

Constrained Matching Problem in Bipartite Graphs

Monaldo Mastrolilli Georgios Stamoulis

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George Zirdelis, NTUA

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Definition 1

Given a graph $G = (V, E)$, a matching M in G is a set of pairwise non-adjacent edges; that is, no two edges share a common vertex.

Definition 2 (Bounded Color Matching)

Input is:

- Bipartite graph $G(V, E)$ with bipartition $V = V_1 \cup V_2$.
- The edge set E is partitioned into k sets, $E_1 \cup E_2 \cup \dots \cup E_k$.
- Each edge set is characterized by a color $j \in [k]$.
- Each edge $e \in E$ has a profit $p_e \in \mathbb{Q}^+$.

Objective is:

- Find a maximum weight matching M .
- In M there are no more than w_j edges of color j where $w_j \in \mathbb{Z}^+$, i.e.
 $M \cap E_j \leq w_j, \quad \forall j \in [k]$.

The relaxation of the IP for the Bounded Color Matching problem:

$$\begin{aligned} & \text{maximize} && \max p^T x \\ & \text{subject to} && \sum_{e \in \delta(v)} x_e \leq 1, \quad \forall v \in V \\ & && \sum_{e \in E_j} x_e \leq w_j, \quad \forall j \in [k] \\ & && 0 \leq x_e \leq 1 \end{aligned}$$

where $\delta(v)$ is the set of edges with one endpoints in v . Integrality Gap is $\frac{1}{2}$ so we cannot hope to achieve a better than $\frac{1}{2}$ approximation algorithm.

Another way to describe the problem, say \mathcal{M}' , is the following:

$$\mathcal{M}' = \left\{ y \in \{0, 1\}^{|E|} : y \in \mathcal{M} \wedge \sum_{e \in E_j} y_e \leq w_j, \forall j \in [k] \right\}$$

where \mathcal{M} is the usual bipartite polytope. Again, we can relax this by setting $y_e \in [0, 1]$.

- The Bounded Color Matching problem is known to be **NP** – *Complete*
- Even if $|E_j| \leq 2$ and $w_j = 1, \forall j$.
- The special case of a 2-regular bipartite graphs where,
 - 1 Each color appears twice.
 - 2 Find a maximum matching with at most one edge per coloris **APX** – *Hard* so a **PTAS** is out of reach.

Definition 3

Let $E' \subseteq E$. Then we define the **characteristic vector** of E' to be the binary vector $\chi_{E'} \in \{0, 1\}^{|E|}$, s.t.

$$\chi_{E'}(e) = 1 \Leftrightarrow e \in E'$$

Definition 4

Let $y \in \mathbb{R}^n$. Then,

$$\text{support}(y) = \{i \in [n] : y_i \neq 0\}$$

i.e. the indices of the non-zero components of y .

Lemma 5

Let x^* be an optimal basic feasible solution for the LP described by \mathcal{M}' s.t. $x_e^* > 0$, $\forall e \in E$. Then, there exist $F \subseteq V$ and $Q \subseteq [k]$ s.t.,

- 1 $\sum_{e \in \delta(v)} x_e^* = 1, \quad \forall v \in F.$
- 2 $\sum_{e \in E_j} x_e^* = w_j, \quad \forall j \in Q.$
- 3 $\{\chi_{\delta(v)}\}_{v \in F}$ and $\{\chi_{E_j}\}_{j \in Q}$ are all linearly independent.
- 4 $|E| = |F| + |Q|$ where $|E|$ is the number of the edges with $x_e^* > 0$.

- If $\sum_{e \in \delta(v)} x_e = 1$ then v is a *tight vertex*.
- If $\sum_{e \in E_j} x_e = w_j$ then j is a *tight color class*.

Define the **residual graph** to be the graph with the same vertex set but we include an edge e if $x_e > 0$ in the LP solution for the original graph.

Lemma 6

Take any basic feasible solution x s.t. $x_e > 0, \forall e$, i.e. we remove any edge with $x_e = 0$. Then one of the following must be true:

- 1 either there is an edge s.t. $x_e = 1$.
- 2 or there is a color class $j \in Q \subseteq [k]$ s.t. $|E_j| \leq w_j + 1$ in the residual graph
- 3 or there is a tight vertex $v \in F$ s.t. the degree of v is 2 in the residual graph.

Using the lemma above will can do iterative rounding to obtain a solution.

C, E will be resp. the set of the available colors and edges, at each round.

Initialize $M = \emptyset$.

While $C \neq \emptyset$ or $E \neq \emptyset$ do:

- 1 Compute an optimal (fractional) basic solution x to the current LP.
- 2 Remove all edges from the graph s.t. $x_e = 0$.
- 3 Remove all vertices of the graph s.t. $\deg(v) = 0$.
- 4 if $\exists e = (u, v) \in E : x_e = 1$ and $e \in C_j$ then $M := M \cup \{e\}$, $V = V \setminus \{u, v\}$, $w_j := w_j - 1$. if $w_j = 0$ then $C := C \setminus C_j$, $E := E \setminus \{e : e \in E_j\}$.
- 5 **(Relaxation:)** while $V \cup C \neq \emptyset$
 - if \exists color class $C_j \in Q$ with $|E_j| \leq w_j + 1$ then remove the constraint for this color class, i.e. define $C := C \setminus C_j$.
 - if \exists vertex $v \in F$ s.t. $\deg(v) = 2$ then remove the constraint for that vertex.

Return M

At each step of the algorithm, either we add an edge to our matching M , or we remove a tight constraint. Thus the algorithm will terminate in at most $|Q| + |F|$.

- Since we remove the degree constraints for a vertex v when $\text{deg}(v) = 2$ we select edges from a graph G' that is a collection of disjoint paths or cycles.
- But a disjoint path or cycle can be partitioned into two matchings, i.e. M_1, M_2 and we select the one with the highest profit, i.e.

$$\max(p(M_1), p(M_2)) \geq \frac{1}{2}p(M_1 \cup M_2)$$

- Therefore we do this for every connected component (disjoint paths and cycles), we get at least $\frac{1}{2}$ of the profit of the matchings but we violate by an additive 1 every color constraint.

As a result of the above there is a polynomial time $(1/2, \text{additive } 1)$ bi-criteria approximation algorithm for the weighted Bounded Color Matching problem.

We now consider the unweighted version of the Bounded Color Matching problem:
Compute a maximum cardinality matching M s.t. in M we have at most w_j edges for color class j .

Recall that from Lemma 7 we have that for any solution to the LP, if $0 < x_e < 1$ then,

- either there exists a tight color class $j \in Q$ s.t. $|\text{support}(x) \cap E_j| \leq w_j + 1$
- or there exists a tight vertex $v \in F$ s.t. $\text{deg}(v) = 2$.

The main idea of the algorithm for the cardinality version consists of the two following steps:

- **Relaxation step:** We identify a tight color class j and we remove its constraint, thus relaxing the problem.
- **Rounding step**
 - We round appropriately some variables to 1 and some others to 0, preserving feasibility.
 - Rounding step comes with a parameter $\lambda \in [0, 1]$. Idea is that if we round x_e to 1, we need to update the color bound of this color class.
 - Using λ we update the color bound by any value in $[x_e, 1]$ (if we use $x_e + \lambda(1 - x_e)$).
 - Values of λ closer to x_e violate more the color constraint whereas values closer to 1 give less violation but worst performance guarantee.

Lemma 7

Let x be the optimal solution in G (as stated in Lemma 4) before the rounding step and \hat{x} be the optimal solution after the rounding step in \hat{G} . Then we have that,

$$\sum_{e \in E(G)} x_e - \sum_{e \in E(\hat{G})} \hat{x}_e \leq 1 + (\gamma + \lambda\gamma)$$

where $\gamma = 1 - x_e$.

The loss due to a single rounding step is at most $\gamma + \lambda\gamma$ which can be at most $\frac{1}{2}(\lambda + 1)$.

C, E will be resp. the set of the available colors and edges, at each round.

Initialize $M = \emptyset$.

While $C \neq \emptyset$ or $E \neq \emptyset$ do:

- 1 Compute an optimal (fractional) basic solution x to the current LP.
- 2 Remove all edges from the graph s.t. $x_e = 0$.
- 3 Remove all vertices of the graph s.t. $\text{deg}(v) = 0$.
- 4 if $\exists e = (u, v) \in E : x_e = 1$ and $e \in C_j$ then $M := M \cup \{e\}$, $V = V \setminus \{u, v\}$, $w_j := w_j - 1$. if $w_j = 0$ then $C := C \setminus C_j$, $E := E \setminus \{e : e \in E_j\}$.
- 5 **(Relaxation:)** If \exists color class $j \in Q$ with $|E| \leq \lceil w_j \rceil + 1$ then remove the constraint for this color class, i.e. set $C := C \setminus C_j$ and iterate.
- 6 **(Rounding:)** if $\exists v \in F$ s.t. $\text{deg}(v) = 2$ then let: u_1, u_2 be the neighbors of v and let e_1, e_2 be the two edges incident on v . Assume w.l.o.g. that $x_{e_1} \geq \frac{1}{2}$ and $e_1 = (u_1, v)$.
 - Round x_{e_1} to 1. Add it (e_1) to M .
 - Round x_{e_2} and all other edges incident to u_1 to zero.
 - If $e_1 \in E_j$ then set $w_j := w_j - x_{e_1} - \lambda(1 - x_{e_1})$.
 - Remove v, u_1 and all the rounded edges from the graph and iterate.

Return M

- From Lemma 7 we have that in each rounding step the objective function decreases by $1 + \gamma + \lambda\gamma$.
- Intuitively, the larger the value of γ is, the fewer iterations the algorithm will perform.
- Because $OPT \leq |V|/2$ and at each rounding step we delete 2 vertices from the current graph, we can perform at most $|V|/4$ rounding steps. So, we can have at most $|V|/4$ values of γ , though they all might be different.

Lemma 8

Let \tilde{x} be the final solution of the algorithm that corresponds to M . Then we have that,

$$\sum_{e \in M} \tilde{x}_e \geq \frac{2}{3 + \lambda} \sum_{e \in E(G)} x_e$$

Proof.

- Since we choose $x_{e_1} \geq 1/2$ assume that in some iteration $\gamma_1 = \frac{p}{q} \in (0, 1/2]$ and also that this γ_1 appears k_1 times during the Rounding steps.
- The total decrease in the objective function is $\frac{q+p(\lambda+1)}{q} = 1 + \gamma_1 + \lambda\gamma_1$
- Maximum number of iterations we can have for this particular γ_1 is $OPT \cdot \frac{q+p(\lambda+1)}{q}$ before it truncates to 0. E.g. for $\gamma_1 = \frac{1}{3}$ and $\lambda = \frac{1}{2}$ in the next iteration of the LP we will have

$$OPT' = OPT - \frac{3}{2} \Rightarrow OPT - OPT' = \frac{3}{2}$$

and so we can have at most $OPT \cdot \frac{2}{3}$ iterations. □

continued.

- Assume that the algorithm performs a fraction f_i of the maximum possible number of iterations for each γ_i . Then we have that,

$$\sum_i f_i \leq 1$$

because in each round we reduce the objective function.

- At the end of the algorithm the final objective function value will be,

$$OPT - \sum_i f_i \cdot \frac{OPT}{1 + \gamma_i + \lambda \gamma_i} \cdot \gamma_i (\lambda + 1)$$



continued.

Set $g(\gamma_i) = \frac{\gamma_i(\lambda+1)}{1+\gamma_i+\lambda\gamma_i}$ which monotonically increases. We have that

$$\begin{aligned} SOL &= OPT - OPT \sum_i f_i \cdot g(\gamma_i) \geq OPT - OPT \sum_i f_i \cdot g(1/2) \\ &= OPT - OPT \sum_i \frac{\lambda+1}{\lambda+3} \\ &\geq \frac{2}{\lambda+3} OPT \end{aligned}$$

Using similar arguments one can show that the color bound w_j of a color j can be violated by at most a factor of $\frac{2}{1+\lambda} w_j + 1$. □

Theorem 9

For any $\lambda \in [0, 1]$, there is a polynomial time $(\frac{2}{3+\lambda}, \frac{2}{1+\lambda} w_j + 1)$ bi-criteria approximation algorithm for the Bounded Color Matching problem.

- The closer λ is to 1 the more we deteriorate from the optimal objective function value but the less we lose in color bounds.
- The closer λ is to 0 the more we violate the color constraints but the better the approximation guarantee is.
- Depending on the application we choose a parameter λ that is more suitable.
- We have a family of algorithms for the unweighted case.



Monaldo Mastrolilli and Georgios Stamoulis.

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