

1 Baby Brooks

Notational note. I use \mathcal{G} to denote the Greedy algorithm.

We wish to prove that if $G = (V, E)$ is connected and non-regular, then $\chi(G) \leq \Delta$.

Let $x_0 \in V$ be s.t. $d(x_0) = \delta$. Since G is connected, running BFS from x_0 adds all vertices to the BFS tree. Let O^{-1} be the ordering of the vertices s.t. z is the i th vertex if it was the i th one to be added by BFS. Trivially, x_0 is the first vertex in O^{-1} . Let O be the reverse order, with x_0 last. We will prove \mathcal{G} colors G with at most Δ colors if it uses the ordering O .

Observe that, in the BFS run, every $x \neq x_0$ is inserted by a neighbor that was already in the tree. In other words, in the O^{-1} order, every vertex has a neighbor that precedes him in the order. Consequently, in O , every $x \neq x_0$ has a neighbor that succeeds him in the order.

It follows that the worst case scenario for the coloring of $x \neq x_0$ is that it has $d(x) - 1$ preceding neighbors. $\therefore \mathcal{G}$ eliminates at most $d(x) - 1 \leq \Delta - 1$ colors. Then x can be colored with a color in $\{1, \dots, \Delta\}$.

When \mathcal{G} reaches x_0 it eliminates at most $d(x_0) = \delta$ colors. Since G is non-regular $\delta < \Delta$. \therefore There is at least one color for x_0 in $\{1, 2, \dots, \Delta\}$.

2 Max flow, min cut

Let f a flow over a network \mathcal{N} . We want to prove two things: (1) $v(f) \leq \text{Cap}(S)$ for any cut S and (2) f is maximal iff there is a cut S s.t. $v(f) = \text{Cap}(S)$.

(1) We know $v(f) = f(S, \bar{S}) - f(\bar{S}, S)$. Since $f(A, B)$ is a sum over f and $0 \leq f(\vec{ab}) \leq c(\vec{ab})$ for any $\vec{ab} \in E$,

$$v(f) = f(S, \bar{S}) - f(\bar{S}, S) \leq f(S, \bar{S})$$

The same logic implies $f(S, \bar{S}) \leq c(S, \bar{S}) = \text{Cap}(S)$. Then $v(f) \leq f(S, \bar{S}) \leq \text{Cap}(S)$. ■

(2: \Leftarrow) Assume there is a cut S s.t. $v(f) = \text{Cap}(S)$. Let g an arbitrary flow. Then $v(g) \leq \text{Cap}(S) = v(f)$. Then f is maximal. Furthermore, it is trivial by definition of Cap that $\text{Cap}(T) \geq v(f)$ for any cut T . Then $\text{Cap}(T) \geq \text{Cap}(S) \Rightarrow S$ is minimal.

(2 : \Rightarrow) Assume f is maximal. Let

$$S = \{s\} \cup \{x \in V : \exists f\text{-camino aumentante entre } s \text{ y } x\}$$

S is a cut because, if $t \in S$, there is an augmenting path $s \dots t$ and the flow can be augmented, which contradicts that f is maximal.

Recall that $v(f) = f(S, \bar{S}) - f(\bar{S}, S)$. The first term in the difference is

$$f(S, \bar{S}) = \sum_{x \in S, z \notin S, \vec{xz} \in E} f(\vec{xz})$$

Let $\vec{xz} \in E$ a side in the range of the sum above. Then there is an augmenting path $s \dots x$ and there is no augmenting path $s \dots z$. But $\vec{xz} \in E$ and $s \dots x \dots z$ is a path. Since it cannot be an augmenting path, we must have $f(\vec{xz}) = c(\vec{xz})$. Then $f(\vec{xz}) = c(\vec{xz})$ for all $x \in S, z \notin S, \vec{xz} \in E$. Therefore

$$f(S, \bar{S}) = \sum_{\dots} f(\vec{xz}) = \sum_{\dots} c(\vec{xz}) = \text{Cap}(S)$$

Now consider the second term in the difference:

$$f(\bar{S}, S) = \sum_{w \notin S, x \in S, \vec{wx} \in E} f(\vec{wx})$$

Let \overrightarrow{wx} an arbitrary side in the sum above. Again, there must be an augmenting path $s \dots x$, but not one $s \dots w$. But \overrightarrow{wx} is a side, and then $s \dots \overleftarrow{xw}$ is not augmenting only if $f(\overleftarrow{xw}) = 0$. This means $f(\overrightarrow{wx}) = 0$ for all \overrightarrow{wx} in the range of the sum above.

$$\therefore v(f) = \text{Cap}(S) - 0 = \text{Cap}(S). \blacksquare$$

3 Networks

3.1 Edmond-Karp: Complexity

The complexity of E.K. is determined by the complexity of finding the f_i -augmenting paths, the number of such paths, and the complexity of updating the flow with such paths.

To find f_i -augmenting paths, EK uses BFS with a complexity $O(m)$. Updating the flow has complexity $O(n)$. So the question is what is the number φ of f_i augmenting paths (or runs of BFS) that are used.

The number of such paths is bounded by the number of sides times the number of times a side may become critical. We shall determine how many times a side may become critical.

Let f_1, f_2, \dots be the iterations of EK. Let \overrightarrow{xz} be a side that became critical at iteration k . There are two options: it either saturated being forward or emptied being backward.

Saturated being forward. If \overrightarrow{xz} saturated being forward, then $p_k = s \dots \overrightarrow{xz} \dots t$ is the form of the augmenting path used to determine f_k .

Assume \overrightarrow{xz} becomes critical again at some iteration j , so that $p_j = s \dots \overrightarrow{xz} \dots t$ is an augmenting path. For this to occur there must exist some iteration i , with $k < i < j$, such that $p_i = s \dots \overleftarrow{xz} \dots t$ is an augmenting path. In other words, z must have returned some of the flow to x . Since we are using EK, the augmenting paths are of minimal length, and the distance between two vertices never decreases. This means

$$\begin{aligned} d_j(t) &\geq d_i(x) + b_i(x) \\ &= d_i(z) + 1 + b_i(x) \\ &\geq d_k(z) + 1 + d_k(x) \\ &= d_k(x) + 1 + 1 + b_k(x) \\ &= d_k(t) + 2 \end{aligned}$$

Emptied being backward. Now, assume \overrightarrow{xz} is a side that saturated being backward on the k th iteration; this is, that $p_k = s \dots \overleftarrow{xz} \dots t$ was the path found. Assume it empties again at some iteration j , so that $p_j = s \dots \overleftarrow{xz} \dots t$ is the f_j path. Then there is some i , with $k < i < j$, such that $p_i = s \dots \overrightarrow{xz} \dots t$ is a path; i.e. the side was used forward. Now, all these paths are of minimal length, because we

are using E.K. And in all successive paths, the distance from s to t never decreases. Then,

$$\begin{aligned}
 d_j(t) &\geq d_i(t) \\
 &= d_i(z) + b_i(z) \\
 &= d_i(x) + 1 + b_i(z) \\
 &\geq d_k(x) + 1 + b_k(z) \\
 &= d_k(z) + 1 + 1 + b_k(z) \\
 &= d_k(t) + 2
 \end{aligned}$$

In both cases, the distance from s to t had to increase by at least two units. But the distance from s to t is bounded by n ; this is, it is $O(n)$. Then a side may become critical $O(\frac{n}{2}) = O(n)$ times. Since there are m sides, the number of times an arbitrary side will become critical is $O(mn)$.

\therefore The complexity is $O(mn) [O(n) + O(m)] = O(mn)O(m) = O(m^2n)$.

3.2 Edmond-Karp: Augmenting paths are non-decreasing

Let $A = \{x \in V : d_{k+1}(x) < d_k(x)\}$ and assume $A \neq \emptyset$. Let $x_0 \in A$ be the vertex whose distance $d_{k+1}(x_0)$ from s is minimal; i.e. $d_{k+1}(x_0) \leq d_{k+1}(y) \forall y \in A$. Since $x_0 \in A$, $d_{k+1}(x_0) < d_k(x_0) \leq \infty$. Then there exists a $\mathcal{P}_{k+1} = s \dots x_0$ of minimal length. Let z be the predecessor of x_0 in this path.

By definition of d_f , the length of the path is $d_{k+1}(x_0)$. Because the path is of minimal length to x_0 , it is of minimal length to any predecessor of x_0 in it, including z . Then $d_{k+1}(z) = d_{k+1}(x_0) - 1$. This implies $z \notin A$.

There are two possible cases: either $\overrightarrow{xz} \in E$ or $\overrightarrow{zx} \in E$.

(Case 1): If $\overrightarrow{xz} \in E$, then $d_{k+1}(z) < d_{k+1}(x_0)$. Since $z \notin A$,

$$d_k(z) \leq d_{k+1}(z) < d_{k+1}(x_0) < \infty$$

Since $d_k(z) < \infty$, there is an f_k -augmenting path from s to z . Then, in principle, $s \dots zx$ could be an augmenting path. But if this were the case,

$$d_k(x_0) \leq d_k(z) + 1 \leq d_{k+1}(z) + 1 = d_{k+1}(x_0)$$

which implies $x_0 \notin A$ (\perp). $s \dots zx$ is not f_k -a.p. This can only happen if $f_k(\overrightarrow{zx_0}) = c(\overrightarrow{zx_0})$ (the side is saturated). Since the side is used in a f_{k+1} -a.p. it must be the case that $f_{k+1}(\overrightarrow{zx_0}) < c(\overrightarrow{zx_0})$. This means \overrightarrow{zx} was used backwards in the k th iteration, or rather that $f_k = s \dots \overleftarrow{x_0z} \dots t$.

Since this is Edmond-Karp, augmenting paths are of minimal length. Then

$$\begin{aligned} d_k(z) &= d_k(x_0) + 1 \\ &> d_{k+1}(x_0) + 1 \\ &= d_{k+1}(z) + 2 \\ &\geq d_k(z) + 2 \end{aligned}$$

which is absurd.

1.3.2 If \overrightarrow{xz} is a side then again

$$d_k(z) \leq d_{k+1}(z) < d_{k+1}(x_0) < \infty$$

Then there is an f_k -a.p. $s \dots z$ and, at least in principle, $s \dots \overrightarrow{xz}$ could also be augmenting. But if this were the case, we would have

$$d_k(x_0) = d_k(z) + 1 \leq d_{k+1}(z) + 1 = d_{k+1}(x_0)$$

which would imply $x_0 \notin A$ (\perp). Then $s \dots \overrightarrow{zx_0}$ is not augmenting, which means the side $\overrightarrow{x_0z}$ cannot be used backwards. This can only be true if $f_k(\overrightarrow{x_0z}) = 0$. But since the side is used backwards in the f_{k+1} -augmenting path, it must be used forward in this iteration. Which means $s \dots \overrightarrow{x_0z} \dots t$ is augmenting. But then

$$\begin{aligned} d_k(z) &= d_k(x_0) + 1 \\ &> d_{k+1}(x_0) + 1 \\ &= d_{k+1}(z) + 2 \\ &\geq d_k(z) + 2 \end{aligned}$$

which is absurd.

Conclusion. In both cases a contradiction arises. The contradiction comes from assuming $A \neq \emptyset$. Then $A = \emptyset$. ■

4 Dinitz: Complexity

We will prove the complexity of Dinitz is $O(n^2m)$.

Recall that Dinitz builds successive auxiliary networks with BFS, uses DFS to find blocking paths in them, and uses these paths to update the flow. Since the level of t augments in each successive network, there are at most $O(n)$ auxiliary networks. So the complexity of Dinitz is

$$O(n) [\text{Comp. of building A.N.} + \text{Comp. of finding blocking flow in A.N.}]$$

The complexity of building the A.N. is the complexity of BFS, which is $O(m)$. The process of finding the blocking path varies in the Western and the original algorithms.

Original version. The original Dinitz algorithm enforced the following invariant: In any given auxiliary network, all vertices have edges connecting to the next level. This invariant implies that the DFS run always reaches t without need to backtrack. In consequence, the complexity of the DFS run is $O(n)$. The complexity of updating the flow is $O(n)$, and each time this happens at least one side is saturated and removed. Then there are $O(m)$ paths. \therefore The complexity of finding the paths and updating the flow is $O(n \times m)$.

However, there is some extra cost associated to preserving this invariant. Each time a path is found, saturated sides are removed with a pruning operation. The pruning operation goes from the highest to the lowest levels in the network, checks for vertices whose exiting edge is saturated, and deletes them. This means that, for each vertex, a process of $O(1)$ complexity checks if it is saturated. This happens each time a path is found, so $O(n \times m)$ times. Once these sides have been detected, they and their neighbors are removed, which has complexity $O(d(x))$. Since this happens at worst for all vertices, the deleting operation is $O(\sum_{x \in V} d(x)) = O(m)$. The total cost of enforcing the invariant is then $O(n \times m) + O(m)$, which means the complexity of finding the blocking paths and updating the flow is $O(n \times m) + O(n \times m) + O(m) = O(n \times m)$. Then the complexity of Dinitz is $O(n^2m)$.

Western version. The Western version finds blocking paths and updates the flow inside a while loop with three clauses. The first clause is running BFS to advance; i.e. setting $x = s$ and iteratively making x be the first neighbor of x until t is reached. This clause executes as long as there is a neighbor and we call it A.

The second clause considers the case where x , the vertex we are traversing, has no neighbors in the next level. Then the algorithm backtracks to its predecessor,

removes the side leading from it to x , and attempts A once more. We call this part R .

The last clause considers the case where t is reached. Here, the flow is augmented using the path found and all saturated sides are removed.

Thus, finding the blocking flow and updating the path can be modeled as a word $A \dots IA \dots RA \dots RA \dots I$; i.e. as a succession of words of the form $A \dots X$ with $X \in \{R, I\}$.

The process R has complexity $O(1)$ because it simply involves going backwards and deleting a side. The process A has complexity $O(1)$ because it simply involves moving forward, but at most $O(n)$ successions of A may occur. This means $O(A \dots A) = O(n)$. The process I has complexity $O(n)$ because the flow is updated across at most all vertices. Then $O(A \dots I) = O(n) + O(n) = O(n)$ and $O(A \dots R) = O(n) + O(1) = O(n)$. The question is how many words of the form $A \dots X$ exist. At worst, all paths are traversed in the process of finding t , so there are $O(m)$ words of this form. Then the complexity of the run is

$$O(m) [O(n) + O(n)] = O(nm)$$

Then the complexity of Dinitz is $O(n^2m)$.

4.1 Wave: Complexity

We know the distance between s and t in auxiliary networks is increasing. The distance is bounded by n . \therefore There are $O(n)$ auxiliary networks. Now, each auxiliary network is first constructed and then used to find a blocking flow. Then the complexity of Wave is

$$O(n) [F + B]$$

where F is the complexity of finding a blocking flow in an auxiliary network and B the complexity of building an auxiliary network. To build the auxiliary network we use BFS. $\therefore B = O(m)$. Let us examine F .

To find blocking flows we attempt to balance in forward and backward waves. When going forward, let V be the steps where a side is saturated, \bar{V} those where a side is not saturated. When going backward, let S be the steps where a side is emptied, \bar{S} those where a side is not emptied.

Complexity of V . Assume \vec{xz} is a side and is saturated in a forward wave. To saturate again it must empty at least a little bit before. If it empties, then z was blocked and returned the flow to x ; and since z is blocked, \vec{xz} will never again be used. Then each side can saturate at most once. \therefore the complexity of V is $O(m)$.

Complexity of S . Assume \vec{zx} is a side and it empties in a backward wave. Since x returned flow, it is blocked and z will not send flow to x again. Then \vec{zx} cannot be emptied ever again. \therefore The complexity of S is $O(m)$.

Complexity of \bar{V} . When a vertex sends flow to its neighbors, it saturates all sides except perhaps one. Then, for any given vertex, each forward wave will send partial flow through at most one side. \therefore The complexity of \bar{V} is $O(n) \times \varphi$, with φ the number of forward waves.

Complexity of \bar{S} . When a vertex returns flow to its predecessors, it empties all sides except perhaps one. Then, for any given vertex in a backward wave, at most one side is partially emptied. \therefore The complexity of \bar{S} is $O(n) = \psi$ where ψ is the number of backward waves.

Now, it is trivially to see $\varphi = \psi$ and there are at most n forward waves. $\therefore O(\varphi) = n$.

\therefore The complexity of \bar{S}, \bar{V} are both $O(n) \times O(n) = O(n^2)$.

$\therefore F = O(n^2) + O(n^2) + O(m) + O(m) = O(n^2) + O(m)$.

But $O(n^2) + O(m) = O(n^2)$, because $m \leq \binom{n}{2} = O(n^2)$.

$\therefore F = O(n^2)$.

\therefore The complexity of Dinitz is $O(n) [O(m) + O(n^2)] = O(n)O(n^2) = O(n^3)$

5 Codes

5.1 Hamming bound

Let $C \subseteq \{0, 1\}^n$. We want to prove

$$|C| = \frac{2^n}{\sum_{i=0}^t \binom{n}{i}}$$

Let $A = \bigcup_{v \in C} D_t(v)$. Recall that $D_t(v)$ is the set of words in $\{0, 1\}^n$ that are at a Hamming distance of t or less from v .

Let $S_v(r) = \{w \in \{0, 1\}^n : d_H(v, w) = r\}$. Then it follows by definition that $D_t(v) = \bigcup_{r=0}^t S_v(r)$. Of course, this union is disjoint. It follows that

$$A = \bigcup_{v \in C} \bigcup_{r=0}^t S_v(r)$$

and

$$|A| = |C| \times \sum_{r=0}^t |S_v(r)|$$

So now we must only determine $|S_v(r)|$. But this is easy to do if we consider that any $w \in S_v(r)$ differs from v by exactly r bits, and is fully determined by this difference. In other words, there is a bijection between any $w \in S_v(r)$ and the set of the r bits out of all n possible bits that make up the difference between w and v . This readily entails that $|S_v(r)| = \binom{n}{r}$. This readily gives

$$\begin{aligned} |A| &= |C| \times \sum_{r=0}^t \binom{n}{r} \\ \Rightarrow |C| &= \frac{|A|}{\sum_{r=0}^t \binom{n}{r}} \end{aligned}$$

We do not know the cardinality of A , but since $A \subseteq \{0, 1\}^n$ we know $|A| \leq 2^n$. Then

$$|C| \leq \frac{2^n}{\sum_{r=0}^t \binom{n}{r}}$$

5.2 $\delta(C) = \min \{j : \exists S \subseteq H_{*n} : |S| = j \wedge S \text{ is LD}\}$

Notation. I use H_{*n} to denote the set with the n columns of H . I use $H^{(i)}$ to denote the i th column of H .

Let $s = \min \{j : \exists S \subseteq H_{*n} : |S| = j \wedge S \text{ is LD}\}$. This implies there are s columns $H^{(j_1)}, \dots, H^{(j_s)}$ s.t. $\sum x_i H^{(j_i)} = 0$ for x_1, \dots, x_s not all null.

(1) Let $w := \sum x_i e_{j_i}$ where e_k is the vector with all zeroes except at the k th coordinate. Since not all x_i are zeroes, $w \neq 0$. Now,

$$\begin{aligned} Hw^t &= H (x_1 e_{j_1} + \dots + x_s e_{j_s})^t \\ &= x_1 H e_{j_1}^t + \dots + x_s H e_{j_s}^t \\ &= \sum x_i H^{(j_i)} && \left\{ \text{Because } H e_j^t = H^{(j)} \right\} \\ &= 0 \end{aligned}$$

Then $w \in \text{Nu}(H) = C$. But $|w| \leq s$ and $w \neq 0$. We know $\delta = \min \{|x| : x \in C, x \neq 0\}$.
 $\therefore \delta \leq |w| \leq s$.

(2) Let $v \in C$ s.t. $\delta = |v|$. Then there are i_1, \dots, i_δ s.t. $v = e_{i_1} + \dots + e_{i_\delta}$. Since $v \in C, Hv^t = 0$, which using the same logic as before gives $\sum H^{(i_j)} = Hv^t = 0$.

This implies $\{H^{(i_1)}, \dots, H^{(i_\delta)}\}$ is LD.

$\therefore s \leq \delta$.

(3) Points (1) and (3) imply $s = \delta$.

5.3 Three statements around a generating polynomial

Let C a code of length n and dimension k with generating polynomial $g(x)$. We will prove:

1. $C = \{p(x) : gr(p) < n \wedge g(x) \mid p(x)\} := C_1$
2. $C = \{v(x) \odot g(x) : v(x) \in F[x]\} := C_2$
3. $gr(g) = n - k$
4. $g(x) \mid (1 + x^n)$

(1 and 2) : We shall prove $C_1 \subseteq C_2 \subseteq C \subseteq C_1$.

Let $p(x) \in C_1$. Then there is some $q(x)$ s.t. $p(x) = g(x)q(x)$ and $gr(g(x)q(x)) < n$.

$$\therefore g(x)q(x) = g(x) \odot q(x) \in C_2.$$

$$\therefore C_1 \subseteq C_2.$$

Now let $p(x) = v(x) \odot g(x) \in C_2$ with $v(x)$ an arbitrary polynomial. Then

$$\begin{aligned} p(x) &= \left(v_0 + v_1x + \dots + v_{gr(v)}x^{gr(v)} \right) \odot g(x) \\ &= v_0 \odot g(x) + v_1 (x \odot g(x)) + v_2 (x^2 \odot g(x)) + \dots + v_{gr(v)} \left(x^{gr(v)} \odot g(x) \right) \\ &= v_0 g(x) + v_1 Rot(g(x)) + v_2 Rot^2(g(x)) + \dots + v_{gr(v)} Rot^{gr(v)}(g(x)) \end{aligned}$$

All rotations of $g(x)$ belong to C .

$$\therefore p(x) \in C.$$

$$\therefore C_2 \subseteq C.$$

Now let $p(x) \in C$. By definition, $gr(p) < n$, which implies $p(x) = p(x) \bmod (1 + x^n)$. We know

$$p(x) = g(x)q(x) + r(x)$$

for some $q(x), r(x)$ s.t. $gr(r) < gr(g)$. Then

$$\begin{aligned}
p(x) &= (g(x)q(x) + r(x)) \mod (1 + x^n) \\
&= g(x) \odot q(x) + (r(x) \mod (1 + x^n)) \\
&= g(x) \odot q(x) + r(x) \quad \{\text{Since } gr(r) < gr(g) < n\}
\end{aligned}$$

$$\therefore r(x) = p(x) + g(x) \odot q(x).$$

We know $p \in C$ and $g(x) \odot q(x) \in C^2 \subseteq C$.

$$\therefore r(x) \in C.$$

But since g is generating polynomial, it is the unique polynomial with the least non-null degree in C .

$$\therefore gr(r) < gr(g) \Rightarrow r(x) = 0.$$

$$\therefore g(x) \mid p(x) \text{ and then } p(x) \in C_1.$$

$$\therefore C \subseteq C_1$$

(3) Let $p(x) \in C$. Then there is $q(x)$ s.t. $p(x) = g(x)q(x)$ with $n > gr(p) = gr(g) + gr(q)$. This readily implies $gr(q) < n - gr(g) < n$. Then $g(x)q(x) \in C$.

This entails there is a bijection between C and the set of polynomials of degree $n - gr(g)$. Then

$$\begin{aligned}
|C| &= |\{p(x) : gr(p) < n - gr(g)\}| \\
\iff 2^k &= 2^{n-gr(g)} \\
\iff k &= n - gr(g) \\
\iff gr(g) &= n - k \blacksquare
\end{aligned}$$

(4) Divide $(1 + x^n)$ by $g(x)$ to obtain

$$1 + x^n = g(x)q(x) + r(x)$$

with $gr(r) < gr(g)$. Taking the modulus,

$$\begin{aligned}
0 &= (1 + x^n) \mod (1 + x^n) \\
&= (g(x)q(x) + r(x)) \mod (1 + x^n) \\
&= (g(x) \odot q(x)) + (r(x) \mod 1 + x^n) \\
g(x) \odot q(x) &= r(x)
\end{aligned}$$

because $gr(r) < gr(g) < n$.

$$\therefore r(x) = g(x) \odot q(x) \in C.$$

But $gr(r) < gr(g)$. $\therefore r(x) = 0$ and $g(x) \mid (1 + x^n)$.

6 Matchings

6.1 Konig

We want to prove that any bipartite and regular graph $G = (V, E)$ has a perfect matching. Let X, Y be the two parts of G . For any $W \subseteq V$ let $E_W := \{wu \in E : w \in W\}$.

(1) Let $S \subseteq X$ and $l \in E_S$. It follows that

$$\exists x \in S, y \in Y : l = xy = yx$$

$\therefore y \in \Gamma(x)$. And since $x \in S$ we have $y \in \Gamma(S)$ and $l \in E_{\Gamma(S)}$.

$\therefore E_S \subseteq E_{\Gamma(S)}$ and $|E_S| \leq |E_{\Gamma(S)}|$.

(2) Let us calculate $|E_W|$ when $W \subseteq X$.

Observe that $E_W = \bigcup_{w \in W} \{wv : v \in \Gamma(w)\}$. Furthermore, the union is disjoint, because $wv \in E_W \Rightarrow w \in X \Rightarrow v \in Y$. Then

$$|E_W| = \sum_{w \in W} |\Gamma(w)| = \sum_{w \in W} d(w)$$

Since G is regular, $d(w) = \delta = \Delta$.

$\therefore |E_W| = \Delta|W|$

(3) Using what we established in (1), it follows from (2) that

$$|S|\Delta \leq |\Gamma(S)|\Delta \Rightarrow |S| \leq |\Gamma(S)|$$

This holds for any $S \subseteq X$. Then Hall's theorem implies there is a complete matching from X to Y . To prove it is perfect, we must prove $|X| = |Y|$.

But since X, Y are the two parts of G , $E = E_X = E_Y$. Then $|E_X| = |E_Y|$, which implies $|X|\Delta = |Y|\delta \Rightarrow |X| = |Y|$.

Alternatively, since there is a complete matching from X to Y , $|X| \leq |Y|$. But the choice of X over Y was arbitrary, and then the same holds for Y . Then $|X| = |Y|$.

In both cases the matching is perfect.

6.2 Hall

Let $G = (V, E)$ a bipartite graph with parts X and Y , and let $Z \in \{X, Y\}$. We want to prove that there is a complete matching from X to Y iff $\forall S \subseteq Z : |S| \leq |\Gamma(S)|$.

(\Rightarrow) The proof is trivial, because if such matching exists, it induces an injective function $f : X \rightarrow Y$ s.t. $f(x) \in \Gamma(x)$. Since it is an injection, $|f(S)| = |S|$ for any S . Then $f(S) \subseteq \Gamma(S) \Rightarrow |S| \leq |\Gamma(S)|$.

(\Leftarrow) Assume the Hall condition $|S| \leq |\Gamma(S)|$ holds. Assume that, after running the algorithm to find a maximal matching, an incomplete matching is found. We will build $S \subseteq X$ that violates our assumption (we could use $S \subseteq Y$ without loss of generality).

(1) Let S_0 be the set of rows unmatched and $T_1 = \Gamma(S_0)$. Observe that, by assumption, $S_0 \neq \emptyset$, and all columns in T_1 have a match that is not in S_0 . Let S_1 the set of rows matching columns of T_1 and $T_2 = \Gamma(S_1) - T_1$. Generally,

$$S_i = \text{Rows matching with } T_i$$

$$T_{i+1} = \Gamma(S_i) - \bigcup_{j=0}^{j=i} T_j$$

The algorithm stops only when it is revising a row and this row has no available neighbors; this is, it only stops passing from a S_i to a T_{i+1} when $T_{i+1} = \emptyset$. Furthermore, since each column only labels a single row (that of its match), and T_i "creates" S_i , we have $|S_j| = |T_j|$.

Define $S = \bigcup S_i$, $T = \bigcup T_i$, and note that all the S_i are disjoint and all the T_i are disjoint. Then

$$\begin{aligned} |S| &= \sum |S_i| \\ &= |S_0| + \sum |T_i| \\ &= |S_0| + |T| \end{aligned}$$

$\therefore |S| > |T|$ (since $S_0 \neq \emptyset$).

We must only prove now that $T = \Gamma(S)$.

(1) T are the labeled columns, and each column is labeled by a row in S . Each row only labels its neighbors. This implies $T \subseteq \Gamma(S)$.

(2) Assume $y \in \Gamma(S)$ and $y \notin T$. Then y was not labeled. But since $y \in \Gamma(S)$ there is an $x \in S$ s.t. $y \in \Gamma(x)$. Then each time the algorithm passes through x it

should label y , which contradicts the fact that y is not labeled. Then $y \in T$. Then $\Gamma(S) \subseteq T$.

$\therefore \Gamma(S) = T$ and $|S| > |\Gamma(S)|$. But this contradicts the hypothesis that the Hall condition holds. The contradiction comes from assuming there wasn't a complete matching. \therefore There is a complete matching. ■

7 P-NP

7.1 2-Color is polynomial

We must give a polynomial algorithm that decides whether an arbitrary $G = (V, E)$ is $\chi(G) = 2$.

```

n_colored := 0
while j < n do
    x := v ∈ V, v not colored
    C(x) := 1
    n_colored := n_colored + 1
    Q = queue with only x
    while Q ≠ ∅ do
        p := pop!(Q)
        for w ∈  $\Gamma(p)$  do
            if w is not colored do
                push!(w, Q)
                C(w) = 3 − C(p)
                n_colored = n_colored + 1
    for {v, w} ∈ E do
        if C(v) = C(w) return False else return True

```

Complexity. The inner **while** traverses the vertices in the connected component of *x*, which means the outer **while** is executed once per connected component. Furthermore, inside the inner **while** we loop through $\Gamma(p)$. Then the complexity of the inner **while** is

$$O \left[\sum_{p \in C(x)} d(p) \right] = O [2 \times \text{edges in } C(x)] = O(\text{\#edges in } C(x))$$

where $C(x)$ is the connected component of *x*. The **for** loop is $O(m)$ of course. \therefore the algorithm is polynomial.

Correctness. It is trivial to note that if the algorithm returns *True* then G is 2-color. Now, assume the algorithm returned *False*. Then there is some $\{v, w\} \in E$ s.t. $C(v) = C(w)$. These belong to the same connected component.

Let x be the root of this connected component from which the queue was built. Without loss of generality, assume v entered the queue first. Then, when v became the first element in the queue, w must have already been in the queue. Otherwise, because they are neighbors, v would have added w with the color $3 - C(v)$, which contradicts the hypothesis. Then there is a chain of inclusion in the queue

$$x = v_r \rightarrow v_{r-1} \rightarrow \dots \rightarrow v_1 \rightarrow v_0 = v$$

$$x = w_t \rightarrow w_{t-1} \rightarrow \dots \rightarrow w_1 \rightarrow w_0 = w$$

Since v came first, we must have $r \leq t$, but since w is the queue when v is its first element, $t \leq r + 1$. Now, since the color depends of the parity of r and t , $C(v) = C(w) \Rightarrow t \equiv r \pmod{2}$. Then $t = r$.

Let k the first index s.t. $v_k = w_k$; then we have a path $v \rightarrow v_1 \rightarrow v_k = w_k \rightarrow w_{k-1} \rightarrow w_{k-2} \dots w$ with $wk + 1$ vertices. But since v, w are an edge, the graph contains C_{2k+1} . $\therefore \chi(G) \leq 3$.

7.2 3SAT es NP-Completo

Let $B = B_1 \wedge \dots \wedge B_m$ an instance of SAT with variables x_1, \dots, x_n . We build an instance of 3-SAT by transforming each B_i into an E_i as follows:

Complete.

$$E_i = (e_1 \vee e_2 \vee y_1) \wedge (\overline{y_1} \vee y_2 \vee e_3) \wedge (\overline{y_2} \vee y_3 \vee e_4) \vee \dots \vee (\overline{y_{k-3}} \vee e_{k-1} \vee e_k)$$

We want to prove

$$\exists \vec{b} : B(\vec{b}) = 1 \iff \exists \vec{a} : \tilde{B}(\vec{b}, \vec{a}) = 1$$

(\Leftarrow) Assume $B(\vec{b}) = 0$. Then $D_i(\vec{b}) = 0$ for some i . Let e_1, \dots, e_k be the literals in D_i .

If $k = 3$ a contradiction ensues trivially. If $k = 2$, then $D_i = e_1 \vee e_2$ and then $E_i = (e_1 \vee e_2 \vee y_1) \wedge (e_1 \vee e_2 \vee \overline{y_1})$. Since $D_i = 0$, $e_1 \vee e_2 = 0$ and therefore $e_1 = e_2 = 0$. From this follows $E_i = y_1 \wedge \overline{y_1} = 1$. (\perp)

If $k = 1$ then $e_1 = 0$ and therefore $E_i = (y_1 \vee y_2) \wedge (y_1 \vee \overline{y_2}) \wedge (\overline{y_1} \vee y_2) \wedge (\overline{y_1} \vee \overline{y_2}) = 0$. But by assumption $E_i = 1$ (\perp).

If $k \geq 4$ we must observe that, since $D_i(\vec{b}) = 0$, we have $e_1 = e_2 = \dots = e_k = 0$. Then this literals are neutral elements in the disjunctions and can be ignored. Since $E_i(\vec{b}, \vec{a}) = 1$, its first term is true; in other words, $e_1 \vee e_2 \vee y_1 = 1 \Rightarrow y_1 = 1$. In all the following cases (except the last), $E_i = \overline{y_{i-1}} \vee y_i$ must be true; this is, $y_i \Rightarrow y_{i+1}$ is true. But the last term is $\overline{y_{k-3}}$, which cannot be true because y_1 and $y_1 \Rightarrow y_2 \Rightarrow \dots \Rightarrow y_{k-3}$. (\perp)

(\Rightarrow) Assume $B(\vec{b}) = 1$. For $k = 1, k = 2$, define $y_i = 0$ for all i . $\therefore D_i(\vec{b}) = 1 \Rightarrow E_i(\vec{b}, \vec{a}) = 1$. For $k = 3$ the result is trivial. Let us consider the case $k \geq 4$.

Since $D_i(\vec{b}) = 1$ is a true disjunction, at least one e_r is true under \vec{b} . Define the following assignment:

$$\begin{aligned} y_1 &= y_2 = \dots = y_{r-2} = 1 \\ y_i &= 0 \text{ para todos los demás } i \end{aligned}$$

Then

$$\begin{array}{ll}
E(\vec{b}, \vec{\alpha}) = (e_1 \vee e_2 \vee y_1) & \{\text{True because } y_1 = 1\} \\
\wedge (\overline{y_1} \vee y_2 \vee e_3) & \{\text{True because } y_2 = 1\} \\
\vdots & \\
\wedge (\overline{y_{r-3}} \vee y_{r-2} \vee e_{r-1}) & \{\text{True because } y_{r-2} = 1\} \\
\wedge (\overline{y_{r-2}} \vee y_{r-1} \vee e_r) & \{\text{True because } e_r = 1\} \\
\wedge (\overline{y_{r-1}} \vee y_r \vee e_{r+1}) & \{\text{True because } y_{r-1} = 0\} \\
\vdots & \\
\wedge (\overline{y_{k-3}} \vee e_{k-1} \vee e_k) & \{\text{True because } y_{k-3} = 0\}
\end{array}$$

\therefore Our assignment makes \tilde{B} true.

7.3 3-Color es NP-Completo

Sabemos que $3\text{-Color} \in NP$. La idea es ver que $3\text{-SAT} \leq_p 3\text{-COLOR}$. Debemos crear un grafo $G = (V, E \cup F)$ tal que B es satisfactible si y solo si $\chi(G) \leq 3$.

Let $B = D_1 \wedge \dots \wedge B_m$ with variables x_1, \dots, x_n and each $B_i = (l_{i1} \vee l_{i2} \vee l_{i3})$. Let $\mathcal{G} = (V, E)$ a graph formed as follows:

- A nucleus t from which n triangles with sides $\{t, v_1, w_1\}, \dots, \{t, v_n, w_n\}$ form.
- m claws, each with a triangle $\{b_{i1}, b_{i2}, b_{i3}\}$, and such that from each b_{ij} sprouts a tip u_{ij} .
- A source s connected to t and to every tip u_{ij} .

Now, let $\psi : \{l_{11}, \dots, l_{m3}\} \rightarrow V$ a function that maps a literal to v_k if the literal is x_k , and to w_k if the literal is $\overline{x_k}$. In other words,

$$\psi(l_{ij}) := \begin{cases} v_k & l_{ij} = x_k \\ w_k & l_{ij} = \overline{x_k} \end{cases}$$

We will use ψ to create our graph $G = (V, E \cup F)$ by letting

$$F = \{u_{ij}\psi(l_{ij}) : 1 \leq i \leq m, 1 \leq j \leq 3\}$$

In other words, we connect each u_{ij} to either v_k or w_k , depending on whether $l_{ij} = x_k$ or $\overline{x_k}$. Now that we have defined G , we must only prove that B is satisfiable iff G is 3-colorable.

Proof of (\Leftarrow): Assume $\chi(G) \leq 3$. Since G has triangles, $\chi(G) = 3$. Let $\vec{b}_k = [c(v_k) = c(s)]$.

We must prove $B_i(\vec{b}) = 1$ for all $1 \leq i \leq m$. Take an arbitrary B_i .

- (1) The triangle $\{b_{i1}, b_{i2}, b_{i3}\}$ must have $c(b_{ij}) = c(t)$ for some j . Then, since $\{b_{ij}, u_{ij}\}$ is a side, $c(u_{ij}) \neq c(t)$. And since $\{u_{ij}, s\}$ is a side, $c(u_{ij}) \neq c(s)$. $\therefore u_{ij}$ was colored with the third color.
- (2) Since $\{u_{ij}, \psi(l_{ij})\}$ is a side, and $\{\psi(l_{ij}), t\}$ is a side, $c(\psi(l_{ij})) = c(s)$.

We have established that $c(\psi(l_{ij})) = c(s)$. By definition, we have two cases.

(1) If $\psi(l_{ij}) = v_k$, $l_{ij} = x_k$ and $c(v_k) = c(s)$, which means $\vec{b}_k = 1$.

Then $l_{ij}(\vec{b}) = 1$. Then $D_i(\vec{b}) = 1$.

(2) If $\psi(l_{ij}) = w_k$, then $l_{ij} = \bar{x}_k$. Since $\{v_k, w_k\}$ is a side, these vertices have different colors. Then $c(v_k) \neq c(s)$. Then $\vec{b}_k = 0$.

Then $l_{ij}(\vec{b}) = \bar{x}_k(\vec{b}) = 1$. Then $D_i(\vec{b}) = 1$.

In both cases, $D_i(\vec{b}) = 1$. This holds for any $i = 1, 2, \dots, m$. Then $B(\vec{b}) = 1$.

■

Proof of (\Rightarrow) : Assume there is some \vec{b} s.t. $B(\vec{b}) = 1$. Let $C = \{C_s, C_t, C\}$ a set of three colors, and let $c(s) = C_s$, $c(t) = C_t$, and

$$c(v_k) = \begin{cases} C_s & b_k = 1 \\ C & b_k = 0 \end{cases}, \quad c(w_k) = \begin{cases} C & b_k = 1 \\ C_s & b_k = 0 \end{cases}$$

Clearly, we are ensuring $c(v_k) \neq c(w_k) \neq c(t)$, so the triangles $\{t, v_i, w_i\}$ are all properly colored. And of course, $\{t, s\}$ is also properly colored. All we have to do is look at the claws.

First, since $\exists \vec{b} : B(\vec{b}) = 1$, then $D_i(\vec{b}) = 1$ for all i . This means, in an arbitrary D_i , there is at least one l_{ij} s.t. $l_{ij}(\vec{b}) = 1$. Let's color the tips of the claw as follows:

$$c(u_{ir}) = \begin{cases} C & r = j \\ C_t & r \neq j \end{cases}$$

Clearly, $\{u_{ir}, s\}$ is properly colored. What about $\{u_{ir}, \psi(l_{ir})\}$? There are two cases:

(Case $r \neq j$) : In this case, $c(u_{ir}) = C_t$, and since $\psi(l_{ir})$ is either a v or a w , $c(\psi(l_{ir})) \neq C_t$.

(Case $r = j$) : Here, $c(u_{ir}) = C$. If $l_{ij} = x_k$, $\psi(l_{ij}) = v_k$. By definition of l_{ij} , $l_{ij}(\vec{b}) = 1$. Then $c(\psi(l_{ij})) = c(v_k) = C_s$.

So, in both cases we are properly coloring $\{u_{ir}, \psi(l_{ir})\}$.

Now that we have colored the tips u_{ir} , we only have to color the triangles in the claw, $\{b_{i1}, b_{i2}, b_{i3}\}$. Let $c(b_{ij}) = C_t$ and the other two be colored in any of the possible ways. Clearly, the triangle is properly colored. Furthermore,

- $\{b_{ij}, u_{ij}\}$ is properly colored, because $c(b_{ij}) = C_t, c(u_{ij}) = C$.
- $\{b_{ir}, u_{ir}\}$ with $r \neq j$ is properly colored, because $c(u_{ir}) = C_t$ and $c(b_{ir}) \neq C_t$.

We have given a proper coloring of all vertices using \vec{b} .

7.4 Trisexual marriage