

Combinatorial Optimization

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0 Organization

- Prerequisites
 - Basic knowledge of graph algorithms
 - Linear Programming (LP Duality)
 - Programming skills in C++
- Exam
 - Qualification requires 50% of the points in theoretical & programming exercises
 - Oral exam
- Books
 - "Combinatorial Optimization", Korte & Vygen
 - "Understanding & Using Linear Programming", B. Gärtner, J. Matoušek
 - Skript (theorems & definitions)
 - Further book recommendations are on the website

1 Matchings

1.1 Introduction

Definition 1.1.

1. A *matching* M in a graph $G = (V, E)$ is a set of pairwise disjoint edges, i.e. they don't have a common endpoint.
 $\nu(G) := \max.$ cardinality of a matching in G
2. An *edge cover* C of a graph $G = (V, E)$ is a subset of E s.t. $V = \bigcup_{e \in C} e$.
 $\zeta(G) := \min.$ cardinality of an edge cover in G
3. A matching is called *perfect* (or *1-factor*) if it is an edge cover
4. $v \in V$ with $v \in e \in M$ is called *M -covered*
5. $v \in V$ is called *M -exposed* if it is not M -covered

Definition 1.2.

1. A *stable set* (independent set) S is a set of pairwise non-adjacent vertices.
 $\alpha(G) := \max.$ cardinality of a stable set

2. A *vertex cover* C is a subset of V s.t. $E = \bigcup_{\{x,y\} \in E, x \in C} \{x, y\}$
 $\tau(G) := \min.$ cardinality of a vertex cover

Lemma 1.3.

1. $\alpha(G) + \tau(G) = |V|$
2. $\nu(G) + \zeta(G) = |V|$ if G has no isolated vertices
3. $\zeta(G) = \alpha(G)$ if G is bipartite and has no isolated vertices

Problem. Cardinalty Matching Problem

Input: Graph $G = (V, E)$

Task: Find a maximum cardinality matching

Problem. Maximum Weight Matching Problem (MWMP)

Input: Graph $G, c : E \rightarrow \mathbb{R}$

Task: Find a matching M maximizing $c(M)$

Problem. Minimum Weight Perfect Matching (MWPM)

Input: Graph $G, c : E \rightarrow \mathbb{R}$

Task: Find a perfect matching of minimum weight or decide that no perfect matching exists in G

Lemma 1.4. *The MWMP is equivalent to the MWPM (i.e. there exists a transformation with linear complexity)*

Proof. Given a MWPM instance (G, c) , define $c' := K - c$ ($K := 1 + \sum_{e \in E} |c(e)|$).

\Rightarrow Any maximum weight matching is a maximum cardinality matching

Given a MVMP instance (G, c) , define G' as 2 copies of G where the 2 copies of a vertex are joined by an edge.

$\Rightarrow G'$ has a perfect matching. Define:

$$c'(e) := \begin{cases} -c(e) & \text{if } e \text{ is in the first copy} \\ 0 & \text{else} \end{cases}$$

A minimum weight perfect matching in G' gives us a maximum weight matching in G . \square

Definition 1.5. Let $G = (V, E)$ be a graph and $M \subseteq E$ a matching in G . A path P is *M-alternating* if its edges are alternatingly in and not in M . If both end points of this path are *M-exposed*, P is an *M-augmenting* path.

Lemma 1.6. *Given a matching M in G and an inclusion-wise maximal M-alternating path P ,*

$$M \Delta P := M \setminus P \cup P \setminus M$$

is a matching. If P is M-augmenting, then $|M \Delta P| = |M| + 1$.



Figure 1: Example of the construction in Theorem 1.8

Theorem 1.7 (Petersen 1891, Berge 1957). *Augmenting Path Theorem*
 Given a graph $G = (V, E)$ and a matching M in G :

$$|M| = \nu(G) \Leftrightarrow \nexists M\text{-augmenting path } P \text{ in } G$$

Proof.

" \Rightarrow ": Clear

" \Leftarrow ": Assume there exists a matching M' with $|M'| > |M|$. Let $G' := (V, M \Delta M')$.

$$\Rightarrow |\delta_{G'}(v)| \leq 2 \quad \forall v \in V$$

$\Rightarrow G'$ is the union of disjoint circuits and paths

\Rightarrow all circuits are even and have the same number of edges from M and M'

$\Rightarrow \exists$ a path P in G' starting and ending with an edge in M'

$\Rightarrow P$ is an alternating path

□

1.2 Bipartite Matching

Theorem 1.8 (König 1931). *If G is bipartite, then $\nu(G) = \tau(G)$*

Proof. Add vertices s and t edges between them to all vertices of the respective partition. Direct all edges from s to t . Then $\nu(G)$ is maximum number of disjoint s - t -paths. Menger \Rightarrow This is equal to the minimum number of vertices that disconnect t from s . □

Theorem 1.9 (Hall 1935). *Let $G = (A \dot{\cup} B, E)$ be a bipartite graph. Then:*

$$G \text{ has a matching covering } A \Leftrightarrow |\Gamma(X)| \geq |X| \quad \forall X \subseteq A$$

Corollary 1.10. *Marriage Theorem*

$$|\Gamma(X)| \geq |X| \quad \forall X \subseteq A \text{ and } |A| = |B| \Leftrightarrow G \text{ has a perfect matching}$$

Definition 1.12. The MWPMP for bipartite graphs is called *Assignment Problem*.

Theorem 1.13. *The Assignment Problem can be solved in time $O(nm + n^2 \log m)$.*

Proof. Use the Successive Shortest Paths algorithm in an auxiliary graph. \square

1.3 The Tutte Matrix & Randomized Matching

Definition 1.14. Let G be a simple, undirected graph. Let G' be an orientation of G and $(X_e)_{e \in E(G)}$. The *Tutte matrix* is defined as

$$T_G(X) := (T_{vw}^*)_{v,w \in V(G)}$$

where

$$t_{vw}^* := \begin{cases} X_{\{v,w\}} & \text{if } (v,w) \in E(G) \\ -X_{\{v,w\}} & \text{if } -(v,w) \in E(G) \\ 0 & \text{else} \end{cases}$$

Remark 1.15. $T_G(X)$ is skew-symmetric (i.e. $T_G(X) = -(T_G(X))^t$). $\text{rank}(T_G(X))$ is independent of the orientation of G . $\det(T_G(X))$ is a polynomial in X .

Theorem 1.16 (Tutte). *A simple graph G has a perfect matching $\Leftrightarrow \det(T_G(X)) \neq 0$*

Proof. Let $V(G) = \{v_1, \dots, v_n\}$ and S_n be the permutation group.

$$\det T_G(X) = \sum_{\pi \in S_n} \text{sgn} \pi \cdot \prod_{i=1}^n t_{v_i, v_{\pi(i)}}^*$$

Let $S'_n := \{\pi \in S_n \mid \prod_{i=1}^n t_{v_i, v_{\pi(i)}}^* \neq 0\}$. Each $\pi \in S_n$ corresponds to a digraph $H_\pi := (V(G), \{(v_i, v_{\pi(i)}) \mid i \in [n]\})$. We have $|\delta^+(v)| = 1 = |\delta^-(v)| \ \forall v \in V(H_\pi) \Rightarrow H_\pi$ is the union of disjoint circuits. If $\pi \in S'_n$, then $H_\pi \subset G'$.

If there exists $\pi \in S'_n$ s.t. H_π is a collection of even circuits, then this immediately yields a perfect matching in G (using every second edge of each circuit).

Otherwise, $\forall \pi \in S'_n$, H_π contains an odd circuit. Let $r(\pi) \in S'_n$ arise from π by reversing edges on the unique odd circuit containing a vertex with minimum index $\Rightarrow r(r(\pi)) = \pi$ and $\text{sgn}(\pi) = \text{sgn}(r(\pi))$. The second part is true since for reversing an odd cycle, we need an even number of swaps. Let $v_{i_1}, \dots, v_{i_{2k+1}}$ be the "first" odd circuit. Then $r(\pi)$ is attained by $2k$ swaps: For $j = 1, \dots, k$ swap $(\pi(i_{2j-1}), \pi(i_{2k}))$ and $(\pi(i_{2j}), \pi(i_{2k+1}))$.

¹This is an abbreviation for $\{1, \dots, n\}$.

$\prod_{i=1}^n t_{v_i v_{\pi(i)}}^* = -\prod_{i=1}^n t_{v_i v_{r(\pi(i))}}^*$ since there is an odd number of sign changes to t^* . $\Rightarrow \det(T_G(X)) = 0$. We have shown that if G has no perfect matching, then $\det T_G(X) = 0$.

Assume that G has a perfect matching M . Define π as $\pi(i) = j, \pi(j) = i$ where $\{i, j\} \in M$. $\Rightarrow \prod_{i=1}^n t_{v_i v_{\pi(i)}}^* = \prod_{e \in M} -X_e^2$ cannot be canceled out. In particular, $\det T_G(X) \neq 0$. \square

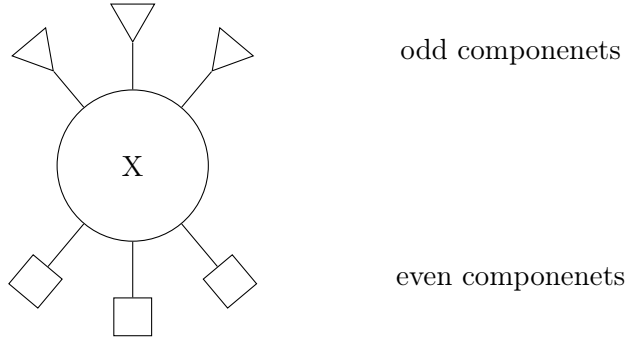
Remark 1.17. Picking $X' \in [0, 1]^{E(G)}$ at random, we almost surely have (since the zero set of a non-zero polynomial is a set of measure zero):

$$\det T_G(X') \neq 0 \Leftrightarrow G \text{ has a perfect matching}$$

Theorem 1.18 (Lovász 1979). *Let G be a simple graph and $X \in [0, 1]^{E(G)}$ chosen randomly. Then almost surely $\text{rank}(T_G(X)) = 2\nu(G)$.*

1.4 Tutte's Matching Theorem

Let $X \subseteq V(G)$. $G - X$ consists of even and odd (in terms of the number of vertices) connected components. We define $q_G(X)$ to be the number of odd components in $G - X$.



Definition 1.19. A graph G satisfies the *Tutte Condition* if $q_G(X) \leq |X|$ for all $X \subseteq V(G)$. $\emptyset \neq X \subseteq V(G)$ is called *barrier* if $q_G(X) = |X|$.

Proposition 1.20. *For any graph G and any $X \subseteq V(G)$:*

$$q_G(X) - |X| \equiv |V(G)| \pmod{2}$$

Definition 1.21. A graph G is *factor-critical* if $G - v$ has a perfect matching for all $v \in V(G)$. A matching is called *near-perfect* if it covers $|V(G)| - 1$ vertices.

Proposition 1.22. *If G is factor-critical, then it is connected.*

Theorem 1.23 (Tutte 1947). *A graph G has a perfect matching \Leftrightarrow Tutte Condition holds (i.e. $q_G(X) \leq |X| \ \forall X \subseteq V(G)$)*

Proof.

" \Rightarrow ": Clear

" \Leftarrow ": We proceed by induction on $|V(G)|$. The case $|V(G)| = 2$ is clear.

Generally, if the Tutte Condition holds, then $|V(G)|$ must be even (pick $X = \emptyset$). Proposition 1.20 $\Rightarrow q_G(X) - |X|$ is even. Every $x \in V(G)$ induces a barrier $\{x\}$. Let X be a maximum barrier. Then $G - X$ doesn't have any even components (since otherwise a single vertex of such a component could be added to X).

Claim: Each odd component is factor-critical.

Let C be an odd component in $G - X$, $v \in V(C)$. Assume that $C - v$ does not have a perfect matching. Induction Hypothesis $\Rightarrow C - v$ violates Tutte Condition.

$\Rightarrow \exists Y \subseteq V(C - v) : q_{C-v}(Y) > |Y|$

$\stackrel{1.20}{\Rightarrow} q_{C-v}(Y) \geq |Y| + 2$

Observe $X \cap \{v\} = Y \cap \{v\} = X \cap Y = \emptyset$:

$$\begin{aligned} q_G(X \cup Y \cup \{v\}) &= q_G(X) - 1 + q_C(Y \cup \{v\}) \\ &= |X| - 1 + q_{C-v}(Y) \\ &\geq |X| - 1 + |Y| + 2 \\ &= |X \cup Y| + 1 \\ &= |X \cup Y \cup \{v\}| \end{aligned}$$

$\Rightarrow X \cup Y \cup \{v\}$ is a barrier

\Rightarrow Claim

Let G' arise from G by contracting each odd component into a single vertex. We have $V(G') = X \dot{\cup} Z$ and G' is bipartite. We have to show that G' has a perfect matching. If not, then $\exists A \subseteq Z : |\Gamma_{G'}(A)| < |A|$
 $\Rightarrow q_G(\Gamma_{G'}(A)) \geq |A| > |\Gamma_{G'}(A)|$ which contradicts the Tutte Condition.

□

Theorem 1.24 (Berge 1958).

$$|V(G)| = 2\nu(G) + \max_{X \subseteq V(G)} (q_G(X) - |X|)$$

Proof. For $X \subseteq V(G)$, any matching has at least $q_G(X) - |X|$ uncovered vertices, so " \geq " holds.

For the other inequality, add $k := \max_{X \subseteq V(G)} (q_G(X) - |X|)$ new vertices and connect them to all existing vertices, yielding a new graph H .

We claim that H has a perfect matching. This then implies:

$$2\nu(G) + k \geq 2\nu(H) - k = |V(H)| - k = |V(G)|$$

Assume that H does not have a perfect matching. Then by Tutte's Theorem, there exists $Y \subseteq V(H)$ with $q_H(Y) > |Y|$. By 1.20, $k \equiv |V(G)| \pmod{2}$, therefore $|V(H)|$ is even, so $Y \neq \emptyset$. Y must contain all new vertices, otherwise $H - Y$ would be connected² and $q_H(Y) \leq 1 \leq |Y|$.

$$\Rightarrow q_G(Y \cap |V(G)|) = q_H(Y) > |Y| = |Y \cap V(G)| + k$$

which is a contradiction to the choice of k . □

1.5 Ear Decompositions of Factor-Critical Graphs

Definition 1.25. Let G be a graph. An *ear decomposition* of G is a sequence r, P_1, \dots, P_k with $G = (r, \emptyset) + P_1 + \dots + P_k$ such that each P_i is either a path with exactly the endpoints located in $\{r\} \cup \bigcup_{j \in [i-1]} V(P_j)$ or a circuit where exactly one of the vertices belongs to $\{r\} \cup \bigcup_{j \in [i-1]} V(P_j)$.

P_1, \dots, P_k are called *ears*. If $|V(P_1)| \geq 3$ and P_2, \dots, P_k are paths we call it a *proper ear decomposition*.

Theorem 1.27 (Whitney 1932). *Let G be an undirected graph. Then:*

$$G \text{ 2-connected} \Leftrightarrow G \text{ has a proper ear decomposition}$$

Definition 1.28. An ear decomposition is *odd* if every ear has odd length (in terms of the number of edges).

Theorem 1.29. *Let G be an undirected graph. Then*

$$G \text{ factor-critical} \Leftrightarrow G \text{ has an odd ear decomposition}$$

The first vertex r of the ear decomposition can be chosen arbitrarily.

Proof.

" \Leftarrow ": Let G be a graph with an odd ear decomposition r, P_1, \dots, P_k . P_1 is an odd circuit, so it is factor-critical. We use induction on the number of ears. Let P be the last ear and G' be G before adding P . By the induction hypothesis, G' is factor-critical. Given $v \in V(G)$, we have to show that $G - v$ has a perfect matching.

Case 1: $v \in V(G')$. Then $G' - v$ has a perfect matching. Adding every second edge of P (excluding the endpoints) to it, yields a perfect matching of $G - v$.

Case 2: $v \in V(G) \setminus V(G')$. Let x, y be the endpoints of P . Without loss of generality let $P_{[v,x]}$ be even. There exists a perfect matching in $G' - x$. Together with every second edge of $P_{[v,y]}$ and $P_{[v,x]}$ this is a perfect matching in $G - v$.

²Note that Y cannot contain all old vertices, since otherwise $q_H(Y) < |Y|$.

" \Rightarrow ": Let $r \in V(G)$ be any vertex. Let M be a perfect matching in $G - r$. Suppose we have an odd ear decomposition for $G' \subseteq G$ with $r \in V(G')$ and $M \cap E(G')$ is a near-perfect matching in G' (i.e. all vertices in G' except for r are matched with other vertices in G').

If $G' \neq G$, there is an edge $\{x, y\} \in E(G) \setminus E(G')$ with $x \in V(G')$ (by Proposition 1.22). If $y \in V(G')$, then $\{x, y\}$ can be chosen as the next ear. Otherwise, construct an M -alternating odd ear, starting with $\{x, y\}$. Let N be a matching in $G - y$. $M \Delta N$ contains a y - r -path P . Let w be the first vertex in $P \cap V(G')$. w is M -exposed in $P_{[y, w]}$, y is N -exposed in $P_{[y, w]}$. Therefore $P_{[y, w]}$ is even and together with $\{x, y\}$ it forms an M -alternating odd ear.

Inductively, this argument yields an odd ear decomposition. □

Definition 1.30. Let G be factor-critical and M a near-perfect matching. An M -alternating ear decomposition is an odd ear decomposition such that each ear is an M -alternating path or circuit C with:

$$|E(C) \cap M| = |E(C) \setminus M| - 1$$

Corollary 1.31. *For any factor-critical graph G and any near-perfect matching M in G , there exists in M -alternating ear decomposition of G .*

Definition 1.32. Let G be factor-critical, M a near-perfect matching and r, P_1, \dots, P_k an M -alternating ear decomposition of G . $\mu, \varphi : V(G) \rightarrow V(G)$ are associated with the ear decomposition if:

- $\{x, y\} \in M \Rightarrow \mu(x) = y$
- $\{x, y\} \in E(P_i) \setminus M$ and $x \notin \{r\} \cup \bigcup_{j \in [i]} V(P_j) \Rightarrow \varphi(x) = y$
- $\mu(r) = \varphi(r) = r$

Proposition 1.33. *Let G be a factor-critical graph and μ, φ functions associated with an M -alternating ear decomposition. Then this ear decomposition is unique up to the order of the ears. The Ear-Decomposition-Algorithm correctly determines an explicit list of the ears in linear time.*

Proof. Step 3 determines ears uniquely. The algorithm clearly runs in linear time. □

Lemma 1.34. *Let G be factor-critical and μ, φ associated with an M -alternating ear decomposition. Then the maximal path given by the initial sequence*

$$x, \mu(x), \varphi(\mu(x)), \mu(\varphi(\mu(x))), \dots \tag{1}$$

defines an M -alternating x - r -path of even length.

Algorithm 1: Ear Decomposition Algorithm

Input: Factor-critical graph G , functions μ, φ associated with an M -alternating ear decomposition

Output: An M -alternating ear decomposition r, P_1, \dots, P_k

```
1  $X := \{r\}$  where  $r$  is the vertex with  $\mu(r) = r$ 
2  $k := 0$ ,  $S :=$  empty stack
3 while  $X \neq V(G)$  do
4   if  $S$  is non-empty then
5     | Let  $v \in V(G) \setminus X$  be an endpoint of the topmost element of
6     | the stack
7   else
8     | Choose  $v \in V(G) \setminus X$  arbitrarily
9    $x := v$ ,  $y := \mu(v)$ ,  $P := (\{x, y\}, \{\{x, y\}\})$ 
10  while  $\varphi(\varphi(x)) = x$  do
11    |  $P := P + \{x, \varphi(x)\} + \{\varphi(x), \mu(\varphi(x))\}$ 
12    |  $x := \mu(\varphi(x))$ 
13  while  $\varphi(\varphi(y)) = y$  do
14    |  $P := P + \{y, \varphi(y)\} + \{\varphi(y), \mu(\varphi(y))\}$ 
15    |  $y := \mu(\varphi(y))$ 
16   $P := P + \{x, \varphi(x)\} + \{y, \varphi(y)\}$ 
17   $P$  is the ear containing  $y$  as an inner vertex. Put  $P$  on  $S$ .
18  while Both endpoints of the topmost element  $P$  of the stack  $S$ 
19  | are in  $X$  do
20    | Delete  $P$  from  $S$ 
21    |  $k := k + 1$ ,  $P_k := P$ ,  $X := X \cup V(P)$ 
22 forall  $\{y, z\} \in E(G) \setminus (E(P_1) \cup \dots \cup E(P_k))$  do
23   |  $k := k + 1$ ,  $P_k := (\{y, z\}, \{\{y, z\}\})$ 
24 return  $r, P_1, \dots, P_k$ 
```

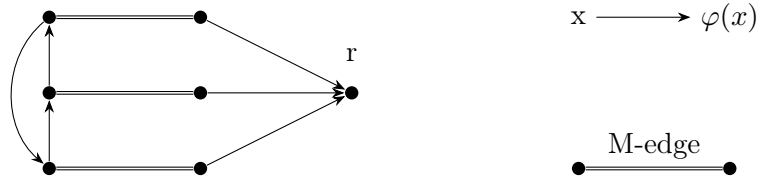


Figure 2: Counter example for the reverse implication of lemma 1.34

Proof. We proceed by induction on the number of ears. Let $x \in V(G) \setminus \{r\}$ and P_i be the ear containing x . A subsequence of (1) is a subpath Q of P_i from x to $y \in \{r\} \cup \bigcup_{j \in [i-1]} V(P_j)$. Q starts with a matching edge and ends with a non-matching edge, so it has even length. If $y = r$, we are done, otherwise the statement follows from the induction hypothesis. \square

1.6 Edmond's Matching Algorithm

Definition 1.35. Let G be a graph, M a matching in G . A *blossom* in G with respect to M is a factor-critical subgraph B of G such that $|M \cap E(B)| = \frac{|V(B)|-1}{2}$. The vertex $r \in V(B)$ that is exposed by M is called the *base* of B .

Definition 1.36. Let G be a graph, M a matching in G , B a blossom and Q a M -alternating v - r -path of even length from $v \in V(G)$ that is M -exposed to the base r of B . Additionally, let $E(Q) \cap E(B) = \emptyset$. $B + Q$ is called a M -flower.

Lemma 1.37. Let G be a graph, M a matching in G . Suppose there is a M -flower $B + Q$. Let G', M' result from G and M by contracting $V(B)$ into a single vertex. Then:

$$M \text{ maximum matching in } G \Leftrightarrow M \text{ maximum matching in } G'$$

Proof.

" \Leftarrow ": Assume that M is not maximum in G . $N := M \Delta E(Q)$ is a matching with $|N| = |M|$.

$\Rightarrow \exists N$ -augmenting path P in G . At least one endpoint x of P is in $V(B)$. If P and B are disjoint, let y be the other endpoint of P . Otherwise, let y be the first vertex on P in B . $P' := P_{[x,y]}$ is an N' -augmenting path in G' , so $|N'| = |M'| < \mu(G')$.

" \Rightarrow ": Assume that M' is not maximum in G' , so there exists a matching N' in G' with $|N'| > |M'|$. Let N_0 arise from N' in G , then N_0 contains ≤ 1 vertex from $V(B)$. Since B is factor-critical, N_0 can be extended by $k := \frac{|V(G)|-1}{2}$ edges to a matching N in G . We have

$$|N| = |N_0| + k = |N'| + k > |M'| + k = |M|$$

so M is not maximum. \square

Lemma 1.39. Let G be a graph, M a matching in G . $X \subseteq V(G)$ is the set of exposed vertices. We can find a shortest M -alternating X - X -walk of positive length in $O(|E(G)|)$ time.

Proof. Define $D := (V(G), A)$ where:

$$A := \{(u, v) \mid \exists x \in V(G) : \{u, x\} \in E(G), \{x, v\} \in M\}$$

A shortest $X - \Gamma_G(X)$ -path in D corresponds to a shortest X - X -walk in G . \square

Theorem 1.40. *Let $P = v_0, \dots, v_t$ be a shortest M -alternating X - X -walk in G . Then either*

- P is an M -augmenting path or
- v_0, \dots, v_j is an M -flower for some $j \leq t$.

Proof. If P is not a path, choose $i < j$ such that $v_i = v_j$ and j minimal. Then v_0, \dots, v_{j-1} are distinct vertices. If $j - i$ is even, deleting v_{i-1}, \dots, v_j from P yields a shorter walk, so $j - i$ is odd.

Case 1: j is even. Then i is odd and therefore $v_{i+1} = v_{j-1}$ must be the matching mate of $V_i = v_j$ which contradicts the minimality of j .

Case 2: j is odd. Then i is even, so v_0, \dots, v_i is an M -alternating path of even length and v_i, \dots, v_j is an M -alternating odd circuit, i.e. a blossom. \square

Algorithm 2: Edmond's Augmenting Path Search

Input: Graph G , matching M

Output: An M -augmenting path (if one exists)

```

1  $X :=$  set of exposed vertices
2 if  $\exists M$ -alternating  $X$ - $X$ -walk of positive length then
3    $P = v_0, \dots, v_t :=$  a shortest such walk
4   if  $P$  is a path then
5     return  $P$ 
6   else
7     Choose  $j$  as in Theorem 1.40
8      $v_0, \dots, v_j$  is an  $M$ -flower with blossom  $B$ 
9     Recurse on  $G/B$ 
10    Augment an  $M/B$ -augmenting path in  $G/B$  to an
       $M$ -augmenting path  $P'$  in  $G$ 
11    return  $P'$ 
12 else
13    $\nexists M$ -augmenting path
```

Theorem 1.41. *Given a graph G , a maximum cardinality matching can be found in time $O(n^2m)$ where $n := |V(G)|$, $m := |E(G)|$*

Proof. Start with $M = \emptyset$ and iteratively find M -augmenting path P , set $M := M \Delta E(P)$. If no such path exists, then M is maximum. P can be found in time $O(mn)$ ³. Since M is maximum after at most $\frac{n}{2}$ augmentation, we have total running time $O(n^2m)$. \square

1.6.1 Growing forest - $O(n^3)$

Definition 1.42. Let G be a graph, M a matching in G . An *alternating forest* with respect to M in G is a forest F in G where:

- $V(F)$ contains all M -exposed vertices, each tree of F contains exactly one exposed vertex, its *root*.
- We call $v \in V(G)$ an outer (inner) vertex if it has even (odd) distance from the root of its component.
- $\forall v \in V(F)$ the unique path from v to the root of its component is M -alternating.
- $v \in V(G) \setminus V(F)$ is called *out-of-forest*.

Clearly, inner vertices always have degree 2 (we always assume that there are no matching edges that can immediately be added to F).

Proposition 1.43. *In any alternating forest, the number of outer vertices that are not the root equals the number of inner vertices.*

Proof. For all outer vertices, there exists exactly one inner vertex on its path to the root. \square

Lemma 1.44. *Given a graph G , a matching M , an alternating forest F with respect to M in G . Then, either M is a maximum matching or \exists outer vertex $x \in V(F)$, an edge $\{x, y\} \notin E(F)$ such that one of the following holds:*

- *Grow:* $y \notin V(F)$ and therefore $\{y, z\} \in M$ with $z \notin V(F)$. In this case, y, z and $\{x, y\}, \{y, z\}$ can be added to F .
- *Augment:* y is an outer vertex in a different connected component in F . In this case, M can be augmented along $P(x) \cup \{x, y\} \cup P(y)$ where $P(z)$ denotes the unique path from $z \in V(F)$ to the root of its connected component.

³Here, m is the time required for finding a walk and the recursion depth is bounded by n .

- *Shrink*: y is an outer vertex in the same component as x . Let r be the first vertex on $P(x)$ that is also on $P(y)$. Then $|\delta_F(r)| \geq 3$, so y is an outer vertex and $|E(F_{[x,r]})|, |E(F_{[y,r]})|$ are even. Together with $\{x, y\}$ these paths form a blossom with ≥ 3 vertices.

Proof. We show that if none of these cases apply, M is maximum. Let X be the set of inner vertices, $s := |X|$ and t be the number of outer vertices. All outer vertices are isolated in $G - X$, so $G - X$ and $q_G(X) - |X| = t - s$. By Berge's formula (1.24), $t - s$ vertices are exposed by any matching, so M is maximum. \square

Definition 1.45. Let G be a graph, M a matching in G . A subgraph F of G is a *general blossom forest* with respect to M if there exists a partition $V(F) = V_1 \dot{\cup} \dots \dot{\cup} V_k$ such that $F_i = F[V_i]$ is a maximal factor-critical subgraph of F with $|M \cap E(F_i)| = \frac{|V_i| - 1}{2}$ ($i \in [k]$) and after contracting each V_i , we obtain an M -alternating forest F' . F_i is called an outer (inner) blossom if V_i is an outer (inner) vertex in F' .

A *special blossom forest* is a general blossom forest where each inner blossom is a single vertex.

Store a special blossom forest with 3 functions $\mu, \varphi, \rho : V(G) \rightarrow V(G)$:

$$\begin{aligned} \mu(x) &:= \begin{cases} x & \text{if } x \text{ is exposed in } M \\ y & \text{if } \{x, y\} \in M \end{cases} \\ \varphi(x) &:= \begin{cases} x & \text{if } x \text{ is the base of an outer blossom or } x \text{ is out-of-forest} \\ y & \text{if } x \text{ is an inner vertex and } \{x, y\} \in E(F) \setminus M \\ y & \text{if } x \text{ is an outer vertex (i.e. in an outer blossom)} \\ & \text{and } \mu, \varphi \text{ are associated with an } M\text{-alternating} \\ & \text{ear decomposition of } x\text{'s blossom, } \{x, y\} \in \\ & E(F) \setminus M \end{cases} \\ \rho(x) &:= \begin{cases} x & \text{if } x \text{ is an inner vertex or out-of-forest} \\ y & \text{if } x \text{ is an outer vertex and } y \text{ is the base of the} \\ & \text{outer blossom containing } x \text{ (} y = x \text{ is possible).} \end{cases} \end{aligned}$$

Proposition 1.46. Let F be a special blossom forest with respect to M and μ, φ, ρ as above. Then:

1. For all outer vertices x , $P(x) :=$ maximal path given by subsequence of $x, \mu(x), \varphi(\mu(x)), \mu(\varphi(\mu(x))), \dots$ is an M -alternating path from x to q where q is the root of the component containing x .
2. A vertex x is

- an outer vertex $\Leftrightarrow \mu(x) = x \vee \varphi(\mu(x)) \neq \mu(x)$

- *an inner vertex* $\Leftrightarrow \varphi(\mu(x)) = \mu(x) \wedge \varphi(x) \neq x$
- *out-of-forest* $\Leftrightarrow \mu(x) \neq x \wedge \varphi(x) = x \wedge \varphi(\mu(x)) = \mu(x)$

Proof.

1. By definition of μ, φ and lemma 1.33 some initial subsequence of $P(x)$ ends at the base r of the blossom containing x . If $r = q$, we are done. Otherwise $\mu(r), \varphi(\mu(r))$ are next elements in a sequence leading to outer vertex $\varphi(\mu(r))$. This can be iterated.
2. Since the conditions are mutually exclusive, it suffices to show one implication for all the statements.
 - If x is outer, it is a root ($\mu(x) = x$) or $P(x)$ is a path of length at least 2, so $\varphi(\mu(x)) \neq \mu(x)$.
 - If x is inner, then $\mu(x)$ is the base of an outer blossom. Therefore $\varphi(\mu(x)) = \mu(x)$. $P(\mu(x))$ is a path of length at least 2, so $\varphi(x) \neq x$.
 - If x is out-of-forest, then x is covered by M so $\mu(x) \neq x$. By definition of φ , $\varphi(x) = x$. $\mu(x)$ is out-of-forest as well, so $\varphi(\mu(x)) = \mu(x)$.

□

Lemma 1.47. *Following invariants hold:*

- a) $\{\{x, \mu(x)\} \mid x \in V(G), \mu(x) \neq x\}$ is a matching
- b) $\{\{x, \mu(x)\} \mid \underbrace{x \in V(G), \varphi(\mu(x)) = \mu(x) \wedge \varphi(x) \neq x}_{\text{inner vertices}}\} \cup \{\{x, \varphi(x)\} \mid x \in V(G), \varphi(x) \neq x\}$ forms the edge set of a special blossom forest.
- c) μ, φ, ρ satisfy the conditions in definition 1.45 (special blossom forest).

Proof. a) holds as μ only changes in *Augment*. b) is correct after initialization and after the reset in the *Augment* step. It is preserved by *Grow* steps.

In a *Shrink* step, r (the first vertex that the paths from x, y to the root share) is a root or has $|\delta(r)| = 3$ (i.e. it is the base of a blossom), so it is an outer vertex. We define a blossom $B := \{v \in V(G) \mid \varphi(v) \in V(P(x)_{[x,r]}) \cup V(P(y)_{[y,r]})\}$. Consider $\{u, v\} \in F$ with $u \in B, v \notin B$. If $\{u, v\} \in M$, we have $u = r, v = \mu(r)$ (since $F[B]$ contains a near-perfect matching). u was an outer vertex before shrinking and $F[B]$ being factor-critical follows from c) and the characterization by ear-decompositions.

For c), it's clear that μ always represents a matching. $\varphi(x) = x$ if x is not an outer vertex. Therefore, $\mu + \varphi$ represent an M -alternating ear decomposition of B . During *Shrink*, $\varphi(v)$ is not changed if $\varphi(v) = r$. Therefore, the

odd ear decomposition for $B' := \text{blossom containing } r$, is the correct starting point. The next ear is $P(x)_{[x,x']} \cup P(y)_{[y,y']} + \{x, y\}$, where x' (y') is the first vertex in B' on $P(x)_{[x,r]}$ ($P(y)_{[y,r]}$).

For each ear Q of a former blossom $B'' \subseteq B$, $Q \setminus (E(P(x)) \cup E(P(y)))$ form a new ear (since it is created by removing an even path). φ, μ represent this ear-decomposition. \square

Theorem 1.48. *Edmond's cardinality matching algorithm correctly determines a maximum matching in $O(n^3)$ time, where $n := |V(G)|$.*

Proof. By lemma 1.47 and proposition 1.46, the algorithm maintains a special blossom forest. Let M, F be the final matching and forest. x an outer vertex implies that $\forall y \in \Gamma(x) : y$ is inner and $\varphi(y) = \varphi(x)$. Define:

$B := \text{set of inner vertices}$

$B := \text{set of bases of (outer) blossoms}$

Then every unmatched vertex is in B . Matched vertices in B have matching mates in X and $|B| = |X| + |V(G)| - 2|M|$. (Outer) blossoms are odd connected components in $G - X$, so by Berge's theorem (1.24), at least $|B| - |X|$ vertices remain uncovered by any matching, so M is maximum.

We now consider the running time: The status (outer, inner, out-of-forest) for a given vertex can be checked in constant time (proposition 1.46). Therefore, *Grow*, *Augment* and *Shrink* can be implemented in $O(n)$ time. There are at most n calls to *Grow* and *Shrink* per augment and at most $\frac{n}{2}$ *Augments*. This implies the running time $O(n^3)$. \square

Remark 1.49. The time for *Shrink* can be reduced to $O(\log n)$ using a binary tree, leading to a running time of $O(nm \log n)$ in total. Tarjan (1974), Gabow & Tarjan (1983) proved a running time of $O(nm\alpha(m, n))$ (where α is the inverse Ackermann function) or $O(nm)$.

Remark 1.50. It's not necessary to reset everything after augmenting. It suffices to reset the 2 trees that were changed by the augmentation. Gabow & Tarjan (1983) showed that it's possible to augment all paths of the same length in $O(m)$ time. There are $2\sqrt{\nu(G)} + 2$ different path lengths, so in total this results in a running time of $O(\sqrt{nm})$.

Remark 1.51 (Skew-symmetric flows). Goldberg & Karzanov (2003) (and Fremuth-Pagen & Jungnickel (2003)) used *Generalized Max-Flow* to achieve a running time of $O(\sqrt{nm} \frac{\log \frac{m}{n}}{\log n})$.

1.7 Gallai-Edmonds Decomposition

Proposition 1.52. *Let G be a graph, $X \subseteq V(G)$ with $|V(G)| - 2\nu(G) = q_G(X) - |X|$. Then any maximum matching of G*

Algorithm 3: Edmond's Cardinality Matching Algorithm

Input: A graph G

Output: A maximum matching M (defined by $\{x, \mu(x)\}$)

1 $\mu(v) := v, \varphi(v) := v, \rho(v) := v, \text{scanned}(v) := \text{false}$ for all $v \in V(G)$

 // Outer Vertex Scan:

2 **while** \exists outer vertex x with $\text{scanned}(x) = \text{false}$ **do**

3 Let y be a neighbor of x such that y is either out-of-forest or y is
 outer with $\rho(y) \neq \rho(x)$

4 **if** *such a y does not exist* **then**

5 $\text{scanned}(x) = \text{true}$, **continue**

 // Grow:

6 **if** y is out-of-forest **then**

7 $\varphi(y) := x$, **continue**

 // Augment:

8 **else if** $P(x)$ and $P(y)$ are vertex-disjoint **then**

9 $\mu(\varphi(v)) = v, \mu(v) = \varphi(v)$ for all $v \in V(P(x) \cap P(y))$ with
 odd distance from x or y on $P(x)$ or $P(y)$, respectively

10 $\mu(x) := y, \mu(y) := x$

11 $\varphi(v) := v, \rho(v) := v, \text{scanned}(v) := \text{false}$ for all $v \in V(G)$

 // Shrink:

12 **else**

13 Let r be the first vertex on $V(P(x)) \cap V(P(y))$ with $\rho(r) = r$
14 **forall** $v \in V(P(x)_{[x,r]}) \cup V(P(y)_{y,r})$ with odd distance from x
 or y on $P(x)_{[x,r]}$ or $P(y)_{[y,r]}$, respectively and $\rho(\varphi(v)) \neq r$

do

15 $\varphi(\varphi(v)) := v$

16 **if** $\rho(x) \neq r$ **then**

17 $\varphi(x) := y$

18 **if** $\rho(y) \neq r$ **then**

19 $\varphi(y) := x$

20 **forall** $v \in V(G)$ with $\rho(v) \in V(P(x)_{[x,r]}) \cup V(P(y)_{[y,r]})$ **do**

21 $\rho(v) := r$

22 **return** μ

- contains a perfect matching in the even components of $G - X$.
- contains a near-perfect matching in odd components of $G - X$.
- matches all $x \in X$ to distinct odd components.

Proof. Follows directly from Berge's theorem (1.24). \square

Theorem 1.53. *Let G be a graph and:*

$$Y := \{v \in V(G) \mid \exists \text{ maximum matching that exposes } v\}$$

Define $X := \Gamma(Y)$ and $W := V(G) \setminus (X \cup Y)$. Then:

1. X attains $\max_{X' \subseteq V(G)} q_G(X') - |X'|$.
2. $G[Y]$ is the union of factor-critical subgraphs and $G[W]$ is the union of even connected components.
3. Any maximum matching in G
 - contains a perfect matching in $G[W]$.
 - contains a near-perfect matching in each component of $G[Y]$.
 - matches all $x \in X$ to distinct connected components

Y, X, W is called Gallai-Edmonds decomposition of G .

Proof. Consider the matching M and special blossom forest F at the end of the algorithm. Let X' (Y') be the set of inner (outer) vertices and W' the set of out-of-forest vertices.

Claim. X', Y', W' satisfy 1., 2. and 3.

(Proof of theorem 1.48).

Proposition 1.52 implies that any maximum matching covers all vertices in $V(G) \setminus Y'$, so $Y \subseteq Y'$. For the other inclusion, let $v \in Y'$. Then $M \Delta P(v)$ is a maximum matching exposing v , so $v \in Y$ and $Y' = Y$. By definition, $X = X'$ and $W = W'$. \square

Corollary 1.54. *A graph G has a perfect matching $\Leftrightarrow \forall U \subseteq V(G)$, $G - U$ has at most $|U|$ factor-critical components.*

1.8 Minimum Weight Perfect Matching

We use the following Integer Programming formulation:

$$\begin{aligned} \min \quad & \sum_{e \in E(G)} c_e x_e \\ \text{s.t.} \quad & \sum_{e \in \delta(v)} x_e = 1 \quad v \in V(G) \\ & x_e \in \{0, 1\} \end{aligned}$$

and the corresponding relaxation where we only require $x_e \geq 0$. The dual problem of this is:

$$\begin{aligned} \max \quad & \sum_{v \in V(G)} z_v \\ \text{s.t.} \quad & z_v + z_w \leq c_e \quad \{v, w\} \in E(G) \end{aligned}$$

Proposition 1.55 (Hungarian Method). *Let G be a graph, $c \in \mathbb{R}^{E(G)}$ and $z \in \mathbb{R}^{V(G)}$ with $z_v + z_w \leq c_e$ for all $e = \{v, w\} \in E(G)$. Define:*

$$G_z := (V(G), \{e = \{v, w\} \mid z_v + z_w = c_e\})$$

Let M be a matching in G_z , F a maximal alternating forest in G_z with respect to M . Let X/Y be the set of inner/outer vertices. Then:

1. *If M is a perfect matching in G_z , then it is a minimum-weight perfect matching in G .*
2. *If $\Gamma_G(y) \subseteq X$ for all $y \in Y$, then M is a maximum matching.*
3. *If neither 1. nor 2. hold, define:*

$$\epsilon := \min \left\{ \min_{e=\{v,w\} \in E(G[Y])} \frac{c_e - z_v - z_w}{2}, \min_{e \in \delta(Y) \cap \delta(V(F))} c_e - z_v - z_w \right\}$$

Set $z'_v := z_v - \epsilon$ for all $v \in X$, $z'_v := z_v + \epsilon$ for all $v \in Y$ and $z'_v := z_v$ for all $v \in V(G) \setminus (X \cup Y)$. Then z' is a feasible dual solution and $M \cup E(F) \subseteq E(G_{z'})$. Additionally, $\Gamma_{G_{z'}}(y) \setminus X \neq \emptyset$ for some $y \in Y$.

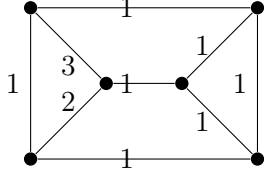
Proof. 1. Let M' be a minimum-weight perfect matching.

$$\begin{aligned} \sum_{e \in M'} c_e &= \sum_{v \in V(G)} z_v + \sum_{e=\{v,w\} \in M'} (c_e - z_v - z_w) \\ &\geq \sum_{v \in V(G)} z_v \\ &= \sum_{v \in V(G)} z_v + \sum_{e=\{v,w\} \in M} (c_e - z_v - z_w) \\ &= \sum_{e \in M} c_e \end{aligned}$$

2. Each outer vertex is an odd blossom (singleton) of $G - x$. By Berge (1.24), at least $|Y| - |X|$ vertices remain uncovered.
3. z' stays feasible by the choice of ϵ . Edges in $E(F), M$ remain tight. By 1. and 2., $\exists y \in Y : \Gamma_{G_{z'}}(y) \setminus X \neq \emptyset$.

□

Remark 1.56. For bipartite graphs, the adjacency matrix is totally unimodular, so the LP has integral vertices.



We define $\mathcal{A} := \{X \subseteq V(G) \text{ odd}\}$ and add the blossom inequalities

$$\sum_{e \in \delta(X)} x_e \geq 1 \quad \forall X \in \mathcal{A}$$

to the LP relaxation. The new dual problem is then:

$$\begin{aligned} & \max \sum_{A \in \mathcal{A}} z_A \\ & \text{s.t.} \quad \sum_{A \in \mathcal{A}, e \in \delta(A)} z_A \leq c_e \\ & \quad z_A \geq 0 \quad (A \in \mathcal{A}, |A| \geq 3) \end{aligned}$$

Edmond's Algorithm starts with an empty matching $x = 0$ and dual feasible solution:

$$z_A := \begin{cases} \frac{1}{2} \min\{c(e) \mid e \in \delta(A)\} & |A| = 1 \\ 0 & \text{else} \end{cases}$$

We always ensure that z is dual feasible and that (x, z) satisfy complementary slackness:

$$\begin{aligned} x_e > 0 &\Rightarrow \sum_{A \in \mathcal{A}, e \in \delta(A)} z_A = c_e \\ z_A > 0, |A| > 1 &\Rightarrow \sum_{e \in \delta(A)} x_e = 1 \end{aligned}$$

Definition 1.57. $c_z(e) := c(e) - \sum_{A \in \mathcal{A}, e \in \delta(A)} z_A$ is the *reduced cost* of e .

Theorem 1.58. *There are at most $\frac{7}{2}|V(G)|^2$ of the repeat-until loop in algorithm 4.*

Proof. \mathcal{B} is laminar at any time, i.e. for $X, Y \in \mathcal{B}$ we have $(X \subseteq Y) \vee (Y \subseteq X) \vee (X \cap Y = \emptyset)$. Therefore $|\mathcal{B}| \leq 2|V(G)|$.

Observation. *Any U added to \mathcal{B} during Shrink will not be "unpacked" before the next Augment.*

Proof. After *Shrink*, there exists an even length M -augmenting R - U -path. It remains in G_z until the next *Augment* or until U is included in another blossom $U' \supseteq U$ which is not resolved before an *Augment* (inductively). \square

Between 2 augments:

- $\# \text{ Unpacks} \leq |\mathcal{B}|$ at beginning of the sequence
- $\# \text{ Shrinks} \leq |\mathcal{B}|$ at the end of the sequence

Therefore, there are at most $4|V(G)|$ *Unpack* and *Shrink* operations between 2 augments. For each dual change without *Unpack*, we have: $z_B > 0 \quad \forall B \in \mathcal{B}$, so ϵ is not determined by z_B . Therefore $\exists e = \{X, Y\}$ with $X \notin \mathcal{X}, Y \in \mathcal{Y}$ where $c_z(e)$ becomes 0.

Case 1: $X \notin \mathcal{Y}$. Then $|V(G_z) \setminus (\mathcal{X} \cup \mathcal{Y})|$ decreases.

Case 2: $X \in \mathcal{Y}$. Then $\exists X$ - Y M -alternating walk in the next iteration.

In particular, such a dual change can occur at most $|V(G)|$ times between 2 augmentations.

In total, there are at most $\frac{1}{2}|V(G)|$ *Augment* steps. Therefore, there are $\frac{1}{2}|V(G)|^2 (4 + |V(G)| + 2|V(G)|)$ \square

Algorithm 4: Minimum-Weight Perfect Matching

Input: Graph G with edge weights $c : E(G) \rightarrow \mathbb{R}$

Output: A minimum-weight perfect matching M in (G, c)

Corollary 1.59. *A minimum-weight perfect matching can be computed in $O(n^2m)$ time where $n := |V(G)|$ and $m = |E(G)|$.*

Proof. Theorem 1.58 times $O(m)$. \square

Remark 1.60. To achieve $O(n^3)$ running time, one can modify the algorithm:

1. Use a General Blossom Forest to avoid recomputing the R - R -walks from scratch. We then have mappings $\mu_v, \varphi_v^i, \rho_v^i$ for $1 \leq i \leq k_v$ where k_v is the number of blossoms that contain v .

2. Store all vertices in a heap (ordered by their criticality for dual-feasibility) to speed up the computation of ϵ .

Gabow (1990) showed a running time of $O(n(m+n \log n))$. Gabow & Tarjan (1991) showed a running time of $O(m \log(nW) \sqrt{n\alpha(m, n) \log n})$ where $W := \max_{e \in E(G)} |c(e)|$.

Theorem 1.61. *Let G be a graph. The set of vectors $x \in \mathbb{R}^{E(G)}$ satisfying*

$$\begin{aligned} x_e &\geq 0 & e &\in E(G) \\ x(\delta(v)) &= 1 & v &\in V(G) \\ x(\delta(A)) &\geq 1 & A \subseteq V(G) \text{ with } |A| \text{ odd} \end{aligned}$$

is the convex hull of all perfect matchings in G . It is called the perfect matching polytope.

Proof. For any objective function $c : E(G) \rightarrow \mathbb{R}$, the minimum-weight perfect matching algorithm produces an integral primal and a dual solution that satisfy complementary slackness. In particular, all vertices of the polytope are integral. \square

Theorem 1.62. *Let G be a graph. The set of vectors $x \in \mathbb{R}^{E(G)}$ satisfying*

$$\begin{aligned} x_e &\geq 0 & e &\in E(G) \\ x(\delta(v)) &\leq 1 & v &\in V(G) \\ x(E(G[A])) &\leq \frac{|A| - 1}{2} & A \subseteq V(G) \text{ with } |A| \text{ odd} \end{aligned}$$

is the convex hull of all matchings in G . It is called the matching polytope.

Proof. Any matching solution x satisfies these conditions. Let x be any solution that satisfies the conditions. We have to show that x is a convex combination of matching solutions. Define H by:

$$\begin{aligned} V(H) &:= \{(v, i) \mid v \in V(G), i \in \{1, 2\}\} \\ E(H) &:= \{ \{(v, i), (w, i)\} \mid \{v, w\} \in E(G), i \in \{1, 2\} \} \\ &\quad \cup \{ \{(v, 1), (v, 2)\} \mid v \in V(G) \} \end{aligned}$$

We set $y_{\{(v, i), (w, i)\}} := x_{\{v, w\}}$ for all $\{v, w\} \in E(G), i \in \{1, 2\}$ and $y_{\{(v, 1), (v, 2)\}} := 1 - x(\delta(v))$ for all $v \in V(G)$. Then $y \geq 0$ and $y(\delta_H(x)) = 1$ for all $x \in V(H)$.

Claim. *y satisfies the inequalities of the perfect matching polytope (in particular the blossom inequalities).*

If this is true, by 1.62 y is a convex combination of perfect matchings. $H[\{(v, 1) \mid v \in V(G)\}]$ is isomorphic to G , so x is a convex combination of matchings in G .

We now prove the claim: Let $X \subseteq V(G)$ with $|X|$ odd. We have to show that $y(\delta_H(X)) \geq 1$. Define:

$$\begin{aligned} A &:= \{v \in V(G) \mid (v, 1) \in X, (v, 2) \notin X\} \\ B &:= \{v \in V(G) \mid (v, 1) \in X, (v, 2) \in X\} \\ C &:= \{v \in V(G) \mid (v, 1) \notin X, (v, 2) \in X\} \end{aligned}$$

Define $A_i := A \cap (V(G) \times \{i\})$ and $B_i := B \cap (V(G) \times \{i\})$. $|B_1 \cup B_2|$ is even, so (since $|X|$ is odd) $|A|$ or $|C|$ is odd. Without loss of generality, let $|A|$ be odd.

$$\begin{aligned} \sum_{e \in \delta_H(X)} y_e &\geq \sum_{v \in A_1} \underbrace{\sum_{e \in \delta_H(v)} y_e}_{=1} - 2 \cdot \sum_{e \in E(H[A_1])} y_e - \sum_{e \in \delta(A_1) \cap \delta(B_1)} y_e \\ &\quad + \sum_{e \in \delta(A_2) \cap \delta(B_2)} y_e \\ &= |A_1| - 2 \cdot \sum_{e \in E(G[A])} x_e \\ &\geq |A_1| - (|A| - 1) \\ &= 1 \end{aligned}$$

□

Theorem 1.63. *The matching polyhedron is TDI (Totally Dual Integral), i.e. for all $c \in \mathbb{Z}^{E(G)}$ for which the dual program of $(\max c^t x \text{ s.t. } \dots)$ has a finite optimum solution, it has an integral optimum solution.*

Proof. The dual is

$$\begin{aligned} \min \quad & \sum_{v \in V(G)} y_v + \sum_{A \in \mathcal{A}, |A| > 1} \frac{|A| - 1}{2} z_A \\ \text{s.t.} \quad & \sum_{v \in e} y_v + \sum_{A \in \mathcal{A}, |A| > 1, e \in E(G[A])} z_A \geq c(e) \quad e \in E(G) \\ & y, z \geq 0 \end{aligned}$$

Let (G, c) be a counterexample such that $|V(G)| + |E(G)| + \sum_{e \in E(G)} |c(e)|$ is minimum. Then:

- $c(e) \geq 1$ for all $e \in E(G)$, since otherwise e could be deleted.
- G has no isolated vertices.

Claim. *In an optimum solution (y, z) , $y = 0$.*

Proof. If $y_v > 0$, then $x(\delta(v)) = 1$ for all optimum solutions x . Decreasing $c(e)$ by 1 for all $e \in \delta(v)$ yields a smaller feasible instance (G, c') where the weight of x is decreased by 1 and x remains optimum. By assumption, (G, c') is not a counterexample, so there exists an integral optimum solution (y', z') . Increasing y'_v by one yields some optimum in (G, c) which has optimum integral solution $(y' + \mathbb{1}_v, z')$. \square

Let $(y = 0, z)$ be a dual optimum solution such that

$$\sum_{A \in \mathcal{A}, |A| > 1} |A|^2 z_A$$

is maximum.

Claim. $\mathcal{F} := \{A : z_A > 0\}$ is laminar.

If not, there exist $X, Y \in \mathcal{F} : X \cap Y, X \setminus Y, Y \setminus X \neq \emptyset$. We proceed by "uncrossing". Let $\epsilon := \{z_X, z_Y\} > 0$.

Case 1: $|X \cap Y|$ is odd. Then $|X \cup Y|$ is odd. Define:

$$\begin{aligned} z'_X &:= z_X - \epsilon \\ z'_Y &:= z_Y - \epsilon \\ z'_{X \cap Y} &:= z_{X \cap Y} + \epsilon & (\text{unless } |X \cap Y| = 1) \\ z'_{X \cup Y} &:= z_{X \cup Y} + \epsilon \\ z'_A &:= z_A & \text{elsewhere} \end{aligned}$$

Then (y, z') is a dual optimum solution.

Case 2: $|X \cap Y|$ is even. Then $|X \setminus Y|$ and $|Y \setminus X|$ are odd. Define:

$$\begin{aligned} z'_X &:= z_X - \epsilon \\ z'_Y &:= z_Y - \epsilon \\ z'_{X \setminus Y} &:= z_{X \setminus Y} + \epsilon & \text{unless } |X \setminus Y| = 1 \\ z'_{Y \setminus X} &:= z_{Y \setminus X} + \epsilon & \text{unless } |Y \setminus X| = 1 \\ z'_A &:= z_A & \text{elsewhere} \\ y'_v &:= \epsilon & \forall v \in X \cap Y \\ y'_v &:= 0 & \forall v \notin X \cap Y \end{aligned}$$

Then (y', z') is feasible. The objective value is:

$$\begin{aligned}
& \sum_{v \in V(G)} y'_v + \sum_{A \in \mathcal{A}, |A| > 1} z'_A \frac{|A| - 1}{2} \\
&= \epsilon \cdot |X \cap Y| + \sum_{A \in \mathcal{A}, |A| > 1} \frac{|A| - 1}{2} \\
&+ \epsilon \left(\frac{|X \setminus Y| - 1}{2} + \frac{|Y \setminus X| - 1}{2} - \frac{|X| - 1}{2} - \frac{|Y| - 1}{2} \right) \\
&= \text{objective}(y, z)
\end{aligned}$$

Therefore (y', z') is an optimum solution with $y' \neq 0$, which is a contradiction to the previous claim.

We can conclude that \mathcal{F} is laminar.

Let $A \in \mathcal{F}$ with $z_A \notin \mathbb{Z}$ and $|A|$ is maximal. Define $\epsilon := z_A - \lfloor z_A \rfloor > 0$. Let A_1, \dots, A_k be the inclusion-wise maximal proper subsets of A in \mathcal{F} . Since \mathcal{F} is laminar, $A_i \cap A_j = \emptyset$ for $i \neq j$. Define:

$$\begin{aligned}
z'_A &:= z_A - \epsilon \\
z'_{A_i} &:= z_A + \epsilon & 1 \leq i \leq k \\
z'_D &:= z_D & \text{elsewhere}
\end{aligned}$$

Then (y, z') is dual feasible with objective value:

$$\sum_{B \in \mathcal{A}, |B| > 1} \frac{|B| - 1}{2} z'_B < \sum_{B \in \mathcal{A}, |B| > 1} \frac{|B| - 1}{2} z_B$$

This contradicts the optimality of (y, z) , so there exists no counter example. \square

Theorem 1.64. *Let G be a graph.*

$$P := \{x \in \mathbb{R}_{\geq 0}^{E(G)} \mid x(\delta(v)) \leq 1 \quad \forall v \in V(G)\}$$

is the functional matching polytope.

$$Q := \{x \in \mathbb{R}_{\geq 0}^{E(G)} \mid x(\delta(v)) = 1 \quad \forall v \in V(G)\}$$

If G is bipartite, then P and Q are integral.

Proof. The adjacency matrices of bipartite graphs are totally unimodular. \square

Theorem 1.65. *Let G be a graph. The vertices of the fractional perfect matching polytope satisfy*

$$x_e = \begin{cases} \frac{1}{2} & \text{if } e \in E(C_1) \cup \dots \cup E(C_k) \\ 1 & \text{if } e \in M \\ 0 & \text{else} \end{cases}$$

where C_1, \dots, C_k are vertex-disjoint odd circuits and M is a perfect matching in $G - (V(C_1) \cup \dots \cup V(C_k))$.

Proof. Exercise 6.3 □

2 T -Joins and b -Matchings

Definition 2.1. Let G be a graph, $T \subseteq V(G)$. A subset $J \subseteq E(G)$ is called T -join if T is the set of odd-degree vertices in $(V(G), J)$.

Proposition 2.2. *Let G be a graph, $T, T' \subseteq V(G)$, J a T -join and J' a T' -join. Then $J \Delta J'$ is a $T \Delta T'$ -join.*

Proof. For $v \in V(G)$:

$$\begin{aligned} |\delta_{J \cap J'}(v)| &\equiv |\delta_J(v)| + |\delta_{J'}(v)| \\ &\equiv |\{v\} \cap T| + |\{v\} \cap T'| \\ &\equiv |\{v\} \cap (T \Delta T')| \pmod{2} \end{aligned}$$

□

Proposition 2.3. *Let G be a graph, $T \subseteq V(G)$.*

$$\exists \text{ } T\text{-join in } G \Leftrightarrow |V(C) \cap T| \text{ for each connected component } C$$

Proof.

" \Rightarrow ": Let J be a T -join. For each connected component C :

$$\sum_{v \in V(C)} |J \cap \delta(v)| = 2 |J \cap E(C)|$$

Therefore $|J \cap \delta(v)|$ is odd for an even number of vertices and $|V(C) \cap T|$ is even.

" \Leftarrow ": Partition T into pairs $\{v_1, w_1\}, \dots, \{v_k, w_k\}$ such that v_i and w_i are in the same component for all i . Let P_i be a v_i - w_i -path in G . Define $J := E(P_1) \Delta E(P_2) \Delta \dots \Delta E(P_k)$. By proposition 2.2, this is a T -join.

□

Theorem 2.4. *Let G be a graph, $c : E(G) \rightarrow \mathbb{R}$ and $T \subseteq V(G)$. In strongly polynomial time (e.g. $O(n^2m)$) we can determine if a T -join exists and if so, compute a minimum-weight T -join.*

Proof. In $O(m)$ ($m := |E(G)|$), we can check if a T -join exists. If so:

1. Eliminate negative weights.

$$\begin{aligned} N &:= \{e \in E(G) \mid c(e) < 0\} \\ U &:= \{v \in V(G) \mid |\delta_N(v)| \text{ odd}\} \\ T' &:= T \Delta U \\ c'(e) &:= |c(e)| \quad e \in E(G) \end{aligned}$$

Claim. *If J' is a minimum T' -join with respect to c' , then $J' \Delta N$ is a minimum T -join with respect to c .*

Let \tilde{J} be a T -join. Then $\tilde{J} \Delta N$ is a T' -join, so $c'(\tilde{J}) \leq c'(\tilde{J} \Delta N)$ and

$$c(J) = c'(J') + c(N) \leq c'(\tilde{J} \Delta N) + c(N) = c(\tilde{J})$$

which proves the claim.

2. We can now assume that $c \geq 0$. A minimum-weight T -join does not have cycles of positive weight. We can eliminate cycles of weight 0 without changing the cost. We can then restrict ourselves to searching for collections of T - T -paths.

Let K_T be the metric closure of T with respect to G . It can be computed in $O(n \cdot (m + n \log n))$ by using Dijkstra for all vertices. Find a minimum-weight perfect matching M in K_T . Each $e = \{s, t\} \in M$ induces a path $P_{s,t}$. Then the symmetric difference $\Delta_{\{s,t\} \in M} E(P_{s,t})$ is a minimum-weight T -join in G .

□

Corollary 2.6. *A maximum-weight T -join can be computed as fast as a minimum-weight T -join.*

Proof. Set $c' := -c$.

□

Corollary 2.7. *Let G be a graph, $c : E(G) \rightarrow \mathbb{R}$. We can find a cycle of negative length in G in $O(n^2m)$ time.*

Proof. Apply theorem 2.4 to $T = \emptyset$. If $c(J) < 0$, $(V(G), J)$ contains a cycle C . If $c(C) = 0$, we can eliminate it and recurse, otherwise return C .

□

2.2 T -Join Applications

2.2.1 TSP Approximation

Let (K_n, c) with c metric be an instance of the TSP. Consider the *Double tree algorithm*:

1. Compute a minimum spanning tree T .
2. $T' := T + T$ (doubling all edges). Then T' is Eulerian.
3. Walk along T' and add vertices to the TSP tour in the order of their first appearance, yielding a tour T^* . Since c is metric, we have $c^*(*) \leq c(T') \leq 2c(T)$. Since the cost of T is a lower bound for the cost of a tour, we have $c(T^*) \leq 2\text{OPT}$ (where OPT is the cost of a shortest TSP tour).

Algorithm 5: Christofides Algorithm (1976)

Input: Complete metric graph (K_n, c)

Output: A TSP-tour T

- 1 Find MST T_{MST} in (K_n, c)
 - 2 $W := \{v \in V(K_n) \mid |\delta_{T_{\text{MST}}}(v)| \text{ odd}\}$
 - 3 $J :=$ minimum-weight W -Join in (K_n, c)
 - 4 Add cities to T in the order of first appearance in a Eulerian walk of $T_{\text{MST}} + J$.
 - 5 **return** T
-

Theorem 2.8. *Algorithm 5 is a $\frac{3}{2}$ -approximation algorithm for the metric TSP, i.e. for the computed tour T we have:*

$$c(T) \leq \frac{3}{2}\text{OPT}$$

Proof. We have $c(T_{\text{MST}}) \leq \text{OPT}$ and $\text{OPT}(W) \leq \text{OPT}(V(K_n))$ (since c is metric). Any tour through the vertices in W can be decomposed into 2 matchings. Therefore, $c(J) \leq \frac{1}{2}\text{OPT}(W) \leq \frac{1}{2}\text{OPT}$. It follows that $c(T) \leq (1 + \frac{1}{2})\text{OPT}$. \square

2.2.2 Shortest Paths in Undirected Graphs

The naive reduction to digraphs requires non-negative weights.

Corollary 2.9. *Given an undirected graph G , $c : E(G) \rightarrow \mathbb{R}$ such that each circuit has length at least 0. Then for $s, t \in V(G)$, a shortest s - t -path can be found in $O(n^2m)$ time, where $n := |V(G)|$, $m := |E(G)|$.*

Proof. Choose $T := \{s, t\}$. Apply theorem 2.4 to get a minimum-weight T -join J . J can be partitioned into circuits of length 0 and an s - t -path of length $c(J)$. \square

2.2.3 Chinese Postman Problem

Definition 2.10. A walk $C = \{v_0, e_1, v_1, \dots, e_t, v_t\}$ is called a Chinese postman tour if $v_0 = v_t$ and each edge in $E(G)$ is visited at least once. The Chinese Postman Problem is the problem of finding a shortest Chinese postman tour in G with respect to $c : E(G) \rightarrow \mathbb{R}_{\geq 0}$.

Corollary 2.11. *The Chinese postman problem can be solved in $O(n^2m)$ time, where $n := |V(G)|$, $m := |E(G)|$.*

Proof. Set $T := \{v \in V(G) \mid |\delta(v)| \text{ odd}\}$ and let J be a minimum-weight T -join. Compute a Eulerian tour C in $G + J$. Let C' be a shortest Chinese postman tour. Let $J' :=$ set of edges occuring in C' an even number of times (at least twice). Then J' is a T -join, so $c(J') \geq c(J)$ and:

$$c(C') \geq c(E(G)) + c(J') \geq c(E(G)) + c(J) = c(C)$$

□

2.3 T -Joins and T -Cuts

Definition 2.12. Let G be a graph and $T \subseteq V(G)$. A T -cut is a cut $C = \delta(X)$ with $X \subseteq V(G)$ and $|X \cap T|$ odd.

Proposition 2.13. *Let G be a graph, $T \subseteq V(G)$, $|T|$ even. Then:*

1. *For any T -join J and any T -cut C : $J \cap C \neq \emptyset$.*
2. *The inclusion-wise minimal T -cuts (T -joins) are exactly the inclusion-wise minimal edge sets intersecting all T -joins (all T -cuts).*

Proof. For 1., let $C = \delta(X)$ with $|X \cap T|$ odd be a T -cut. Then the edges in $J \cap C$ either belong to a path passing through X or have an endpoint in T . Therefore $|J \cap C|$ is odd, in particular the set is non-empty.

For 2., we prove in an exercise that each edge set intersecting all T -joins (T -cuts) contains a T -cut (T -join). Therefore minimal such sets are T -cuts (T -joins). Remark: The minimum cardinality of a T -join is at least as large as the maximum number of edge-disjoint T -cuts⁴. □

Theorem 2.14 (Seymour (1981)). *Let G be bipartite, $T \subseteq V(G)$ such that there exists a T -join. Then:*

$$\min. \text{ cardinality of a } T\text{-join} = \max. \text{ number of edge-disjoint } T\text{-cuts}$$

The maximum is attained by a crossfree family \mathcal{C} of cuts, i.e.

$$\forall X, Y \in \mathcal{C} : X \subseteq Y \vee Y \subseteq X \vee X \cap Y = \emptyset \vee X \cup Y = V(G)$$

⁴In general, the two numbers are not equal: Consider K_4 and $T = V(K_4)$. A minimum T -join consists of 2 edges but there are no 2 edge-disjoint T -cuts.

Proof. If $T = \emptyset$, the statement is clear. Let $T \neq \emptyset$. We proceed by induction on $|V(G)| + |T|$. Let J be a minimum-cardinality T -join. Set:

$$c(e) := \begin{cases} -1 & e \in J \\ 1 & e \in E(G) \setminus J \end{cases}$$

Claim. *Every circuit C has $c(C) \geq 0$.*

$$\begin{aligned} c(C) &= c(C \setminus J) + c(C \cap J) + |J \setminus C| - |J \cap C| \\ &= \left| \underbrace{C \Delta J}_{T\text{-join}} \right| - |J| \geq 0 \end{aligned}$$

Let P be a minimum length walk in (G, c) traversing no edge more than once such that $|E(P)|$ is minimum. Then P is a path. Let t be the last vertex in P and f the edge entering t . Then $f \in J$, otherwise $c(f) = 1$ and deleting f would yield a shorter path. Furthermore, $|\delta_J(t)| = 1$, otherwise we could add the other edge from $J \cap \delta(t)$ to shorten $c(P)$.

Claim. *Each circuit C that contains t but not f has $c(C) > 0$.*

- Case 1: t is the only vertex in $V(C) \cap V(P)$. Let $e \ni t$ be an edge on C incident to t . Then $c(e) = 1$ (since $\delta_J(t) = \{f\}$) and $P' := P + C - e$ yields a shorter walk if $c(C) \leq 0$.
- Case 2: $V(C) \cap V(P)$ contains another vertex x . Let u be the last vertex on P before t that is also on C . Define $P' := P_{[u, t]}$. C can be split into 2 u - t -paths C', C'' . By minimality of P , $c(P') < 0$. $P' + C', P' + C''$ are circuits (by choice of u). By the first claim, $c(C'), c(C'') > 0$, so also $c(C) > 0$.

□