 'order', so it is safest to avoid this term.

Graphs G and H are *isomorphic* if there exists a bijection $\varphi \colon V(G) \to V(H)$ such that, for each $x, y \in V(G)$, we have $xy \in E(G)$ iff $\varphi(x)\varphi(y) \in E(H)$. We say that φ is an *isomorphism*,¹ and write $G \cong H$. It is easy to check that isomorphism of graphs is an equivalence relation, and simply amounts to a 'relabeling' of the vertices. Often we do not make a distinction between isomorphic graphs, treating them as the same.²

make a distinction between isomorphic graphs, treating them as the same.² A graph G is complete if $E(G) = \binom{V(G)}{2}$. We write $K_n = ([n], \binom{[n]}{2})$ for the complete graph on the vertex set [n]. Clearly any complete graph of order n is isomorphic to K_n . A graph G is empty if $E(G) = \emptyset$. We write $E_n = ([n], \emptyset)$ for the empty graph on the vertex set [n]. Clearly any empty graph of order n is isomorphic to E_n . A graph G is a cycle on n vertices, or a cycle of length n, $(n \ge 3)$ if it is isomorphic to $C_n :=$ $([n], \{12, 23, \ldots, (n-1)n, n1\})$. A graph G is a path on n vertices, or a path of length n-1 $(n \ge 1)$ if it is isomorphic to $P_n := ([n], \{12, 23, \ldots, (n-1)n\})$.

Warning. Some authors define P_n as a path with n edges, while in this course P_n has n vertices. Similarly, the length of a path is usually the number of edges, but some use it to mean the number of vertices. Always check which definitions are being used! Sometimes the term *edge length* is used to emphasise that it is the number of edges being counted.

A graph H is a subgraph of G if $V(H) \subseteq V(G)$ and $E(H) \subseteq E(G)$. A spanning subgraph is one that includes all the vertices: V(H) = V(G). An induced subgraph is one that includes all possible edges, i.e., all edges with both endvertices in V(H): $E(H) = E(G) \cap \binom{V(H)}{2}$. Indeed, given any $W \subseteq V(G)$ there is a unique induced subgraph with vertex set W which we write as G[W], which we call the subgraph of G induced by W.

We often say that H is a subgraph of G (or more precisely that G contains a copy of H) to mean that G has a subgraph isomorphic to H.³

The complement of a graph G = (V, E) is $\overline{G} = (V, \binom{V}{2} \setminus E)$. Thus $\overline{K}_n = E_n$ and $\overline{E}_n = K_n$. For an edge e, we write G - e for the subgraph $(V, E \setminus \{e\})$, obtained by deleting the edge e from G. For $e \in E(\overline{G})$, $G + e = (V, E \cup \{e\})$ is the graph obtained by adding the edge e to G. For a vertex v, we write G - v for the subgraph induced by $V \setminus \{v\}$, i.e., the subgraph obtained from G by deleting v and (as we must) all edges incident with v.

¹More generally we can define a graph homomorphism as a map $\varphi \colon V(G) \to V(H)$ such that $xy \in E(G)$ implies $\varphi(x)\varphi(y) \in V(H)$. An isomorphism is then just an invertible map φ such that both φ and φ^{-1} are homomorphisms.

²Sometimes we talk of 'labelled graphs', meaning the actual values of vertices are important, and 'unlabelled graphs' for graphs considered only up to isomorphism.

³Equivalently, there is an injective homomorphism from H to G.

⁴Some people write $G \setminus e$ for G - e. Not to be confused with G/e, defined later.

In much of the following, unless otherwise indicated, the implicitly assumed setting is an arbitrary graph G = (V, E).

Degrees

The *degree* of a vertex v is the number of incident edges,

$$d(v) = \big| \{ w \in V : vw \in E \} \big|.$$

We write $d_G(v)$ if we want to specify the graph. A vertex w is a *neighbour* of v if v and w are adjacent, i.e., $vw \in E$. The set $N(v) = N_G(v) = \{w \in V : vw \in E\}$ is the *neighbourhood* of v, so d(v) = |N(v)|. Some authors write $\Gamma(v)$ instead of N(v).

A graph G is r-regular if every vertex has degree r. If d(v) = 0, v is an isolated vertex. If $V = \{v_1, \ldots, v_n\}$, the degree sequence of G is the sequence $d(v_1), d(v_2), \ldots, d(v_n)$, often arranged in nondecreasing order.

Lemma 1.1 (Handshaking Lemma). For any graph G = (V, E),

$$\sum_{v \in V} d(v) = 2e(G).$$

Proof. Consider the number of pairs (v, e) where v is a vertex of G and e is an edge of G incident with v. We count them in two different ways. Firstly, each vertex v is in exactly d(v) such pairs, so there are $\sum_{v \in V} d(v)$ pairs in total. Secondly, each edge e of G is in exactly two such pairs, so there are 2|E| = 2e(G) pairs.

Corollary 1.2. For any graph G, the number of vertices with odd degree is even. \Box

Paths, cycles and walks in graphs

A path on t vertices in G is a subgraph of G isomorphic to P_t ; a cycle of length t in G is a subgraph isomorphic to C_t . We usually just list the vertices to describe a path or cycle. Thus $v_1v_2\cdots v_t$ is a path of length t in G if and only if v_1,\ldots,v_t are distinct vertices of G and $v_1v_2,\ldots,v_{t-1}v_t$ are edges of G^{5} . Similarly, $v_1v_2\cdots v_tv_1$ is a cycle in G if and only if $t \ge 3, v_1,\ldots,v_t$ are distinct vertices of G, and $v_1v_2,\ldots,v_{t-1}v_t,v_tv_1$ are edges of G. A graph is acyclic if it contains no cycles.

We say $v_0v_1 \cdots v_t$ is a *walk* in *G* if v_0, v_1, \ldots, v_t are (not necessarily distinct) vertices of *G* such that $v_iv_{i+1} \in E$ for each $i = 0, 1, \ldots, t - 1$. The *length* of a walk is the number of steps, here *t*. If $x = v_0$ and $y = v_t$ then we speak of a walk from *x* to *y*, or an *x*-*y* walk; an *x*-*y* path is defined similarly. A walk $v_0v_1\cdots v_t$ is *closed* if $v_t = v_0$.

Exercise. Let G be a graph and $x, y \in V(G)$. Then G contains an x-y walk if and only if G contains an x-y path.

⁵The two definitions of path in G are not quite the same: for existence, they are equivalent, but for counting paths, they differ by a factor of 2 for $t \ge 2$ as the subgraph corresponding to $v_1v_2...v_t$ is the same as for $v_tv_{t-1}...v_1$. A similar comment applies to cycles with a different factor.

In other words, if we want to get from x to y, then allowing ourselves to revisit vertices does not help. This simple observation is useful, allowing us to switch back and forth between using paths and walks to define connectedness, at any point using whichever definition is easiest to work with. The cleanest proof of the exercise is to consider a shortest x-y walk and show that it is a path.

A graph G is connected if for all $x, y \in V$ there is at least one x-y path in G. The components of a general graph G are the maximal connected subgraphs. It is easy to check that G is the disjoint union of its components. Indeed, consider the relation \sim on V(G) defined by " $x \sim y$ iff there exists an x-y walk". (Equivalently, there exists an x-y path.) It is easy to check that this is an equivalence relation, and that the components are the subgraphs induced by the equivalence classes.

We finish this section with a simple lemma giving a condition under which we are guaranteed that G contains a cycle.

Lemma 1.3. Let G be a finite graph in which every vertex has degree at least 2. Then G contains a cycle.

Proof. Pick $v_0 \in V(G)$ and $v_1 \in N(v_0)$. Now for each $i \ge 1$ we can successively pick $v_{i+1} \in N(v_i)$ such that $v_{i+1} \ne v_{i-1}$ (since $|N(v_i)| \ge 2$). Thus we have a sequence v_0, v_1, v_2, \ldots such that $v_{i-1}v_i \in E(G)$ for every i, and v_{i-1}, v_i and v_{i+1} are always distinct. Since G has only finitely many vertices, some vertex must appear more than once. Pick i < j such that $v_i = v_j$ and j is minimal. Then $j - i \ge 3$, and by minimality v_i, \ldots, v_{j-1} are distinct. Thus $v_i \cdots v_{j-1}v_i$ is a cycle in G.

2 Trees

A *tree* is simply an acyclic connected graph. A general acyclic graph, or equivalently, a graph in which each component is a tree, is called a *forest*.

Lemma 2.1. The following are equivalent, where minimality/maximality is with respect to deleting/adding edges.

- (a) T is a tree,
- (b) T is a minimal connected graph,
- (c) T is a maximal acyclic graph.

The precise meaning of (b) is that T = (V, E) is connected, but that for any strict subset E' of E, (V, E') is not connected. Equivalently, T is connected, but for any edge eof T, T - e is not connected. Similarly in (c), T = (V, E) is acyclic, but (V, E') contains a cycle for any strict superset $E' \supset E$. Equivalently, T is acyclic, but for any edge e of \overline{T} , T + e is not acyclic.

Proof. Revision from Part A or exercise, as applicable.

A spanning tree of a graph G is a spanning subgraph of G that is a tree, i.e., a subgraph of G that is a tree containing all the vertices of G.

Corollary 2.2. Every connected graph G has at least one spanning tree.

Proof. Remove edges one-by-one, keeping the graph connected, until we can remove no more. The graph T that remains is a minimal connected graph with vertex set V(G); by Lemma 2.1, T is a tree.

A vertex v of any graph G is called a *leaf* if d(v) = 1. This term is most often used in the context of trees/forests.

Lemma 2.3. Every tree on $n \ge 2$ vertices has at least one leaf.

Proof. T is connected, so it has no isolated vertices (vertices of degree 0). But T has no cycle, so by Lemma 1.3 it must have a vertex of degree less than 2. Therefore it has a vertex v with degree 1.

In fact, every tree with at least 2 vertices has at least two leaves; there are many proofs of this fact. One involves modifying the argument above slightly. Another way is to consider a longest path $v_1 \ldots v_t$ in the tree. Clearly $t \ge 2$ as the tree is connected and has at least two vertices (and so must have an x-y path for some $x \ne y$). If v_1 we adjacent to any v_i , i > 2, we would obtain a cycle $v_1 \ldots v_i v_1$, a contradiction, while if v_1 were adjacent to any $v \notin \{v_1, \ldots, v_t\}$ then $vv_1 \ldots v_t$ would be a longer path. Thus the only neighbour of v_1 is v_2 and so v_1 is a leaf. However, exactly the same argument also applies to v_t , so we have at least two leaves.

The significance of leaves is shown by the following simple result.

Lemma 2.4. Let v be a leaf of a graph G. Then G is a tree iff G - v is a tree.

Proof. Revision/Exercise.

Lemma 2.5. If T is a tree on n vertices, then e(T) = n - 1.

Proof. We use induction on n; the case n = 1 is trivial. Let T be any tree with $n \ge 2$ vertices. By Lemma 2.3, T has a leaf v. By Lemma 2.4, T' = T - v is a tree. Since T' has n - 1 vertices, by induction it has n - 2 edges. Thus T has n - 1 edges. \Box

Combining Lemmas 2.1 and 2.5 gives some further characterisations of trees.

Corollary 2.6. Let G be a graph with n vertices. TFAE (the following are equivalent): (a) G is a tree,

- (b) G is connected and e(G) = n 1,
- (c) G is acyclic and e(G) = n 1.

Proof. (a) implies (b) and (c) by the definition of a tree and Lemma 2.5. Suppose that (b) holds. Then G has a spanning tree T which, by Lemma 2.5, has n - 1 = e(G) edges. A spanning subgraph includes all the vertices by definition, and since e(T) = e(G), in this case it includes all the edges too. Thus T = G and so G is a tree, completing the proof that (b) implies (a). Now suppose that (c) holds. Each component C_i of G is a tree. Now $e(C_i) = |C_i| - 1$, so $e(T) = \sum e(C_i) = \sum (|C_i| - 1) = n - c$, where c is the number of components. Thus c = 1 and G is connected.

Counting trees

Let's start with a simpler question: how many graphs G = (V, E) are there with vertex set [n]? Each of the $\binom{n}{2}$ possible edges may or may not be included in E, with all possibilities allowed, so the answer is $2^{\binom{n}{2}}$. Note that we are not asking how many *isomorphism* classes there are: this is a much harder question. (Sometimes, counting graphs on a given vertex set is referred to as 'counting labelled graphs'; counting isomorphism classes is referred to as 'counting unlabelled graphs'.)

Counting trees is much harder than counting all graphs. The answer was found (but not really proved) by Cayley in 1889, though implicitly earlier by Borchardt in 1860; it is now known as *Cayley's formula*.

Theorem 2.7. For any $n \ge 1$ there are exactly n^{n-2} trees T with vertex set [n].

Proof. The result is trivial for n = 1 and 2, so fix $n \ge 3$. We shall map each tree on [n] to its *Prüfer code* $\mathbf{c} = (c_1, c_2, \ldots, c_{n-2})$, where $1 \le c_i \le n$. (The c_i need not be distinct.) Since there are n^{n-2} possible codes, it suffices to show that the map gives a bijection between trees on [n] and codes.

Given a tree T on [n] we construct its code as follows:

 $T_1 := T$ has at least one leaf. Find the leaf v_1 with the smallest number, remove it, and write down the number c_1 of the (unique) vertex v_1 was adjacent to. Repeat until exactly two vertices remain. Thus, for example, v_2 is the smallest leaf of $T_2 := T - v_1$, and c_2 is the vertex of T_2 that v_2 is adjacent to. In general v_i is the smallest leaf of $T_i := T - v_1 - \cdots - v_{i-1}$ and c_i is the vertex of T_i it is adjacent to. Note that c_1, \ldots, c_{n-2} form the code, not v_1, \ldots, v_{n-2} .



Figure 3: Example of a Prüfer code.

The following observation is key to the proof: a vertex w with degree d in T appears exactly d-1 times in the code \mathbf{c} . Indeed, we write w down in the code each time we delete a neighbour of w, i.e., each time its degree decreases. The final degree of w is always 1: either w is deleted when it is a leaf, or w is left at the end as one of the two final vertices, which are then leaves. More generally, if the degree of w in $T - v_1 - v_2 - \cdots - v_{i-1}$ is d, then w occurs d-1 times in c_i, \ldots, c_{n-2} . It follows from this that

$$v_{1} = \min\{[n] \setminus \{c_{1}, \dots, c_{n-2}\}\}$$

$$v_{2} = \min\{[n] \setminus \{v_{1}, c_{2} \dots, c_{n-2}\}\}$$

$$\dots$$

$$v_{i} = \min\{[n] \setminus \{v_{1}, \dots, v_{i-1}, c_{i}, \dots, c_{n-2}\}\} \quad i \leq n-2.$$
(1)

Let us write v_{n-1} and v_n (with WLOG $v_{n-1} < v_n$) for the two vertices left at the end when we constructed the code, so

$$\{v_{n-1}, v_n\} = [n] \setminus \{v_1, \dots, v_{n-2}\}.$$
(2)

Then, since we deleted the edge $v_i c_i$ at step *i*, and were left with the edge $v_{n-1}v_n$ between the final two vertices,

$$E(T) = \{v_1c_1, \dots, v_{n-2}c_{n-2}, v_{n-1}v_n\}.$$
(3)

The formulae above describe T, the tree that we started with, in terms of its code $\mathbf{c} = (c_1, \ldots, c_{n-2})$. Does this mean that the proof is complete? No! We started by assuming that T was a tree, with code \mathbf{c} , and then showed that given \mathbf{c} , we could identify T (i.e., the map from trees to codes is injective). So for any code *coming from a tree*, there is a unique tree with that code. We still need to show that for every code \mathbf{c} , there is a tree with code \mathbf{c} (i.e., the map from trees to codes is surjective).

The formulae above tell us where to look: if there is a tree with code \mathbf{c} , it *must* be as described above. So let us check.

Formally, let **c** be any possible code (c_1, \ldots, c_{n-2}) . Then we may use (1) to define v_1, \ldots, v_{n-2} . (Each time we take the minimum of a non-empty set, which makes sense.) Also, from (1) we see that v_i is not equal to any of v_1, \ldots, v_{i-1} . Thus v_1, \ldots, v_{n-2} are distinct.

Next, we define $v_{n-1} < v_n$ to be the two remaining elements of [n], as in (2), so v_1, \ldots, v_n are distinct; they are $1, 2, \ldots, n$ in some order.

Finally, we let T be the graph with vertex set [n] and edge set given by (3). We need to check that T is indeed a tree, and that it has code c. We do this step-by-step: first note that from our definition (1) of v_i , it is distinct from c_j , $j \ge i$. Thus c_j is distinct from v_i , $i \le j$, so for each j, $c_j \in \{v_{j+1}, \ldots, v_n\}$. Let T_i be the graph with

$$V(T_i) = \{v_i, \dots, v_n\}$$
 and $E(T_i) = \{v_i c_i, \dots, v_{n-2} c_{n-2}, v_{n-1} v_n\}.$

(This makes sense since the ends of the edges are distinct and lie in $V(T_i)$.) Then T_{n-1} is a tree with two vertices. Also, T_i is constructed from T_{i+1} by adding a new vertex v_i and one edge $v_i c_i$. So, by Lemma 2.4, T_i is a tree for $i = n - 2, n - 3, \ldots, 2, 1$. In particular, $T = T_1$ is a tree. That the code of T is **c** is an exercise; see Problem Sheet 1.

3 Long circuits, paths and cycles

An Euler circuit (Euler tour)⁶ in a graph G is a closed walk that contains every edge of G exactly once. (If |G| = 1 we say that G has a trivial Euler circuit.)

 $^{^{6}}$ A *trail* in a graph is a walk which does not repeat edges (although it may repeat vertices). A *circuit* in a graph is a closed trail, i.e., a closed walk that is a trail. Hence an Euler circuit is just a circuit that uses every edge of the graph. Warning: some authors use 'circuit' to mean 'cycle' and vice versa.

Theorem 3.1. Let G be a connected graph. Then G has an Euler circuit if and only if the degree of every vertex is even.

Proof. For the (easier) 'only if' direction, pick $v \in V(G)$. If an Euler circuit enters v k times then it leaves v k times, and so it uses 2k edges incident with v. Thus d(v) must be even.

For the converse, we proceed by induction on e(G), with the result being trivial for e(G) = 0. For the induction step, take any G with e(G) > 0 and assume the result holds for all graphs with fewer edges than G. Since G is connected, each vertex has degree at least 1. As all degrees are even, all vertices have degree at least 2. By Lemma 1.3, G contains a cycle C. The graph H obtained from G by removing the edges of C still satisfies the condition that all of its vertices have even degree. It is possibly no longer connected, but all of its components H_i are, and have fewer edges than G. Therefore, by the induction hypothesis, we can find an Euler circuit E_i in each component H_i . Moreover, each of the components H_i must have at least one vertex in common with the cycle C, otherwise G would have been disconnected in the first place. (Consider a path in G from any vertex in H_i to any vertex in C. The first vertex of this path that lies in C also lies in H_i .) Thus each E_i must be of the form $x_i x_{i,1} x_{i,2} \ldots x_{i,n_i} x_i$ with x_i some vertex of C. We can thus 'merge' each of the E_i into C by, e.g., replacing the first occurrence of x_i in C by the walk $x_i x_{i,1} x_{i,2} \ldots x_{i,n_i} x_i$, to obtain an Euler circuit of the original graph G.



 $C = ebcdfce, E_i = badb \implies ebadbcdfce$

Figure 4: Inserting circuit E_i into circuit C.

A Hamilton cycle in a graph G is a cycle in G that contains every vertex; a graph is called Hamiltonian if it has a Hamilton cycle.

Superficially, the following two problems may seem similar: in a given graph G, is there a closed walk using every edge exactly once (Euler circuit), and is there a closed walk using every vertex exactly once (Hamilton cycle)? But it's easy to tell (using Theorem 3.1) what the answer to the first question is. The second is much harder; for those interested in complexity theory, it is an NP-complete problem.

The minimum degree of a graph G is $\delta(G) = \min_{v \in V} d(v)$, the maximum degree is $\Delta(G) = \max_{v \in V} d(v)$, and the average degree is

$$\bar{d}(G) = \frac{1}{|G|} \sum_{v \in V} d(v) = \frac{2 e(G)}{|G|}$$

It is not hard to see that any graph with $\delta(G) \ge d$ contains a path of edge length at least d: start at any v_0 and, given $v_0 \cdots v_i$ with i < d, choose v_{i+1} to be a neighbour of v_i

not among v_0, \ldots, v_{i-1} . In fact, for connected graphs with many more than d vertices, we can find a path of roughly twice this length.

Lemma 3.2. If G is a connected graph which is not Hamiltonian, then the edge length of a longest path in G is at least the length of a longest cycle.

Proof. Let C be a longest cycle in G, with length ℓ . We have $\ell < n$ since G is not Hamiltonian, so there are vertices not on C. Since G is connected, there is at least one edge vw with $v \in V(C)$ and $w \notin V(C)$ (pick $x \in V(C)$, $y \notin V(C)$, and take the first edge of an x-y walk that leaves V(C)). But then the edge vw and C between them contain a path of length ℓ .



Theorem 3.3. Let G be a connected graph with $n \ge 3$ vertices in which every pair v, w of non-adjacent vertices satisfies $d(v) + d(w) \ge k$. If k < n then G contains a path of edge length k; if $k \ge n$ then G is Hamiltonian.

Proof. If G has a Hamilton cycle, then it also has a path of edge length n-1 (the maximum possible k < n) and we are done. So suppose not. Let $P = v_0 v_1 \cdots v_\ell$ be a longest path in G. Since G is connected and has at least 3 vertices, $\ell \ge 2$. By Lemma 3.2, G contains no cycle of length $\ell+1$. In particular, $v_0 v_\ell \notin E(G)$, so $d(v_0) + d(v_\ell) \ge k$. If, for some $1 \le i \le \ell$, both $v_0 v_i$ and $v_{i-1}v_\ell$ were edges, then we would have a cycle $v_0 v_1 \cdots v_{i-1}v_\ell v_{\ell-1} \cdots v_i$ of length $\ell+1$.



Hence $A = \{i \in [\ell] : v_0 v_i \in E(G)\}$ and $B = \{i \in [\ell] : v_{i-1}v_{\ell} \in E(G)\}$ are disjoint subsets of $[\ell]$. Thus, noting that all neighbours of v_0 and v_{ℓ} are on P (otherwise we would have a longer path), we have

$$\ell \ge |A| + |B| = d(v_0) + d(v_\ell) \ge k.$$

This is impossible if k = n as $\ell \leq n - 1$. If k < n it shows that G contains a path of length k, as required.

Corollary 3.4. If G is connected, |G| = n, and $\delta(G) \ge d$, then G contains a path of edge length (at least) min $\{2d, n-1\}$.

Proof. Trivial for n = 1, 2. For $n \ge 3$, apply Theorem 3.3 with k = 2d.

As another corollary we obtain the following result.

Theorem 3.5 (Dirac's Theorem). Let G be a graph with $n \ge 3$ vertices. If $\delta(G) \ge \frac{n}{2}$, then G contains a Hamilton cycle.

Proof. If $\delta(G) \ge n/2$ then any two non-adjacent vertices have at least one common neighbour, so G is connected. (Or: if G is not connected then there is a component C with at most n/2 vertices. Then any $v \in V(C)$ has degree at most |C| - 1 < n/2.) Now Theorem 3.3 applies with k = n.

This result is best possible, in that we cannot replace the lower bound by $\lceil \frac{n}{2} \rceil - 1$. (For n even, consider the disjoint union of two complete graphs $K_{n/2}$. For $n \ge 5$ odd, consider two $K_{(n+1)/2}$ s sharing a single vertex.)

Theorem 3.3 of course implies a slightly stronger result than Dirac's Theorem, known as Ore's Theorem.

Theorem 3.6 (Ore's Theorem). If G has order $n \ge 3$, and if $d(x) + d(y) \ge n$ whenever $xy \in E(\overline{G})$, then G has a Hamilton cycle.

Theorem 3.3 also lets us relate the length of the longest path in G to the *average* degree of G.

Theorem 3.7. If G is a graph with n vertices containing no path of edge length $k, k \ge 2$, then $e(G) \le \frac{k-1}{2}n$.

Proof. Induction on n. For $n \leq k$ we have $e(G) \leq e(K_n) = \frac{n-1}{2}n \leq \frac{k-1}{2}n$, so we are done. Now suppose n > k. We may assume G is connected; otherwise apply the induction hypothesis to the components (which do not contain P_{k+1}). Then G is connected with no P_{k+1} and $n > k \geq 2$. So by Theorem 3.3 there are (non-adjacent) vertices v, w of G such that $d(v) + d(w) \leq k - 1$. WLOG $d(v) \leq d(w)$, so $d(v) \leq \frac{k-1}{2}$. Since G - v has n - 1 vertices and contains no P_{k+1} , applying the induction hypothesis to G - v we have

$$e(G) = d(v) + e(G - v) \leq \frac{k-1}{2} + \frac{k-1}{2}(n-1) = \frac{k-1}{2}n,$$

oof.

completing the proof.

The result above can be rephrased to say that G always contains a path of edge-length at least $\overline{d}(G)$. We do not get the extra factor of 2 we had in Corollary 3.4, but we assuming something only about average degree, not about the degree of *every* vertex.

4 Vertex colourings

A proper vertex colouring (or simply a colouring) of a graph G is an assignment of a colour to each vertex such that adjacent vertices receive different colours. The least number of colours in such a colouring is the chromatic number $\chi(G)$. For example $\chi(K_n) = n$, $\chi(E_n) = 1$, $\chi(C_4) = 2$ and $\chi(C_5) = 3$. In fact, any even cycle (cycle of even length) has chromatic number 2, and any odd cycle has chromatic number 3.

Often we use positive integers as the colours: a (proper) k-colouring of G is a function $f: V(G) \to \{1, \ldots, k\}$ so that $f(u) \neq f(v)$ whenever $uv \in E(G)$. G is k-colourable if it has a k-colouring, so $\chi(G)$ is the least k for which G is k-colourable.



Figure 5: Chromatic numbers of some small graphs.

As an example of an application, suppose we have to schedule exams, where each exam takes one period. Construct a graph G with a vertex for each exam and an edge uvwhenever one or more students need to take both exams u and v. Then a feasible exam schedule corresponds to a colouring of G, and the least number of periods possible is $\chi(G)$.

Certainly, if H is a subgraph of G, then $\chi(H) \leq \chi(G)$. Clearly, a disconnected graph is k-colourable if and only if its components are, so the chromatic number of G is the maximum of the chromatic numbers of its components. In fact, we can extend this to graphs overlapping in certain ways.

The union of two graphs G_1 and G_2 is the graph with vertex set $V(G_1) \cup V(G_2)$ and edge set $E(G_1) \cup E(G_2)$.

Lemma 4.1. Let G_1 and G_2 be graphs with $V(G_1) \cap V(G_2) = W$ such that $G_1[W]$ and $G_2[W]$ are complete. Then $\chi(G_1 \cup G_2) = \max\{\chi(G_1), \chi(G_2)\}$.

Proof. As each G_i is a subgraph of $G_1 \cup G_2$, $\chi(G_1 \cup G_2) \ge \max{\chi(G_1), \chi(G_2)}$ is trivial. Now let $k = \max{\chi(G_1), \chi(G_2)}$ so that both G_1 and G_2 are k-colourable; we must show that $G_1 \cup G_2$ is also k-colourable. Let c_i be a k-colouring of G_i , and let $W = {w_1, \ldots, w_r}$. Since c_1 assigns distinct colours to w_1, \ldots, w_r , we may permute the colours (i.e., keep fixed which sets of vertices get the same colour, but assign different colours to these sets) to obtain a new k-colouring \tilde{c}_1 of G_1 in which w_1, \ldots, w_r get colours $1, 2, \ldots, r$ in this order. Do the same for G_2 , and then combine the colourings \tilde{c}_1 and \tilde{c}_2 , which agree on W, to obtain a colouring of $G_1 \cup G_2$. As the colouring is proper on both G_1 and G_2 and there are no edges from $G_1 - W$ to $G_2 - W$, this is a proper k-colouring of $G_1 \cup G_2$.

A cutvertex v in a connected graph G is a vertex such that G - v is disconnected. (In a general graph, it's a vertex whose deletion increases the number of components of the graph.) Lemma 4.1 may be applied in particular to any graph G with a cutvertex.

Now we consider what happens when G is k-colourable for small k.

A graph G has $\chi(G) = 1$ if and only if G has no edges, so is an empty graph.

A graph G = (V, E) is *bipartite* if V can be split into disjoint sets X and Y such that $E \subseteq \{xy : x \in X, y \in Y\}$. (We allow one of X or Y to be empty, so K_1 is bipartite.) The complete bipartite graph $K_{m,n}$ has $V = \{a_1, \ldots, a_m, b_1, \ldots, b_n\}$ and $E = \{a_i b_j : i = 1, \ldots, m, j = 1, \ldots, n\}$. The connection to colouring is that $\chi(G) \leq 2$ if and only if G is bipartite: consider $X = \{v : c(v) = 1\}$ and $Y = \{v : c(v) = 2\}$.

Deciding whether a (connected) graph is 2-colourable (i.e., bipartite) is very easy: start somewhere with one colour (it doesn't matter which) and work outwards from there – having coloured a vertex, the colours of its neighbours are forced, and we either get stuck or we don't. The next simple lemma gives a criterion.

Lemma 4.2. A graph G is 2-colourable (bipartite) if and only if it contains no odd cycles.

Proof. If G is 2-colourable then, in any 2-colouring, the colours around any cycle C alternate, implying that C has even length.

For the reverse implication we use induction on |G|; the base case |G| = 1 is trivial. For the induction step let G be a graph with $n \ge 2$ vertices with no odd cycle. We may assume that G is connected (else colour its components). It follows that there is (at least) one vertex v such that G - v is connected (take v to be a leaf of a spanning tree of G). By induction we may 2-colour G - v. If all neighbours of v have the same colour in this colouring, then we may extend the colouring to G by using the opposite colour for v. So we may suppose that v has neighbours x and y with different colours. Then G - v contains a path P from x to y; along this path the colours alternate, so P has odd length and together with vx and vy forms an odd cycle in G, contradicting our assumption.

In general, finding the chromatic number of a graph is very hard; even the question 'is $\chi(G) \leq 3$ ' is hard. However, we can give some general bounds on $\chi(G)$.

A copy of K_k in a graph G is called a *complete subgraph* or a *clique*. The *clique number* $\omega(G)$ of G is the largest k such that G contains a copy of K_k . A set S of vertices is an *independent set* in G (or *stable set*) if G[S] has no edges, i.e., no two vertices of S are adjacent in G. Thus, a (proper) colouring of G corresponds to a partition of V(G) into independent sets. The *independence number* $\alpha(G)$ is the maximum size of an independent set in G. For example, $\omega(C_5) = \alpha(C_5) = 2$. Note that $\alpha(G) = \omega(\overline{G})$.

Lemma 4.3. $\chi(G) \ge \max\{\omega(G), \frac{|G|}{\alpha(G)}\}.$

Proof. All vertices in a clique must get different colours in any colouring, so $\chi(G) \ge \omega(G)$. Also, since the vertices of each colour form an independent set, each colour is used on at most $\alpha(G)$ vertices, so we need at least $\frac{|G|}{\alpha(G)}$ colours.

Given an ordering v_1, \ldots, v_n of the vertices of a graph G, the greedy algorithm constructs a (proper) colouring of G with positive integers by colouring the vertices in order: each vertex receives the least colour not already assigned to one of its neighbours.

Lemma 4.4. $\chi(G) \leq \Delta(G) + 1$.

Proof. Take any ordering of the vertices and apply the greedy algorithm: each vertex has at most $\Delta(G)$ forbidden colours, and so will get a colour from $\{1, 2, \ldots, \Delta(G) + 1\}$. \Box

This bound is tight in some cases: in particular if G is complete or an odd cycle. But usually we can do better; we start with two simple lemmas.

Lemma 4.5. Let G be a connected graph with n vertices and let $v \in V(G)$. Then we may order the vertices as $v_1, \ldots, v_{n-1}, v_n = v$ so that each vertex other than v has at least one neighbour coming after it.

Proof. See problem sheet 2.

Lemma 4.6. Let G be a connected graph with $\Delta(G) \leq d$ and $\delta(G) < d$. Then $\chi(G) \leq d$.

Proof. Pick a vertex v with d(v) < d, take an ordering as in the last lemma, and apply the greedy algorithm: each vertex has at most d-1 forbidden colours.



Figure 6: Decomposition of G into G_1, G_2 , overlapping in $\{x, y\}$, and definition of G_1^+, G_2^+ .

(We won't need this lemma in the proof that follows, but it encapsulates one key idea of that proof.)

Dealing with the d-regular case will be significantly harder, though we now have the tools we need.

Theorem 4.7 (Brooks' Theorem). Let G be a connected graph. If G is neither an odd cycle nor a complete graph then $\chi(G) \leq \Delta(G)$.

Proof. For $\Delta(G) \leq 2$ the result is easy as the graph is either a cycle or a path, so suppose $\Delta(G) \geq 3$. It is convenient to restate the result slightly as follows: let $d \geq 3$ and let G be any graph with $\Delta(G) \leq d$ which does not contain a copy of K_{d+1} . Then $\chi(G) \leq d$. Since a connected graph with maximum degree d that contains a copy of K_{d+1} must be K_{d+1} , this restatement (applied with $d = \Delta(G)$) implies the theorem. We prove the restated result by induction on n = |G|.

If G is disconnected we are done by induction, so suppose G is connected. If G has a cutvertex v, then we may write $G = G_1 \cup G_2$ where G_1 and G_2 overlap precisely in v and $|G_1|, |G_2| < n$. By induction $\chi(G_1) \leq d$ and $\chi(G_2) \leq d$, so by Lemma 4.1 $\chi(G) \leq d$. Hence we may assume G has no cutvertex.

Let v be a vertex of G with degree d. (If there is none, then $\chi(G) \leq \Delta(G) + 1 \leq d$ by Lemma 4.4.) Since G contains no K_{d+1} , we can find neighbours x and y of v such that $xy \notin E(G)$. Suppose that G - x - y is connected. Then we may order the vertices of G - x - y as in Lemma 4.5, ending at v. Putting x and y at the beginning of this ordering, we obtain an ordering of the vertices of G in which each vertex apart from v precedes at least one of its neighbours. Moreover, the greedy algorithm gives x and y the same colour, so when it comes to assign a colour to v, at most d - 1 colours are forbidden. Therefore the greedy algorithm uses at most d colours with this ordering.

Suppose instead that G - x - y is not connected. Then $V(G) \setminus \{x, y\}$ can be partitioned into non-empty sets A and B with $e(A, B) = 0^7$. Let $G_1 = G[A \cup \{x, y\}]$ and $G_2 = G[B \cup \{x, y\}]$, so G consists of its subgraphs G_1 and G_2 overlapping in the non-adjacent vertices x and y (see Figure 6). Both x and y must have neighbours in each of A and B(if say x had no neighbours in A then G - y would be disconnected, so G would have a cutvertex). Hence x and y have degree at most d - 1 in G_1 and in G_2 . Let $G_j^+ = G_j + xy$. Then $\Delta(G_j^+) \leq d$. If neither G_1^+ nor G_2^+ contains K_{d+1} then by Lemma 4.1 and induction

$$\chi(G) \leqslant \chi(G + xy) = \chi(G_1^+ \cup G_2^+) = \max\{\chi(G_1^+), \chi(G_2^+)\} \leqslant d.$$

So suppose that one, say G_1^+ , contains a copy of K_{d+1} . Note that this copy must include x and y, since $G_1 \subseteq G$ contains no K_{d+1} . Since G is connected, in fact G_1^+ is isomorphic to

 $^{{}^7}e(A,B)$ is the number of edges ab of G with $a \in A$ and $b \in B$

 K_{d+1} . Since x and y have degree d - (d - 1) = 1 in G_2 , we can d-colour G_2 with x and y having the same colour. Indeed, by induction we can d-colour G[B], and each of x and y has only one colour ruled out, so since $d \ge 3$ we can choose the same colour for both. But now we can extend this colouring to G_1 , and hence to all of G.

The chromatic polynomial

Given a graph G, for k = 1, 2, ..., let $P_G(k)$ be the number of (proper) k-colourings of G, i.e., colourings with [k] as the set of available colours (not all colours have to be used). For example, $P_{K_n}(k) = k(k-1)(k-2)\cdots(k-n+1)$ and, trivially, $P_{E_n}(k) = k^n$.

It turns out that with k fixed we can calculate $P_G(k)$ inductively, using two operations on graphs.

If e = uv is an edge in a graph G, we let G/e denote the graph obtained by contracting e; that is, G/e is obtained from G by deleting the vertices u and v and adding a new vertex adjacent to each vertex in $(N(u) \cup N(v)) \setminus \{u, v\}$. In other words, we 'merge' u and v into a single vertex and each vertex $z \neq u, v$ is joined to this merged vertex if and only if either zu or zv was originally an edge of G. (There is a slightly different notion of contraction for multigraphs.)

Lemma 4.8. For each edge e of G and positive integer k, $P_{G-e}(k) = P_G(k) + P_{G/e}(k)$.

Proof. Suppose that e = uv. Let S be the set of k-colourings of G - e, let $S_1 = \{c \in S : c(u) \neq c(v)\}$ and let $S_2 = \{c \in S : c(u) = c(v)\}$. Clearly $|S| = |S_1| + |S_2|$. Also, $P_{G-e}(k) = |S|$, $P_G(k) = |S_1|$ (since these are exactly the colourings of G), and $P_{G/e}(k) = |S_2|$ (since these correspond to the colourings of G/e, taking the common colour of u and v for the new vertex and vice versa).

Theorem 4.9. For every graph G there is a unique polynomial $p_G(x) \in \mathbb{Z}[x]$, the chromatic polynomial of G, such that

$$p_G(k) = P_G(k) \quad \text{for each } k = 1, 2, \dots$$
(4)

Moreover, for every edge e of G we have $p_G(x) = p_{G-e}(x) - p_{G/e}(x)$.

Proof. Uniqueness is immediate since two polynomials that agree on all positive integers must be the same. For existence we use induction on e(G). For the base case e(G) = 0, $G \cong E_n$ for some n, so $P_G(k) = k^n$ for every k and the polynomial x^n has the required properties.

For the inductive step, pick any edge e of G and note that G - e and G/e have fewer edges than G. So by induction there are polynomials p_{G-e} and $p_{G/e}$ satisfying (4) for the corresponding graphs. Consider $p_G = p_{G-e} - p_{G/e}$; this is a polynomial. By Lemma 4.8, for every positive integer k we have $p_G(k) = p_{G-e}(k) - p_{G/e}(k) = P_{G-e}(k) - P_{G/e}(k) = P_G(k)$, as required. The final statement follows immediately: we have shown that there is a polynomial p_G satisfying (4), and know that it is unique. We have also shown that for any edge e, $p_{G-e} - p_{G/e}$ is such a polynomial, so $p_{G-e} - p_{G/e} = p_G$.

From now on we write $p_G(k)$ for the number of k-colourings of G, since this number is an evaluation of the chromatic polynomial. In general, identities for numbers of k-colourings

valid for all k give polynomial identities. As a simple example, if G has components G_1, \ldots, G_j then $p_G(x) = p_{G_1}(x) \cdots p_{G_j}(x)$; this is valid for each $x \in \mathbb{N}$ (since $p_H(k)$ is the number of k-colourings of H), and both sides are polynomials.

Theorem 4.10. Let G be a graph with n vertices and m edges. Then

$$p_G(x) = \sum_{i=0}^{n-1} (-1)^i a_i x^{n-i} = a_0 x^n - a_1 x^{n-1} + \dots + (-1)^{n-1} a_{n-1} x^{n-1}$$

where $a_0 = 1$, $a_1 = m$ and $a_i \ge 0$ for all i.

Proof. We argue by induction on m. For m = 0 we have $G \cong E_n$, so $p_G(x) = x^n$, and we are done. For m > 0, pick an edge e of G. Then |G - e| = n and e(G - e) = m - 1 so by the induction hypothesis,

$$p_{G-e}(x) = x^n - (m-1)x^{n-1} + \sum_{i=2}^{n-1} (-1)^i a_i x^{n-i},$$

where each $a_i \ge 0$. Also |G/e| = n - 1 and $e(G/e) \le m - 1$, so

$$p_{G/e}(x) = x^{n-1} + \sum_{j=1}^{n-2} (-1)^j b_j x^{n-1-j} = x^{n-1} + \sum_{i=2}^{n-1} (-1)^{i-1} b_{i-1} x^{n-i}$$

where each $b_j \ge 0$. By the last part of Theorem 4.9,

$$p_G(x) = p_{G-e}(x) - p_{G/e}(x) = x^n - mx^{n-1} + \sum_{i=2}^{n-1} (-1)^i (a_i + b_{i-1}) x^{n-i},$$

and $a_i + b_{i-1} \ge 0$ for each *i*.

5 Edge colourings

A function $f: E(G) \to [k]$ is a proper edge-colouring of G if edges that meet (i.e., share an endvertex) always receive distinct colours. The edge-chromatic number $\chi'(G)$ (also called the chromatic index) is the smallest k such that G has such an edge-colouring.

Proposition 5.1. If e(G) > 0, then $\Delta(G) \leq \chi'(G) \leq 2\Delta(G) - 1$.

Proof. Since the edges incident with a given vertex must get different colours we have $\chi'(G) \ge \Delta(G)$. For the upper bound, list the edges in any order and apply the greedy algorithm to colour the edges. When we come to colour an edge uv, the number of colours unavailable is at most $d(u) - 1 + d(v) - 1 \le 2\Delta(G) - 2$.

Amazingly, given the maximum degree Δ of a graph, there are only two possible values for the edge-chromatic number, Δ and $\Delta + 1$. The proof involves a 'colour chasing' argument. (More precisely, the proof combines two simple colour chasing arguments in a clever way.)

г		٦.
		н
		н
		н

Theorem 5.2 (Vizing's Theorem). $\chi'(G) = \Delta(G)$ or $\Delta(G) + 1$.

Proof. We need to prove that $\chi'(G) \leq \Delta(G) + 1$. We argue by induction on m = e(G). The result is trivial if m is 0 (or 1), so let G be a graph with m > 0 edges, let xy_1 be any edge of G, and assume (applying the induction hypothesis to $G - xy_1$) that we have coloured every edge of G except xy_1 with colours $1, \ldots, \Delta(G) + 1$. Our aim is to show that we can recolour so that we can colour the edge xy_1 as well.

For any vertex v, since $d(v) < \Delta(G) + 1$, there is at least one colour missing at v, i.e., not appearing on any edges incident with v. Let t_1 be a colour missing at y_1 . We define a sequence y_1, y_2, \ldots, y_j of distinct neighbours of x and a sequence t_1, t_2, \ldots, t_j of colours as follows.

If colour t_1 is missing at x, colour xy_1 with t_1 and we are done. If not, there is an edge xy_2 with colour t_1 , and some colour $t_2 \ (\neq t_1)$ is missing at y_2 . If t_2 is missing at x, colour xy_2 with t_2 and xy_1 with t_1 , and we are done. Otherwise, there is an edge xy_3 with colour t_2 , and there is a colour t_3 missing at y_3 . If t_3 is missing at x we can recolour as above; otherwise there is an edge xy with colour t_3 . This could be a 'new' edge, but it could instead be xy_2 .



Figure 7: Recolouring a fan when some t_j is missing at x.

In general, suppose that we have distinct neighbours y_1, \ldots, y_j of x and distinct colours t_1, \ldots, t_{j-1} such that t_i is missing at y_i for each $i = 1, \ldots, j-1$, the edge xy_1 is uncoloured, and xy_i has colour t_{i-1} for each $i = 2, \ldots, j$. We call this a fan of size j. Note that there is a fan of size 1, consisting of the uncoloured edge xy_1 . Let t_j be a colour missing at y_j .

If t_j is missing at x, then recolour xy_i with t_i for each $i = 1, \ldots, j$, and we are done. (There are no conflicts at the y_i since t_i was missing at y_i , and no conflicts at x since t_1, \ldots, t_{j-1} were already present on xy_2, \ldots, xy_j and t_j was missing.) Otherwise there is an edge xy with colour t_j . Note that $y \neq y_1$ (since xy_1 is uncoloured), and $y \neq y_j$, since t_j is missing at y_j . If $y \notin \{y_2, \ldots, y_{j-1}\}$ then y is a 'new' vertex, so let $y_{j+1} = y$ – we now have a fan of size j + 1.

The process must terminate (consider a fan of maximal size), and if we are not done then we have distinct neighbours y_1, \ldots, y_j of x and colours t_1, \ldots, t_j such that (a) t_i is missing at y_i for each $i = 1, \ldots, j$, (b) xy_1 is uncoloured and xy_i has colour t_{i-1} for each $i = 2, \ldots, j$, and (c) the colour $t := t_j$ appears on xy_i for some $2 \leq i < j$ (and so $t = t_{i-1}$).



Figure 8: Recolouring a fan when some $t = t_j = t_{i-1}$.

Let s be a colour missing at x. Note for later that $t = t_{i-1} = t_j$ and t_i, \ldots, t_{j-1} appear on distinct edges incident with x, and s is missing at x, so

the colours
$$s, t, t_i, \dots, t_{j-1}$$
 are distinct. (5)

For k = 1, ..., i - 1 we recolour xy_k with t_k , and we remove the colour from xy_i (so now t is missing at y_i). We now have:

- t is missing at y_j and also at y_i (since previously xy_i had colour t),
- xy_i is the only uncoloured edge,
- for $k = i + 1, \ldots, j$ the edge xy_k has colour t_{k-1} ,
- for $k = i, \ldots, j 1$ colour t_k is missing at y_k , and
- colour s is missing at x.

We can of course swap the colours s and t everywhere in the colouring, without causing any conflicts. But this doesn't gain anything. Let H be the spanning subgraph of Gconsisting of all edges coloured s or t. Then we can swap s and t within any component of H without causing conflicts. Since $\Delta(H) \leq 2$, H consists of paths and cycles. At each of x, y_i and y_j , at least one of s and t is missing, so each of these vertices has degree ≤ 1 in H. Hence the components of H containing x, y_i and y_j are paths (possibly of length 0), with each of x, y_i and y_j being an end of one of these paths. One of the two following possibilities must occur.

Case 1: x and y_i are in different components of H. Swap s and t in the component containing x. Now t is missing at x, and is still missing at y_i , so we can colour xy_i with t. **Case 2**: x and y_i are in the same component of H. Since a path has at most two ends, it cannot be that x, y_i and y_j are all in the same component of H. Thus x and y_j are in different components of H. Swap s and t in the component containing x. Now t is missing at x, and is still missing at y_j . We can colour xy_k with t_k for each $k = i, i + 1, \ldots, j - 1$ (since, recalling (5), swapping s and t did not affect which edges have colours t_k , $i \leq k < j$, or which vertices these colours are missing at) and colour xy_j with t, and we are done. \Box

Proper edge colourings of any graph G correspond exactly to proper vertex colourings of the *line graph* L(G). This is (as it must be for the previous sentence to be true) the graph with a vertex for each edge of G in which two vertices are adjacent if and only if the corresponding edges of G meet. So in a sense, edge colouring is a special case of vertex colouring, though this viewpoint is not likely to be helpful in proving results such as Vizing's Theorem.

6 Planar Graphs

The graph K_4 may be drawn in the plane with no edges crossing. What about $K_{3,3}$ (Dudeney's problem), or K_5 ? We first need to define clearly what we mean by such a drawing of a graph.



Figure 9: A plane drawing of K_4 .

A simple curve in the plane is the image of a continuous injection $\phi: [0,1] \to \mathbb{R}^2$. Its endpoints are $\phi(0)$ and $\phi(1)$. A simple closed curve is the image of a continuous map $\phi: [0,1] \to \mathbb{R}^2$ that is injective except that $\phi(0) = \phi(1)$. A curve is polygonal if it is formed from a finite number of straight-line segments, i.e., ϕ is piecewise linear.

A drawing of a graph G = (V, E) in the plane is a representation consisting of distinct points x_v for the vertices $v \in V$, and simple polygonal curves c_{uv} for the edges $uv \in E$, such that c_{uv} has x_u and x_v as its endpoints, and the interiors of the curves (i.e., the curves without their endpoints) are disjoint from each other and from the x_v . In other words, the points and curves meet only 'as they should' according to the incidence relation of the graph.

In fact, the usual definition allows the edges to be drawn as simple curves that need not be polygonal; it is an exercise in analysis (that we will not do) to show that the two definitions coincide: a general drawing can be 'converted' into a polygonal drawing.⁸

A graph together with a drawing in the plane is often called a *plane* graph. We tend to use the notation G for a plane graph without explicitly indicating the drawing. A graph is *planar* if it has a drawing in the plane.

Given a plane graph, if we omit from the plane the points corresponding to the vertices and edges, what remains falls into open connected components, the *faces*, exactly one of which is unbounded. To study plane (and planar) graphs we need surprisingly little topology. The next lemma may seem obvious, but not all drawings of planar graphs are as simple as one might hope. (E.g., we may have one face inside another, meeting at a cutvertex.)

 $^{^{8}}$ In fact drawings of simple graphs can be converted to ones where all the edges are straight line segments – this is Fáry's Theorem.



Figure 10: A plane graph with three faces.

Lemma 6.1. Let e be an edge of a plane graph G. Then e is in the boundary of two distinct faces if and only if e is in a cycle in G. Moreover, if G is not a forest, then the boundary of every face contains a cycle.

Proof. Suppose e is in a cycle C. Then the drawing of C is a simple closed (polygonal) curve in the plane, which separates the plane into its inside and outside. (This is the easy part of the Jordan curve theorem.) The face on one side of e is inside, the other outside.

In the other direction, let F and F' be the faces with e in the boundary. Let H be the spanning subgraph of G consisting of all edges h such that h is in the boundary of F and some other face, i.e., h separates F from non-F. Going around a small circle centred at a vertex v, so that we cross each of the d(v) edges incident with v exactly once and do not cross any other edges, we enter and leave F the same number of times. Thus $d_H(v)$ is even. It is easy to check (exercise) that in a graph with all degrees even, every edge is in a cycle. So e is in a cycle in H, and hence in G.

For the last part, if G is not a forest, then it contains a cycle and so has more than one face. For any face F define H as above; then H contains a cycle which consists of edges in the boundary of F.

A bridge in a graph G is an edge whose deletion would disconnect the component of G that it lies in. Note that e is a bridge if and only if e is not in any cycle. The result above shows that that in a plane graph, e has the same face on both sides if and only if it is a bridge. Note that being a bridge is an abstract graph property, that does not depend on the drawing in the plane.

Theorem 6.2 (Euler's Formula). Let G be a connected plane graph with n vertices, m edges and f faces. Then

$$n - m + f = 2.$$

Proof. By induction on f. If f = 1 then G does not contain a cycle, so it is a tree, and m = n - 1.

Suppose now that $f \ge 2$ and the result holds for smaller values of f. Pick an edge e in the boundary of two faces. By Lemma 6.1, there is a cycle C in G containing e. Thus G - e is connected. When we delete e from the drawing, two faces join up to form a new face, while all other faces remain unchanged. So by induction n - (m - 1) + (f - 1) = 2 and hence n - m + f = 2 as required.

Corollary 6.3. Let G be a planar graph with $n \ge 3$ vertices. Then $e(G) \le 3n - 6$.

Proof. We may assume G is connected (otherwise consider each component and sum) and not a tree (as $n-1 \leq 3n-6$ for $n \geq 3$). Let m = e(G). Consider a drawing of G in the plane, with f faces F_1, \ldots, F_f . Let $e(F_i)$ be the number of edges in the boundary of F_i

counting any bridges twice. Since each non-bridge is in the boundary of two faces and each bridge of only one, we have $\sum_i e(F_i) = 2m$. By the last part of Lemma 6.1, $e(F_i) \ge 3$ for every face F_i , so $3f \le 2m$. Hence $2 = n - m + f \le n - m + 2m/3 = n - m/3$ and the result follows.

We now see that K_5 is not planar, since $e(K_5) = 10 > 9 = 3 \times 5 - 6$. It is an exercise to show that any triangle-free planar graph with $n \ge 3$ vertices has at most 2n - 4 edges; this shows that $K_{3,3}$ is not planar.

Dual graphs

Slightly informally, a multigraph consists of a set V of vertices and a set E of edges, where each $e \in E$ either joins some vertex v to itself (such an edge is called a *loop*) or joins some (unordered) pair $\{u, v\}$ of vertices. There may be several edges joining the same pair of vertices, and there may be several loops at a given vertex v. (Formally, we may define a multigraph as a triple (V, E, ϕ) , where V and E are finite sets, and $\phi: E \to {V \choose 2} \cup {V \choose 1}$ encodes the ends (or end for a loop) of an edge $e \in E$.) It is clear how to extend the definition of a drawing in the plane to multigraphs; for example, a loop at v is drawn as a (polygonal) simple closed curve from x_v to itself meeting the other edges only at x_v .

If G is a plane (multi-)graph then G has a dual G^* obtained as follows: take one vertex F^* for each face F of G, and one edge e^* for each edge e of G, joining the vertices F_1^* and F_2^* corresponding to the faces F_1 and F_2 of G on the two sides of e. (For a bridge e, $F_1 = F_2$, so e^* is a loop.) We may draw G^* so that each vertex F^* is a point in the corresponding face F of G, and each edge e^* crosses the corresponding edge e of G at one point, and is otherwise disjoint from G. We note that G^* is always connected, and if G is connected, then it is easy to check that G^* has one face for every vertex of G, and indeed that $(G^*)^*$ is isomorphic to G. Note that the dual of a connected simple graph (i.e., a graph – no loops or multiple edges) may be a multigraph.



Figure 11: A planar graph and its dual. The dual is a multigraph.

A map is a connected bridgeless plane (multi-)graph. One of the most famous problems in graph theory, posed in 1852, is: can the faces of every map be coloured with 4 colours so that faces sharing an edge get different colours? Taking duals, it is not hard to check that this is equivalent to asking whether every planar (simple) graph G has $\chi(G) \leq 4$. The answer is yes; the result is known as the 'Four Colour Theorem'.

If G is planar and has $n \ge 3$ vertices, then $e(G) \le 3n - 6$, so $\sum_v d(v) = 2e(G) < 6n$, and G must have $\delta(G) \le 5$. It follows easily by induction on |G| that every planar graph G has $\chi(G) \le 6$. With not too much work, we can improve this by one, to obtain the 'Five Colour Theorem'.

Theorem 6.4 (Heawood, 1890). If G is planar then $\chi(G) \leq 5$.

Proof. We argue by induction on n = |G|. If $n \leq 5$ then the result is trivial, so suppose G is planar and has $n \geq 6$ vertices, and every planar graph with fewer vertices is 5-colourable.

As shown above, G has some vertex v with $d(v) \leq 5$. Draw G in the plane, and let c be a 5-colouring of the plane graph G - v. If any of the 5 colours does not appear on a neighbour of v we can extend the colouring to G, and we are done. So we may assume that d(v) = 5 and that the colours of the neighbours of v are distinct. Let the neighbours of v be v_1, v_2, \ldots, v_5 in cyclic order, and without loss of generality suppose that $c(v_i) = i$.



Let H be the subgraph of G - v induced by the vertices with colour 1 or 3. If v_1 and v_3 are in different components of H then, swapping colours 1 and 3 in the component of H containing v_3 , say, we find a new 5-colouring c' of G - v in which $c'(v_1) = c'(v_3) = 1$; this colouring extends to all of G and we are done. Thus we may assume there exists a path P_1 in G - v joining v_1 to v_3 in which all vertices have colour (in c) 1 or 3. Similarly, there exists a path P_2 in G - v joining v_2 to v_4 in which all vertices have colour (in c) 2 or 4. The paths P_1 and P_2 are vertex disjoint, so in the drawing they do not cross. Since the cycle vP_1v separates the plane and P_2 starts and ends on different sides of this cycle this gives a contradiction.

The paths described above are often called 'Kempe chains'. Kempe thought he had proved the *four* colour theorem in 1879. The theorem was actually first proved by Appel and Haken in 1977 making extensive use of computers. A simpler, but still computer-based, proof was given by Robertson, Sanders, Seymour and Thomas in 1997. As of today there is no simple proof known.

7 Flows, connectivity and matchings

Imagine a road network in which each road has a certain 'capacity', or maximum flow in cars/hour. How can we work out the maximum traffic flow from one or more 'sources' to one or more 'sinks' or target destinations? Since the capacity of a road may not be the same in the two directions (for example if it is one-way) it makes sense to consider this question in the context of *directed graphs*.

Formally, a directed graph or digraph $\overrightarrow{G} = (V, \overrightarrow{E})$ consists of a set V, the set of vertices, and a set \overrightarrow{E} of ordered pairs of distinct elements of V, the (directed) edges. We write \overrightarrow{E} to remind ourselves the graph is directed; often the edge-set is just denoted E. We think of $(x, y) \in \overrightarrow{E}$ as an edge from x to y, and write \overrightarrow{xy} or simply xy. Note that a directed graph cannot contain more than one edge from x to y, but can contain both edges xy and yx. We write

$$N^+(x) = \{ y \in V : xy \in \overrightarrow{E} \}$$

for the *out-neighbourhood* of $x \in V$, and

$$N^{-}(x) = \{ y \in V : yx \in \overrightarrow{E} \}$$

for its *in-neighbourhood*.

A flow in G with source s and sink t is a function $f \colon \overrightarrow{E} \to [0, \infty)$ such that for every $x \in V \setminus \{s, t\}$ we have

$$\sum_{y\in N^+(x)}f(xy)=\sum_{y\in N^-(x)}f(yx)$$

i.e., the flow out of x is equal to the flow into x. Here s and t are distinct vertices.

Given any function $f \colon \vec{E} \to \mathbb{R}$, for $x \in V$ let

$$I_f(x) = \sum_{y \in N^+(x)} f(xy) - \sum_{y \in N^-(x)} f(yx).$$

We may think of $I_f(x)$ as the amount of flow that must be injected into the graph at x to maintain balance; in a flow, $I_f(x) = 0$ for $x \in V \setminus \{s, t\}$.

For any flow (or, indeed, any function on \vec{E}),

$$\sum_{x \in V} I_f(x) = \sum_{x \in V} \left(\sum_{y \in N^+(x)} f(xy) - \sum_{y \in N^-(x)} f(yx) \right) = 0,$$

since for every $uv \in \vec{E}$, f(uv) appears exactly twice, once with x = u and once with x = v. For a flow, the terms with $x \neq s, t$ are zero, so $I_f(s) = -I_f(t)$, i.e.,

$$\sum_{y \in N^+(s)} f(sy) - \sum_{y \in N^-(s)} f(ys) = \sum_{y \in N^-(t)} f(yt) - \sum_{y \in N^+(t)} f(ty).$$

In other words, the net flow leaving s equals the net flow arriving at t. This common value is called the *value* of f, and written v(f). (Usually, $v(f) = I_f(s) = -I_f(t)$ is positive – otherwise we would regard the flow as having t as source and s as sink.) We can think of flow as being 'conserved' at every vertex, but with flow v(f) injected into the graph at s and flow v(f) extracted at t.

A capacity function on a directed graph $G = (V, \vec{E})$ is just a function $c: \vec{E} \to [0, \infty)$. A flow f is feasible (w.r.t. c) if $f(xy) \leq c(xy)$ for every $xy \in \vec{E}$. The key question in the theory of flows is: what is the maximum value of a feasible flow in a given graph with given source s, sink t and capacity function c? To avoid repeating the definitions, we shall call a directed graph with a given sink, source and capacity function a *network*. (Of course, the word 'network' has many different meanings, depending on the context.) When we say f is a flow in a given network, it is always understood that f is feasible.

Given sets S and T of vertices of a directed graph (V, \vec{E}) , let $\vec{E}(S, T) = \{xy \in \vec{E} : x \in S, y \in T\}$ be the set of edges from S to T.

A *cut* in a network is a partition of the vertex set into disjoint sets S and T with $s \in S$ and $t \in T$. (Alternatively, we may say that a corresponding set $\overrightarrow{E}(S,T)$ of edges is a cut.) The *capacity* of a cut (S,T) is

$$c(S,T) = \sum_{xy \in \overrightarrow{E}(S,T)} c(xy),$$

i.e., the maximum conceivable flow from S to T (ignoring what happens within S and T). Clearly, in any feasible flow $f, v(f) \leq c(S, T)$. Indeed,

$$v(f) = I_f(s) = \sum_{x \in S} I_f(x) = \sum_{x \in S} \left(\sum_{y \in N^+(x)} f(xy) - \sum_{y \in N^-(x)} f(yx) \right)$$
$$= \sum_{xy \in \overrightarrow{E}(S,T)} f(xy) + \sum_{xy \in \overrightarrow{E}(S,S)} f(xy) - \sum_{yx \in \overrightarrow{E}(T,S)} f(yx) - \sum_{yx \in \overrightarrow{E}(S,S)} f(yx)$$
$$= \sum_{xy \in \overrightarrow{E}(S,T)} f(xy) - \sum_{yx \in \overrightarrow{E}(T,S)} f(yx) \leqslant c(S,T).$$
(6)

Thus the maximum value of a feasible flow is at most the minimum capacity of a cut. The remarkable 'max-flow min-cut' theorem tells us that we have equality.

Theorem 7.1 (Max-Flow Min-Cut). In any network $(\overrightarrow{G}, s, t, c)$ we have

$$\sup \{v(f) : f \text{ is a feasible flow}\} = \min \{c(S,T) : (S,T) \text{ is a cut}\}.$$

Moreover, the supremum is attained.

The key ingredient of the proof is the notion of an *augmenting path*, or 'slack path'. Let f be a flow in a network. We say that an ordered pair (x, y) is ε -slack if either $xy \in \overrightarrow{E}$ and $f(xy) \leq c(xy) - \varepsilon$ or $yx \in \overrightarrow{E}$ and $f(yx) \geq \varepsilon$ (or both). A path $x_0x_1 \cdots x_r$ in the undirected graph associated to \overrightarrow{G} is ε -slack if $x_{i-1}x_i$ is ε -slack for $1 \leq i \leq r$, and slack (or augmenting) if it is slack for some $\varepsilon > 0$.

Lemma 7.2. Let f be a flow in a network. If $x_0x_1 \cdots x_r$ is an ε -slack path with $x_0 = s$ and $x_r = t$ then v(f) is not maximal; in particular, there is a flow f' with $v(f') = v(f) + \varepsilon$.

Proof. For each i we can either increase the flow along $x_{i-1}x_i$ by ε , or decrease the flow along x_ix_{i-1} by ε . Doing either increases $I_f(x_{i-1})$ by ε and decreases $I_f(x_i)$ by the same amount. Making such a change for each i = 1, 2, ..., r, we see that $I_f(x)$ is unchanged for every $x \neq s, t$ (so we still have a flow), and that $v(f) = I_f(s) = -I_f(t)$ is increased by ε .

Proof of Theorem 7.1. First, we show that the supremum is attained. As noted earlier, for any flow f and cut (S,T) we have $v(f) \leq c(S,T)$. In particular, $v(f) \leq \sum_{y \in N^+(s)} c(sy) < \infty$ so the set $\{v(f) : f \text{ a flow}\}$ is bounded. So there are flows f_i with $v(f_i) \to M < \infty$, where M is the supremum. Let $xy \in \vec{E}$. Then, passing to a subsequence, we may assume that $f_i(xy)$ converges. Repeating this for each edge, we find a (sub)sequence of flows with $v(f_i) \to M$ such that $f_i(xy)$ converges for each $xy \in \vec{E}$. But then $f(xy) = \lim_{i \to \infty} f_i(xy)$ defines a flow with value $\lim_{i \to \infty} v(f_i) = M$.

Let f be a flow attaining the supremum. It suffices to find a cut with capacity v(f). Let

 $S = \{x \in V : \text{ there is a slack path from } s \text{ to } x\},\$

and let $T = V \setminus S$. Clearly $s \in S$ (consider a path of length 0). By Lemma 7.2, $t \notin S$. Thus (S,T) is a cut. Suppose $x \in S$ and $y \in T$ with (x, y) slack. Then taking a slack path $s = x_0 \cdots x_r = x$ and appending $x_{r+1} = y$ gives a slack path ending at y, contradicting $y \in T$. Hence, for every $xy \in \overrightarrow{E}(S,T)$ we have f(xy) = c(xy), and for every $yx \in \overrightarrow{E}(T,S)$ we have f(yx) = 0. I.e., equality holds in (6), so c(S,T) = v(f).

A maximal flow is one with maximum value. A function (here f or c) is *integral* if all its values are integers.

Theorem 7.3. Let $(\overrightarrow{G}, s, t, c)$ be a network in which the capacity function c is integral. Then there is a maximal flow f which is integral.

Proof. We have essentially described an algorithm to find such an f; the key point is that if the capacity function c and flow f are integral, then any slack path is 1-slack. Start with the flow with f(xy) = 0 for all edges, and repeat the following: if there is a slack (and hence 1-slack) path from s to t augment the flow along this path by 1 as above, obtaining a new integral flow with larger value; repeat. Otherwise, by the last part of the proof above, there is a cut (S,T) with v(f) = c(S,T), so f is maximal.

The algorithm defined above is in fact reasonably efficient: it is easy to check for the existence of slack paths by (for example) breadth-first search.

A directed path in a directed graph $\overrightarrow{G} = (V, \overrightarrow{E})$ is a sequence $x_0x_1 \cdots x_r$ of distinct vertices such that $x_0x_1, \ldots, x_{r-1}x_r \in \overrightarrow{E}$. A set $\overrightarrow{X} \subseteq \overrightarrow{E}$ of edges separates s from t if $\overrightarrow{G} - \overrightarrow{X}$ contains no directed path (or, equivalently, no directed walk) from s to t. If (S,T)is a cut, then $\overrightarrow{E}(S,T)$ separates s from t. Conversely, if \overrightarrow{X} separates s from t then it contains $\overrightarrow{E}(S,T)$ for some cut (S,T) – for example, take S to be the set of vertices x such that $\overrightarrow{G} - \overrightarrow{X}$ contains a directed s-x path. Let $c(\overrightarrow{X}) = \sum_{xy \in \overrightarrow{X}} c(xy)$. Then we see that

$$\min\left\{c(S,T):(S,T) \text{ is a cut}\right\} = \min\left\{c(\overrightarrow{X}): \overrightarrow{X} \text{ separates } s \text{ from } t\right\}.$$
(7)

This gives an alternative formulation of the max-flow min-cut theorem. Note, however, that cuts arise in the proof in an essential way, and it is necessary to consider reducing flow along backwards edges as well as increasing it along forwards ones.

The max-flow min-cut theorem has many variants, some of which we leave as exercises. For example, we may consider several sources s_1, \ldots, s_k and several sinks t_1, \ldots, t_ℓ . In this context, a cut (S, T) is a partition of the vertices with all sources in S and all sinks in T. A flow must satisfy $I_f(x) = 0$ for every vertex that is neither a source nor a sink, and its value is $\sum_{i=1}^k I_f(s_i)$. Theorems 7.1 and 7.3 apply *mutatis mutandis* to this setting.

Another important variation allows some edges to have infinite capacity, meaning that the flow along xy can take any finite value. The results hold in this setting too, with the proviso that if there is no cut with finite capacity, then $\{v(f)\}$ is unbounded, so of course there is no flow with maximum value.

One more substantial variant is to impose capacity restrictions on the vertices rather than edges. Let \overrightarrow{G} be a directed graph with source s and sink t, and let c be a (vertex) capacity function assigning every vertex $x \neq s, t$ a capacity $c(x) \in [0, \infty)$. A flow in \overrightarrow{G} is feasible if for each vertex $x \neq s, t$ we have

$$\sum_{y \in N^-(x)} f(yx) = \sum_{y \in N^+(x)} f(xy) \leqslant c(x),$$

i.e., the flow through x is at most c(x). (The equality is the definition of a flow; feasibility is the inequality.)

A vertex-cut is a set $S \subseteq V \setminus \{s, t\}$ of vertices such that in $\overrightarrow{G} - S$ there is no directed path from s to t. The capacity of S is $\sum_{x \in S} c(x)$.

Theorem 7.4. Let \overrightarrow{G} be a directed graph with source s, sink t and vertex capacity function c. Then the maximum value of a feasible flow from s to t is the minimum capacity of any vertex-cut. Furthermore, if c is integral, then there is a flow with maximum value that is integral.

Proof. Rather than modify the proof of Theorem 7.1, we modify the network so that we can apply that result.

Form a directed graph \overrightarrow{H} with source s and sink t by replacing each vertex $x \neq s, t$ by two vertices x^- and x^+ joined by an edge x^-x^+ with capacity c(x). For each edge of \overrightarrow{G} there is an edge of \overrightarrow{H} with infinite (or very large) capacity; edges that start/end at s/tin \overrightarrow{G} do so in \overrightarrow{H} ; every edge of \overrightarrow{G} ending at $x \neq s, t$ now ends at x^- , and every edge starting at x now starts at x^+ . It is easy to check that feasible flows in \overrightarrow{G} are in 1-to-1 correspondence with feasible flows in \overrightarrow{H} . In \overrightarrow{H} , a set \overrightarrow{X} of edges with $c(\overrightarrow{X})$ finite must be of the form $\overrightarrow{X}_S = \{x^-x^+ : x \in S\}$ for some $S \subseteq V \setminus \{s,t\}$. Moreover, \overrightarrow{X}_S is separating if and only if S is a vertex-cut. The result thus follows from Theorem 7.1 and (7) and, for integrality, Theorem 7.3.

Connectivity and Menger's Theorem

Let G be an (undirected) graph and $S \subseteq V(G)$. We say that S separates G if G - S is disconnected. For vertices x, y of G, S separates x and y if they are in different components of G - S.

For a non-negative integer k, a graph G is k-connected if $|G| \ge k+1$ and no set of (at most) k-1 vertices separates G. (Every graph is 0-connected. A graph G is 1-connected iff it is connected and $|G| \ge 2$. The only k-connected graph with |G| = k+1 is K_{k+1} .)

The (vertex) *connectivity* of a graph G is defined as

$$\kappa(G) = \max\{k : G \text{ is } k \text{-connected}\}.$$

Equivalently, $\kappa(G)$ is the minimum number of vertices that must be deleted to either disconnect G, or reduce it to a single vertex. It follows easily from the definition that $\kappa(G-x) \ge \kappa(G) - 1$, and that if H is a spanning subgraph of G then $\kappa(G) \ge \kappa(H)$. It is an exercise to check that if e is an edge of G then $\kappa(G-e) \ge \kappa(G) - 1$.

We now define a 'local' version of connectivity. For distinct *non-adjacent* vertices x and y of G we write

$$\kappa(x, y) = \kappa_G(x, y) = \min\{|S| : S \text{ separates } x \text{ and } y\}.$$

Note that adjacent vertices can never be separated by deleting other vertices. Also, it is easy to check that for any non-complete graph G,

$$\kappa(G) = \min_{xy \in E(\overline{G})} \kappa_G(x, y).$$

Two distinct x-y paths are *independent* (or *internally vertex-disjoint*) if the only vertices they share are x and y. A set of x-y paths is independent if the paths are pairwise independent.

Theorem 7.5 (Menger's Theorem). Let x and y be distinct non-adjacent vertices of G. Then the maximum size of an independent set of x-y paths is $\kappa_G(x, y)$.

Proof. If there are k independent x-y paths then $\kappa_G(x, y) \ge k$ as we must remove at least one vertex from each path to separate x from y. So it is enough to show that there are $\kappa_G(x, y)$ independent paths.

Turn G into a network with source x and sink y by replacing each edge uv by two directed edges \vec{uv} and \vec{vu} , and assigning each vertex other than x and y capacity 1. Then a vertex-cut S is simply a set of vertices separating x and y, and its capacity is just |S|. Hence, by Theorem 7.4, there is an *integral* flow f from x to y with value $\kappa_G(x, y)$. Given the vertex capacities, f can only take values 0 and 1, so f corresponds to a set of edges consisting of independent x-y paths and perhaps some directed cycles. The value of f is the number of paths, so there are $\kappa_G(x, y)$ paths as required.

Corollary 7.6. A graph G is k-connected iff $|G| \ge k+1$ and every pair of non-adjacent vertices is joined by k independent paths.

We can also define *edge connectivity*, and prove a form of Menger's Theorem for edgedisjoint paths. More specifically, we say a graph with at least 2 vertices is k-edge-connected if removing at most k-1 edges cannot disconnect the graph. We define the edge connectivity as

 $\lambda(G) = \max\{k : G \text{ is } k \text{-edge-connected.}\}$

Given distinct vertices x and y, define $\lambda_G(x, y)$ to be the minimum number of edges one needs to remove from G so that x and y are no longer in the same component (i.e., the minimum size of a separating set of edges). Clearly $\lambda(G) = \min_{x \neq y} \lambda_G(x, y)$.

Theorem 7.7 (Menger's Theorem, edge version). Let x and y be distinct vertices of G. Then the maximum number of pairwise edge-disjoint x-y paths is $\lambda_G(x, y)$.

Proof. Exercise.

Hall's Theorem

A matching M in a graph G is a set of pairwise disjoint edges of G; its size |M| is the number of edges. Let G be a bipartite graph with vertex classes V_1 and V_2 . A complete matching from V_1 to V_2 is a matching such that every vertex in V_1 is incident with some edge in the matching, i.e., a matching of size $|V_1|$.

Given a set S of vertices in a graph G, we write $N(S) = \bigcup_{s \in S} N(s)$ for the set of vertices with at least one neighbour in S.⁹

Theorem 7.8 (Hall's Marriage Theorem). Let G be a bipartite graph with bipartition (V_1, V_2) . Then G contains a complete matching from V_1 to V_2 iff $|N(S)| \ge |S|$ for each $S \subseteq V_1$.

⁹Some authors exclude elements of S from N(S). In the present context, where G is bipartite and $S \subseteq V_1$, it makes no difference.

The condition that $|N(S)| \ge |S|$ for each $S \subseteq V_1$ is called *Hall's condition*. It is trivially necessary as each element of S must be matched to a different element of N(S). We give two proofs of sufficiency.

Proof. We can deduce the result from Menger's Theorem. Instead, here is an outline of a proof directly from Theorem 7.3.

Form a directed graph by orienting every edge from V_1 to V_2 , and adding a new vertex s with an edge sx for every $x \in V_1$ and a new vertex t with edges $xt, x \in V_2$. Assign all the new edges capacity 1, and the edges from V_1 to V_2 some very large (or infinite) capacity; $|V_1| + 1$ is large enough. Let (S, T) be a cut, and let $S_i = S \cap V_i$ and $T_i = T \cap V_i$. Either (i) $\vec{E}(S,T)$ contains some edge from V_1 to V_2 . Then $c(S,T) > |V_1|$. Or (ii) not, i.e., $N(S_1) \subseteq S_2$. Then



by Hall's condition. Hence the capacity of any cut is at least $|V_1|$, so by Theorem 7.3 there is an integral flow f with value $|V_1|$. But it is easy to check that f can only take the values 0 and 1, and that the edges e from V_1 to V_2 with f(e) = 1 correspond to a complete matching in G.

Here is a direct proof of Hall's Theorem.

Proof. We argue by induction on $n = |V_1|$. If n = 1, the result is trivial. For the induction step, let $n \ge 2$ and suppose that the result holds for all graphs with $|V_1| < n$. Consider a graph G with $|V_1| = n$ and assume that Hall's condition holds. There are two cases.

(a) Suppose first that |N(S)| > |S| for each $\emptyset \neq S \subsetneq V_1$. Let xy be any edge of G with $x \in V_1$ and $y \in V_2$. Form G' by deleting the vertices x and y from G. Then G' satisfies Hall's condition (since if $\emptyset \neq S \subseteq V_1 \setminus \{x\}$ then $|N'(S)| = |N(S) \setminus \{y\}| \ge |N(S)| - 1 \ge |S|$), and so by induction G' has a complete matching from $V_1 \setminus \{x\}$ to $V_2 \setminus \{y\}$. Now adding the edge xy gives the required matching.

(b) If case (a) does not hold then |N(S)| = |S| for some $\emptyset \neq S \subsetneq V_1$. The bipartite subgraph induced by $S \cup N(S)$ still satisfies Hall's condition, so by induction there is a complete matching M_1 from S to N(S). Now consider $T = V_1 \setminus S$ and $U = V_2 \setminus N(S)$. We shall see that the bipartite subgraph H induced by $T \cup U$ also satisfies Hall's condition. For each $A \subseteq T$ we have



since $|N(A \cup S)| \ge |A \cup S|$ and |N(S)| = |S|. So Hall's condition holds in H, and by induction there is a complete matching M_2 from T to U. Then $M_1 \cup M_2$ is the required matching from V_1 to V_2 .

Tutte's 1-factor Theorem

Although especially natural in bipartite graphs, it makes perfect sense to consider matchings in general graphs. A *k*-factor in a graph G is a spanning *k*-regular subgraph. Thus a 1-factor is exactly the same as a matching covering all vertices.

We call a component of a graph G odd if it has an odd number of vertices, and even otherwise. Let q(G) denote the number of odd components of G, and note that $q(G) \equiv |G| \mod 2$.

Theorem 7.9 (Tutte's 1-factor theorem). A graph G has a 1-factor if and only if, for every $S \subseteq V(G)$, we have

$$q(G-S) \leqslant |S|. \tag{8}$$

Proof. In any 1-factor (complete matching) M, every odd component C of G - S contains at least one vertex paired with some vertex outside C. Since the only edges leaving C in G go to S, a vertex of C must be paired with a vertex of S, and so $|S| \ge q(G - S)$. This shows that (8) is necessary. We prove sufficiency by induction on |G|. The case |G| = 1(or |G| = 2) is trivial.

Suppose then that G satisfies (8), and that the result holds for all smaller graphs. Taking $S = \emptyset$ in (8) we see that q(G) = 0, and in particular |G| is even. Also, for any vertex v of G, q(G - v) is odd (since |G - v| is). Hence $q(G - v) \ge 1$ and (since we are assuming (8)), for $S = \{v\}$ we have q(G - S) = |S|.

Let S be a subset of V(G) for which q(G - S) = |S| with s = |S| maximal. From the above, $s \ge 1$, so S is not empty. Let O_1, \ldots, O_s be the odd components of G - S and $E_1, \ldots, E_k, k \ge 0$, the even components (if there are any). We shall prove the following three statements.

(a) each E_i has a 1-factor,

- (b) if v is any vertex of any O_i , then $O_i v$ has a 1-factor, and
- (c) there is a matching s_1v_1, \ldots, s_sv_s in G such that $\{s_1, \ldots, s_s\} = S$ and $v_i \in O_i$ for $1 \leq i \leq s$.

Clearly, if (a)–(c) hold then G has a 1-factor: apply (c) first, then (b) and (a). It remains to prove (a)-(c).

To see (a), let $A \subseteq V(E_i)$. The components of $G - (A \cup S)$ are O_1, \ldots, O_s , all E_j other than E_i , and the components of $E_i - A$, so $q(G - (A \cup S)) = s + q(E_i - A)$ and

$$q(E_i - A) = q(G - (A \cup S)) - s \leq |A \cup S| - s = |A| + s - s = |A|.$$

Hence E_i satisfies (8) and by induction E_i has a 1-factor.

For (b), let v be a vertex of O_i . Let $A \subseteq V(O_i - v)$. Then the components of $G - (A \cup \{v\} \cup S)$ are the E_j , all O_j other than O_i , and the components of $(O_i - v) - A$. Hence

$$q((O_i - v) - A) = q(G - (A \cup \{v\} \cup S)) - (s - 1) < |A \cup \{v\} \cup S| - s + 1 = |A| + 2,$$

where the inequality is from (8) and the maximality of |S|. Modulo 2,

$$q((O_i - v) - A) \equiv |(O_i - v) - A| = |O_i| - 1 - |A| \equiv |A|,$$

since O_i is odd. Since x < y + 2 and $x \equiv y \mod 2$ imply $x \leq y$, it follows that $q((O_i - v) - A) \leq |A|$, so $O_i - v$ satisfies (8), and (b) follows by induction.

Finally, for (c) let H be the bipartite graph with $V_1 = S$ and $V_2 = \{o_1, \ldots, o_s\}$, with an edge xo_i whenever $x \in S$ and there is at least one edge in G from x to O_i . It suffices to find a complete matching in H, so we check Hall's condition. Let $A \subseteq V_1 = S$. If $o_i \in V_2 \setminus N_H(A)$ then in G there are no edges from A to O_i , so O_i is a component of $G - (S \setminus A)$. Hence, $q(G - (S \setminus A)) \ge |V_2 \setminus N_H(A)| = s - |N_H(A)|$. Thus, by (8),

$$|N_H(A)| \leq q(G - (S \setminus A)) \leq |S \setminus A| = s - |A|.$$

Hence $|N_H(A)| \ge |A|$, so Hall's condition holds in H, and by Hall's Theorem H has the required complete matching.

8 Extremal Problems

If G has a subgraph isomorphic to H we say 'G contains (a copy of) H' for short, and sometimes write $G \supseteq H$.¹⁰ For a graph H with e(H) > 0 and $n \ge 1$ an integer, define

$$ex(n, H) = \max \left\{ e(G) : |G| = n, G \text{ contains no copy of } H \right\}$$

and

$$\mathrm{EX}(n,H) = \{G : |G| = n, e(G) = \mathrm{ex}(n,H), G \text{ contains no copy of } H\}.$$

The graphs in EX(n, H) are called the *extremal graphs*; we often describe EX(n, H) by listing one graph from each isomorphism class. ex(n, H) is the *extremal number* for H (a function of n, of course).

For example, if G contains no copy of P_3 then all edges must be disjoint. Thus $ex(n, P_3) = \lfloor \frac{n}{2} \rfloor$, and $EX(n, P_3)$ is $\{\frac{n}{2}K_2\}$ if n is even and $\{\frac{n-1}{2}K_2 \cup K_1\}$ if n is odd, where mK denotes the disjoint union of m copies of K.

What is $ex(n, K_3)$? Good candidate extremal graphs are the complete bipartite graphs $K_{k,n-k}$; to maximize the number k(n-k) of edges we take $k = \lfloor n/2 \rfloor$. More generally, what is $ex(n, K_{r+1})$?

A graph G is *r*-partite if V(G) is the disjoint union of r sets V_1, \ldots, V_r (the vertex classes) with $e(G[V_i]) = 0$ for each *i*, i.e., no edges within each V_i . In other words, all edges go between parts. This is exactly the same as saying that G is r-colourable. A graph G is complete r-partite if in addition every possible edge between parts is present.

Note that empty parts are allowed: the key point is that inside any part with at least two vertices, edges are forbidden. Clearly, any *r*-partite graph does not contain a copy of K_{r+1} as two of the vertices of the K_{r+1} would have to lie in the same part.

¹⁰Perhaps we should not write this, since in other contexts it means that H itself is a subgraph of G, i.e., that the particular vertices and edges of H are present in G. But usually it is clear from context whether or not we are considering isomorphic copies.

Before continuing we make a trivial observation: if a_1, \ldots, a_r are integers with average $\bar{a} = \frac{1}{r} \sum a_i$ then all a_i are within 1 of each other (i.e., $\max a_i \leq \min a_i + 1$) if and only if every a_i is equal to $\lfloor \bar{a} \rfloor$ or $\lceil \bar{a} \rceil$. (There are two cases: all $a_i = m = \bar{a}$ for some integer m, or some $a_i = m$, some $a_i = m + 1$; then $m < \bar{a} < m + 1$.) Moreover, given r and $\sum a_i$, there is only one way (up to reordering) to choose the a_i so that these conditions hold. I.e., there is only one way to divide a given number of (indivisible) objects among r people 'as fairly as possible'.

For $n, r \ge 1$, the Turán graph $T_r(n)$ is the complete r-partite graph on n vertices with the vertex class sizes as equal as possible, i.e., each is $\lfloor \frac{n}{r} \rfloor$ or $\lceil \frac{n}{r} \rceil$. The Turán number $t_r(n)$ is $e(T_r(n))$. Note that if $n \le r$ then $T_r(n) = K_n$.

For example, $T_2(n)$ is $K_{\lfloor n/2 \rfloor, \lceil n/2 \rceil}$. Thus $t_2(n)$ is $\frac{n^2}{4}$ if n is even, and $\frac{n^2-1}{4}$ if n is odd: $t_2(n) = \lfloor \frac{n^2}{4} \rfloor$.

 $T_3(10)$ has class sizes 3, 3 and 4, and 9 + 12 + 12 = 33 edges, so $t_3(10) = 33$. $T_1(n)$ has no edges, so $t_1(n) = 0$.

Facts about Turán graphs

- 1. Among all r-partite graphs with n vertices, $T_r(n)$ is the unique (up to isomorphism) one with the largest number of edges. Indeed, only complete r-partite graphs are candidates, and if two classes differ in size by 2 or more, moving a vertex from the larger to the smaller gains at least one edge (since it reduces the number of vertices it is not adjacent to).
- 2. Since each vertex class has size $\lfloor \frac{n}{r} \rfloor$ or $\lceil \frac{n}{r} \rceil$, $\delta(T_r(n)) = n \lceil \frac{n}{r} \rceil$ and $\Delta(T_r(n)) = n \lfloor \frac{n}{r} \rfloor$, so $\Delta - \delta \leq 1$. Hence $\delta(T_r(n)) = \lfloor d(T_r(n)) \rfloor$ and $\Delta(T_r(n)) = \lceil \overline{d}(T_r(n)) \rceil$, where $\overline{d}(G) = 2e(G)/|G|$ is the average degree of a graph G.
- 3. To get from $T_r(n)$ to $T_r(n-1)$ we delete any vertex from a largest vertex class, i.e., any vertex of minimum degree. So $t_r(n) - \delta(T_r(n)) = t_r(n-1)$.

Theorem 8.1 (Turán's Theorem). For all positive integers n and r we have

$$ex(n, K_{r+1}) = t_r(n)$$
 and $EX(n, K_{r+1}) = \{T_r(n)\}.$

Proof. We fix r and use induction on n. If $n \leq r$ then $ex(n, K_{r+1}) = \binom{n}{2} = t_r(n)$, and $EX(n, K_{r+1}) = \{K_n\} = \{T_r(n)\}$, as required.

Now let n > r and suppose that the result holds for n - 1. Let G be a graph with n vertices and $t_r(n)$ edges containing no copy of K_{r+1} . We will show that $G \cong T_r(n)$, from which the result follows. (If H had |H| = n, $e(H) > t_r(n)$ and contained no copy of K_{r+1} , then some spanning subgraph G would have $t_r(n)$ edges; so $G \cong T_r(n)$ and then H would contain a copy of K_{r+1} as adding any edge to $T_r(n)$ forms a K_{r+1} .)

First note that by Fact 2

$$\delta(G) \leqslant \lfloor \bar{d}(G) \rfloor = \lfloor \bar{d}(T_r(n)) \rfloor = \delta(T_r(n)).$$

Let $v \in V(G)$ have degree $d(v) = \delta(G)$. Then for H = G - v we have

$$e(H) = e(G) - d(v) = t_r(n) - \delta(G) \ge t_r(n) - \delta(T_r(n)) = t_r(n-1),$$

using Fact 3. But H contains no K_{r+1} so by the induction hypothesis, $H \cong T_r(n-1)$.

Now v cannot have a neighbour in each vertex class of H (or we would get a copy of K_{r+1}), so its neighbours must miss some vertex class V_i completely. Adding v to this class, we see that G is r-partite. Now by Fact 1, $G \cong T_r(n)$.

The density of a graph G is $e(G)/\binom{|G|}{2}$. For r fixed and $n \to \infty$, $T_r(n)$ has density $1 - \frac{1}{r} + o(1)$. Thus by Turán's Theorem, $ex(n, K_{r+1})/\binom{n}{2} = 1 - \frac{1}{r} + o(1)$. We say that K_{r+1} 'appears' at density $1 - \frac{1}{r}$. Thus K_3 appears at density $\frac{1}{2}$, K_4 at density $\frac{2}{3}$, K_5 at density $\frac{3}{4}$, and so on.

What about other graphs? If $\chi(H) \ge r+1$, then, since H is not r-partite, $T_r(n)$ contains no copy of H. Thus for any H, letting $r = \chi(H) - 1$ we have

$$\exp(n, H) \ge t_r(n) = (1 - \frac{1}{r} + o(1))\binom{n}{2}.$$

In particular, H cannot appear before the density $1 - \frac{1}{r}$ at which K_{r+1} appears. Are there 'big' graphs with a given chromatic number that appear significantly later? Amazingly, the answer turns out to be no.

For $s, t \ge 1$ let $K_s(t)$ be the complete s-partite graph with t vertices in each class. For example, $K_1(t)$ is the empty graph E_t , $K_2(t)$ is the complete bipartite graph $K_{t,t}$, and $K_s(1) = K_s$. In general $K_s(t) = T_s(st)$.

Theorem 8.2 (Erdős–Stone Theorem). Let $r, t \ge 1$ be integers and let $\varepsilon > 0$. There is a constant $n_0 = n_0(r, t, \varepsilon)$ such that every graph G with $n \ge n_0$ vertices and

$$e(G) \ge \left(1 - \frac{1}{r} + \varepsilon\right) \binom{n}{2}$$

contains a copy of $K_{r+1}(t)$.

Note that for t = 1 this follows (and is somewhat weaker than) Turán's theorem. We prove this theorem in two steps. We first show that any graph with a given density contains a subgraph with a relatively large minimum degree.

Lemma 8.3. Let $0 \leq \alpha < \beta \leq 1$. If G is a graph with |G| = n and $e(G) \geq \beta {n \choose 2}$ then G has an (induced) subgraph H with

$$\delta(H) \ge \alpha(|H| - 1)$$

and $|H| \ge \sqrt{\varepsilon}n$, where $\varepsilon = \beta - \alpha$.

Proof. Define a sequence G_n, G_{n-1}, \ldots of graphs with $|G_t| = t$ as follows. Set $G_n = G$. If $\delta(G_t) \ge \alpha(t-1)$ then stop. Otherwise, remove a vertex of G_t with minimum degree to get G_{t-1} . The construction must stop at some point $(G_1$ at the latest); let G_k be the final graph, so $\delta(G_k) \ge \alpha(|G_k| - 1)$ by construction. Now

$$e(G_k) = e(G_n) - \delta(G_n) - \delta(G_{n-1}) - \dots - \delta(G_{k+1})$$

$$\geqslant \beta \binom{n}{2} - \alpha ((n-1) + (n-2) + \dots + k)$$

$$= \beta \binom{n}{2} - \alpha \binom{n}{2} + \alpha \binom{k}{2} \geqslant \varepsilon \binom{n}{2}.$$

Since $e(G_k) \leq \binom{k}{2}$ it follows that $\binom{k}{2} \geq \varepsilon\binom{n}{2}$. Multiplying by $2k/(k-1) \geq 2n/(n-1)$ gives $k^2 \geq \varepsilon n^2$, so $k \geq \sqrt{\varepsilon}n$.

Lemma 8.4. Let $r, t \ge 1$ be integers and let $\varepsilon > 0$. There is a constant $n_1 = n_1(r, t, \varepsilon)$ such that every graph G with $n \ge n_1$ vertices and

$$\delta(G) \ge \left(1 - \frac{1}{r} + \varepsilon\right)(n-1)$$

contains a copy of $K_{r+1}(t)$.

Note that Theorem 8.2 will follow easily: apply Lemma 8.3 with $\beta = 1 - \frac{1}{r} + \varepsilon$ and $\alpha = 1 - \frac{1}{r} + \frac{\varepsilon}{2}$, and then Lemma 8.4 with $\varepsilon/2$ in place of ε .

Proof. We use induction on r, proving the base case and induction step together. More precisely, for $r \ge 1$ let \mathbb{H}_r be the statement that for every $t \ge 1$ and $\varepsilon > 0$ there is an n_1 such that We shall prove \mathbb{H}_r assuming, for $r \ge 2$, that \mathbb{H}_{r-1} holds. Then we will have shown that \mathbb{H}_1 holds, and that \mathbb{H}_{r-1} implies \mathbb{H}_r for all $r \ge 2$, so by induction \mathbb{H}_r holds for all $r \ge 1$. A key point is that in proving \mathbb{H}_r , $r \ge 2$, we must consider all t and all $\varepsilon > 0$; but for a given t we may use the fact that \mathbb{H}_{r-1} holds for some other, perhaps *much* larger value of t.

Let $r \ge 1$. To prove \mathbb{H}_r , let $t \ge 1$ and $\varepsilon > 0$ be given, and let G be a graph with |G| = nand $\delta(G) \ge (1 - \frac{1}{r} + \varepsilon)(n - 1)$. Set

$$T = \lceil 2t/(\varepsilon r) \rceil.$$

If $r \ge 2$ then since

$$\delta(G) \ge (1 - \frac{1}{r} + \varepsilon)(n - 1) \ge (1 - \frac{1}{r - 1} + \varepsilon)(n - 1),$$

we know by the induction hypothesis \mathbb{H}_{r-1} that, if *n* is large enough (depending on r, t, ε), *G* must contain a copy of $K_r(T)$. If r = 1 then $K_r(T)$ is an empty graph with *T* vertices, so if *n* is large enough (i.e., $n \ge T$) then *G* certainly contains a copy of this graph.

In either case, let H be a subgraph of G isomorphic to $K_r(T)$. Denote its vertex classes by S_1, \ldots, S_r and let S = V(H) be their union. Let U be the set of vertices in $V(G) \setminus S$ which have at least t neighbours in each class S_i ; these vertices are the *useful* ones.

If $|U| > (t-1) {T \choose t}^r$ then there are at least t vertices in U that have at least t common neighbours in each S_i , giving a $K_{r+1}(t)$; to see this assign to each $u \in U$ an r-tuple (A_1, \ldots, A_r) where $A_i \subseteq S_i \cap N(u)$ and $|A_i| = t$. As there are ${T \choose t}^r$ possible r-tuples, the average number of times an r-tuple is chosen is $|U|/{T \choose r}^t > t - 1$, and so some r-tuple is chosen $\geq t$ times, i.e., we have t vertices in U all joined to the same copy $H' \subseteq H$ of $K_r(t)$. So we may suppose that $|U| \leq (t-1) {T \choose t}^r$. Let $B = V(G) \setminus (S \cup U)$. We count the

So we may suppose that $|U| \leq (t-1)\binom{t}{t}$. Let $B = V(G) \setminus (S \cup U)$. We count the number N of edges of \overline{G} between S and B in two different ways. Firstly, the degree in \overline{G} of any vertex v is

$$n-1-d_G(v) \leqslant n-1-\delta(G) \leqslant (n-1)\left(\frac{1}{r}-\varepsilon\right) \leqslant \left(\frac{1}{r}-\varepsilon\right)n,$$

so counting from S we find that

$$N \leq |S| \left(\frac{1}{r} - \varepsilon\right) n = rT \left(\frac{1}{r} - \varepsilon\right) n = (T - \varepsilon rT) n.$$

On the other hand, for each $u \in B$ there is a vertex class S_i of H such that, in G, the vertex u has at most $t-1 \leq t$ neighbours in S_i . Then, in \overline{G} , u has at least T-t neighbours in $S_i \subseteq S$. Hence, counting from B, we see that

$$N \ge |B|(T-t).$$

Since we chose T so that $\varepsilon rT \ge 2t$, it follows that

$$|B| \leqslant \frac{T - \varepsilon rT}{T - t} n \leqslant \frac{T - 2t}{T - t} n = (1 - c)n$$

for some constant c > 0 (depending on r, t, ε). But now in total there are

$$n = |S| + |B| + |U| \le rT + (1 - c)n + (t - 1){\binom{T}{t}}^r = (1 - c)n + O(1)$$

vertices, a contradiction if n is large enough.

Corollary 8.5. Let H be any graph with e(H) > 0. Then

$$ex(n, H) = \left(1 - \frac{1}{\chi(H) - 1} + o(1)\right) \binom{n}{2}$$

as $n \to \infty$, where $\chi(H)$ is the chromatic number of H.

Proof. Let $r = \chi(H) - 1$, so H has chromatic number r + 1. Since H is not r-partite, $T_r(n)$ does not contain any copies of H, so

$$\exp(n, H) \ge t_r(n) = \left(1 - \frac{1}{r} + o(1)\right) \binom{n}{2}.$$

On the other hand, for large enough t (e.g., t = |H|), the graph $K_{r+1}(t)$ does contains a copy of H. Therefore

$$\operatorname{ex}(n,H) \leqslant \operatorname{ex}(n,K_{r+1}(t)) \leqslant \left(1 - \frac{1}{r} + o(1)\right) \binom{n}{2},$$

where the second inequality is from Theorem 8.2.

Corollary 8.5 answers, at some level, the basic extremal question for any graph H. However, there is a weak point: while for $\chi(H) \ge 3$ it tells us asymptotically what value ex(n, H) has, for $\chi(H) = 2$ it only tells us that $ex(n, H) = o(n^2)$, leaving a wide range of possible functions (e.g., roughly $n^2/\log n$, roughly n, roughly $n^{3/2}$ etc). Can we say something more precise in this case?

The Zarankiewicz Problem

Let G be a bipartite graph where the vertex classes have a given size n. How many edges can G have if it does not contain a copy of some given graph H? This makes sense only if H is bipartite, and in particular we consider $H = K_2(t) = K_{t,t}$, i.e., the bipartite analogue of the Turán problem. Formally, let

$$z(n,t) = \max \{ e(G) : G \subseteq K_{n,n} \text{ and } G \text{ contains no copy of } K_{t,t} \}$$

Theorem 8.6. If $n \ge t \ge 2$ then

$$z(n,t) \leqslant (t-1)^{1/t} n^{2-1/t} + (t-1)n^{2-1/t} + (t-1)$$

In particular, as $n \to \infty$ with t fixed we have $z(n,t) = O(n^{2-1/t})$.

Proof. Let G be a maximal bipartite graph with vertex classes X and Y such that |X| = |Y| = n and G contains no $K_{t,t}$. Note that by maximality, $d(v) \ge t - 1$ for every vertex v. (Otherwise, add a new edge incident with v. The degree of v would still be less than t, so the new edge cannot be in any $K_{t,t}$.)

We say that a vertex v covers a set S of vertices if $S \subseteq N(v)$. A vertex $v \in X$ covers exactly $\binom{d(v)}{t}$ t-element subsets of Y. On the other hand, a t-element subset of Y is covered by at most t-1 vertices in X; otherwise we have a $K_{t,t}$. Hence,

$$\sum_{v \in X} \binom{d(v)}{t} \leqslant (t-1)\binom{n}{t}.$$

From here on it is just calculation. Firstly, the polynomial $x(x-1)\cdots(x-t+1)$ is convex on $[t-1,\infty)$, so $\binom{x}{t} = x(x-1)\cdots(x-t+1)/t!$ is convex as a function of the real variable x when $x \ge t-1$. Let $d = \frac{1}{n} \sum_{v \in X} d(v)$ be the average degree in X. Then, by Jensen's inequality,

$$\binom{d}{t} \leqslant \frac{1}{n} \sum_{v \in X} \binom{d(v)}{t} \leqslant \frac{t-1}{n} \binom{n}{t}.$$
(9)

Hence

$$\frac{t-1}{n} \ge {\binom{d}{t}}/{\binom{n}{t}} = \frac{d(d-1)\cdots(d-t+1)}{n(n-1)\cdots(n-t+1)} \ge {\left(\frac{d-t+1}{n}\right)^t}.$$

Rearranging gives

$$d - t + 1 \leq (t - 1)^{1/t} n^{1 - 1/t}$$

and so

$$e(G) = dn \leqslant (t-1)^{1/t} n^{2-1/t} + (t-1)n.$$

Remark. The same method of proof works to show that $ex(n, K_{t,t}) = O(n^{2-1/t})$; we count the number of copies of $K_{1,t}$ in two ways. (As in the proof above, but in the bipartite case we had the extra restriction that the special vertex of $K_{1,t}$ should be in X.)

The special case t = 2 is the same but with simpler calculations.

Theorem 8.7. For $n \ge 2$ we have

$$z(n,2) \leq \frac{n}{2}(1+\sqrt{4n-3}) \sim n^{3/2}$$

as $n \to \infty$.

Proof. We have (9) as before. With t = 2 this becomes $d(d-1) \leq n-1$, rearranging and noting as before that e(G) = nd gives the result.

The bounds just given are only upper bounds. Are they close to the truth? In general, this is an open question. The case t = 2 is particularly nice. Here we have equality if and only if G is regular, any two vertices in Y have exactly one common neighbour in X, and vice versa. A structure with these properties is called a *finite projective plane*: think of the vertices in X as points, those in Y as lines, and edges of G as representing incidence.

It turns out that, except for some degenerate cases, for equality we must have $n = q^2 + q + 1$, each point on q + 1 lines and each line having q + 1 points. Is this possible? For q any prime power the answer is yes: take the projective plane over a finite field with q elements. (This is enough to show that in fact $z(n, 2) \sim n^{3/2}$). However, for n not of this form, even the value of z(n, 2) is unknown in general.