

Chapter 36

Matchings

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36.1. Definitions

Definition 36.1.1. For a graph $G = (V, E)$ a set $M \subseteq E$ of edges is a **matching** if no pair of edges of M has a common vertex.

A matching is **perfect** if it covers all the vertices of G . For a weight function w , which assigns real weight to the edges of G , a matching M is a **maximal weight matching**, if M is a matching and $w(M) = \sum_{e \in M} w(e)$ is maximal.

Definition 36.1.2. If there is no weight on the edges, we consider the weight of every edge to be one, and in this case, we are trying to compute a **maximum size matching**.

Problem 36.1.3. Given a graph G and a weight function on the edges, compute the maximum weight matching in G .

36.2. Unweighted matching in a bipartite graph

We remind the reader that there is a simple way to do a matching in a bipartite graph using network flow. Since this was already covered, we will not repeat it here.

36.3. Matchings and Alternating Paths

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Consider a matching M . An edge $e \in M$ is a **matching edge**. Naturally, Any edge $e' \in E(G) \setminus M$ is **free**. In particular, a vertex $v \in V(G)$ is **matched** if it is adjacent to an edge in M . Naturally, a vertex v' which is not matched is **free**.

An **alternating path** is a simple path that its edges are alternately matched and free. An **alternating cycle** is defined similarly. The **length** of a path/cycle is the number of edges in it.

For an alternating path/cycle π , its **weight** is

$$\gamma(\pi, M) = \sum_{e \in \pi \setminus M} w(e) - \sum_{e \in \pi \cap M} w(e). \quad (36.1)$$

Namely, it is the total weight of the free edges in π minus the weight of the matched edges. This is a natural concept because of the following lemma.

Lemma 36.3.1. *Let M be a matching, and let π be an alternating path/cycle with positive weight such that*

$$M' = M \oplus \pi = (M \setminus \pi) \cup (\pi \setminus M)$$

is a matching, then $w(M')$ is bigger; namely, $w(M') > w(M)$.

Proof: Just observe that $w(M') = w(M) + \gamma(\pi, M)$. ■

Definition 36.3.2. An alternating path is **augmenting** if it starts and ends in a free vertex.

Observation 36.3.3. *If M has an augmenting path π then M is not of maximum size matching (this is for the unweighted case), since $M \oplus \pi$ is a larger matching.*

Theorem 36.3.4. *Let M be a matching, and T be a maximum size matching, and $k = |T| - |M|$. Then M has k vertex disjoint augmenting paths. At least one of length $\leq n/k - 1$.*

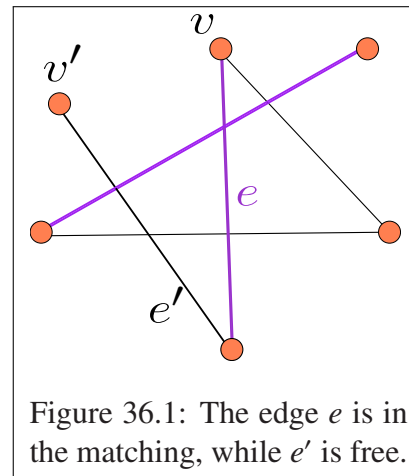
Proof: Let $E' = M \oplus T$, and let $H = (V, E')$. Clearly, every vertex in H has at most degree 2 because every vertex is adjacent to at most one edge of M and one edge of T . Thus, H is a collection of paths and (even length) cycles. The cycles are of even length since the edges of the cycle are alternating between two matchings (i.e., you can think about the cycle edges as being 2-colorable).

Now, there are k more edges of T in $M \oplus T$ than of M . Every cycle have the same number of edges of M and T . Thus, a path in H can have at most one more edge of T than of M . In such a case, this path is an augmenting path for M . It follows that there are at least k augmenting paths for M in H .

As for the claim on the length of the shortest augmenting path, observe that if all these (vertex disjoint) augmenting paths were of length $\geq n/k$ then the total number of vertices in H would be at least $(n/k + 1)k > n$. A contradiction. ■

Theorem 36.3.5. *Let M be a matching of maximum weight among matchings of size $|M|$. Let π be an augmenting path for M of maximum weight, and let T be the matching formed by augmenting M using π . Then T is of maximum weight among matchings of size $|M| + 1$.*

Proof: Let S be a matching of maximum weight among all matchings with $|M| + 1$ edges. And consider $H = (V, M \oplus S)$.



Consider a cycle σ in H . The weight $\gamma(\sigma, M)$ (see Eq. (36.1)) must be zero. Indeed, if $\gamma(\sigma, M) > 0$ then $M \oplus \sigma$ is a matching of the same size as M which is heavier than M . A contradiction to the definition of M as the maximum weight such matching.

Similarly, if $\gamma(\sigma, M) < 0$ then $\gamma(\sigma, S) = -\gamma(\sigma, M)$ and as such $S \oplus \sigma$ is heavier than S . A contradiction.

By the same argumentation, if σ is a path of even length in the graph H then $\gamma(\sigma, M) = 0$ by the same argumentation.

Let U_S be all the odd length paths in H that have one edge more in S than in M , and similarly, let U_M be the odd length paths in H that have one edge more of M than an edge of S .

We know that $|U_S| - |U_M| = 1$ since S has one more edge than M . Now, consider a path $\pi \in U_S$ and a path $\pi' \in U_M$. It must be that $\gamma(\pi, M) + \gamma(\pi', M) = 0$. Indeed, if $\gamma(\pi, M) + \gamma(\pi', M) > 0$ then $M \oplus \pi \oplus \pi'$ would have bigger weight than M while having the same number of edges. Similarly, if $\gamma(\pi, M) + \gamma(\pi', M) < 0$ (compared to M) then $S \oplus \pi \oplus \pi'$ would have the same number of edges as S while being a heavier matching. A contradiction.

Thus, $\gamma(\pi, M) + \gamma(\pi', M) = 0$. Thus, we can pair up the paths in U_S to paths in U_M , and the total weight of such a pair is zero, by the above argumentation. There is only one path μ in U_S which is not paired, and it must be that $\gamma(\mu, M) = w(S) - w(M)$ (since everything else in H has zero weight as we apply it to M to get S).

This establishes the claim that we can augment M with a single path to get a maximum weight matching of cardinality $|M| + 1$. Clearly, this path must be the heaviest augmenting path that exists for M . Otherwise, there would be a heavier augmenting path σ' for M such that $w(M \oplus \sigma') > w(S)$. A contradiction to the maximality of S . ■

The above theorem imply that if we always augment along the maximum weight augmenting path, than we would get the maximum weight matching in the end.

36.4. Maximum Weight Matchings in A Bipartite Graph

Let $G = (L \cup R, E)$ be the given bipartite graph, with $w : E \rightarrow \mathbb{R}$ be the non-negative weight function. Given a matching M we define the graph G_M to be the directed graph, where if $rl \in M$, $l \in L$ and $r \in R$ then we add $(r \rightarrow l)$ to $E(G_M)$ with weight $\alpha((r \rightarrow l)) = w(rl)$ Similarly, if $rl \in E \setminus M$ then add the edge $(l \rightarrow r) \in E(G_M)$ to G_M and set $\alpha((l \rightarrow r)) = -w(rl)$

Namely, we direct all the matching edges from right to left, and assign them their weight, and we direct all other edges from left to right, with their negated weight. Let G_M denote the resulting graph.

An augmenting path π in G must have an odd number of edges. Since G is bipartite, π must have one endpoint on the left side and one endpoint on the right side. Observe, that a path π in G_M has weight $\alpha(\pi) = -\gamma(\pi, M)$.

Let U_L be all the unmatched vertices in L and let U_R be all the unmatched vertices in R .

Thus, what we are looking for is a path π in G_M starting U_L going to U_R with maximum weight $\gamma(\pi)$, namely with minimum weight $\alpha(\pi)$.

Lemma 36.4.1. *If M is a maximum weight matching with k edges in G , than there is no negative cycle in G_M where $\alpha(\cdot)$ is the associated weight function.*

Proof: Assume for the sake of contradiction that there is a cycle C , and observe that $\gamma(C) = -\alpha(C) > 0$. Namely, $M \oplus C$ is a new matching with bigger weight and the same number of edges. A contradiction to the maximality of M . ■

The algorithm. So, we now can find a maximum weight in the bipartite graph G as follows: Find a maximum weight matching M with k edges, compute the maximum weight augmenting path for M , apply it, and repeat till M is maximal.

Thus, we need to find a minimum weight path in G_M between U_L and U_R (because we flip weights). This is just computing a shortest path in the graph G_M which does not have negative cycles, and this can just be done by using the **Bellman-Ford** algorithm. Indeed, collapse all the vertices of U_L into a single vertex, and all the uncovered vertices of U_R into a single vertex. Let H_M be the resulting graph. Clearly, we are looking for the shortest path between the two vertices corresponding to U_L and U_R in H_M and since this graph has no negative cycles, this can be done using the **Bellman-Ford** algorithm, which takes $O(nm)$ time. We conclude:

Lemma 36.4.2. *Given a bipartite graph G and a maximum weight matching M of size k one can find a maximum weight augmenting path for G in $O(nm)$ time, where n is the number of vertices of G and m is the number of edges.*

We need to apply this algorithm $n/2$ times at most, as such, we get:

Theorem 36.4.3. *Given a weight bipartite graph G , with n vertices and m edges, one can compute a maximum weight matching in G in $O(n^2m)$ time.*

36.4.1. Faster Algorithm

It turns out, in fact, that the graph here is very special, and one can use the Dijkstra algorithm. We omit any further details, and just state the result. The interested student can figure out the details (warning: this is not easy).

Theorem 36.4.4. *Given a weight bipartite graph G , with n vertices and m edges, one can compute a maximum weight matching in G in $O(n(n \log n + m))$ time.*

36.5. The Bellman-Ford Algorithm - A Quick Reminder

The **Bellman-Ford** algorithm computes the shortest path from a single source s in a graph G that has no negative cycles to all the vertices in the graph. Here G has n vertices and m edges. The algorithm works by initializing all distances to the source to be ∞ (formally, for all $u \in V(G)$, we set $d[u] \leftarrow \infty$ and $d[s] \leftarrow 0$). Then, it n times scans all the edges, and for every edge $(u \rightarrow v) \in E(G)$ it performs a **Relax** (u, v) operation. The relax operation checks if $x = d[u] + w((u \rightarrow v)) < d[v]$, and if so, it updates $d[v]$ to x , where $d[u]$ denotes the current distance from s to u . Since **Relax** (u, v) operation can be performed in constant time, and we scan all the edges n times, it follows that the overall running time is $O(mn)$.

We claim that in the end of the execution of the algorithm the shortest path length from s to u is $d[u]$, for all $u \in V(G)$. Indeed, every time we scan the edges, we set at least one vertex distance to its final value (which is its shortest path length). More formally, all vertices that their shortest path to s have i edges, are being set to their shortest path length in the i th iteration of the algorithm, as can be easily proved by induction. This implies the claim.

Notice, that if we want to detect negative cycles, we can ran **Bellman-Ford** for an additional iteration. If the distances changes, we know that there is a negative cycle somewhere in the graph.