CS217 – Algorithm Design and Analysis Homework 6

Not Strong Enough

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 Γ_1

Let $\nu(G)$ denote the size of a maximum matching of G. Obviously, $\operatorname{val}(\operatorname{MLP}(G)) \geq \nu(G)$ for all graphs. Show that $\nu(G) = \operatorname{val}(\operatorname{MLP}(G))$ for all bipartite graphs G. Do this without referring to Kőnig's Theorem.

Solution. Suppose \mathbf{x} is a solution to MLP(G). E_F is the set of edges with fractional values in \mathbf{x} . First do as follows:

- (1) If E_F doesn't contain a cycle, terminate.
- (2) Find a cycle in E_F with fractional values, denote it $C := \{e_i \in E_F : 1 \le i \le n, i \in \mathbb{N}\}.$
- (3) Add $x_{e_{2k-1}}$ by ϵ and $x_{e_{2k}}$ by $-\epsilon$ for $\{k \in \mathbb{N} : k \le n/2\}$.
- (4) Increase ϵ until there exists i such that $x_{e_i} = 0$ or 1.
- (5) go to step(2) if there exists a cycle in E_F .

Since G is a bipartite, C can only be a even cycle, so step (3) makes sense. If we modify the solution by step(3), obviously the constraints of MLP(G) will still be satisfied, and the target function will remain unchanged. The process will terminate because $|E_F|$ decreases by at least 1 in each iteration.

Then do as follows:

- (1) If E_F is empty, terminate.
- (2) Choose 2 vertex v_1, v_2 that $|\{x_e \in (0,1) : v_1 \in e\}| = 1$, $|\{x_e \in (0,1) : v_2 \in e\}| = 1$ and there's a path between v_1 and v_2 in E_F , denote it $P := \{e_i' \in E_F : 1 \le i \le m, i \in \mathbb{N}\}$
- $(3) \ \mathrm{Add} \ x_{e'_{2k-1}} \ \mathrm{by} \ \epsilon \ \mathrm{for} \ \{k \in \mathbb{N} : 2k-1 \leq m\} \ \mathrm{and} \ x_{e'_{2k}} \ \mathrm{by} \ -\epsilon \ \mathrm{for} \ \{k \in \mathbb{N} : 2k \leq m\}$
- (4) Increase ϵ until there exists i such that $x_{e'_i} = 0$ or 1
- (5) Go to step (2) if E_F is not empty

Since there's no cycle in E_F , we can definitely choose v_1, v_2 that satisfy the requirements if E_F is not empty. In each iteration, the target function will not decrease. Since v_1, v_2 can only be touched by edges with fractional value or value 0, the constraints of v_1, v_2 can be satisfied. In other words, $\sum_{e \in E: v_1 \in e} x_e = x_{e'_1} \le 1$, $\sum_{e \in E: v_2 \in e} x_e = x_{e'_m} \le 1$ The constraints of other vertices v in P will obviously be maintained. The process will terminate because $|E_F|$ decreases by at least 1 in each iteration.

Finally, after 2 processes, the solution becomes integral and the value of target function does not decrease, which means $\operatorname{int-val}(MLP(G)) \geq \operatorname{val}(MLP(G))$. And we already have $\operatorname{int-val}(MLP(G)) \leq \operatorname{val}(MLP(G))$. So $\nu(G) = \operatorname{int-val}(MLP(G)) = \operatorname{val}(MLP(G))$

Γ_2

We know that $\nu(G) = \tau(G)$ for all bipartite graphs (Kőnig's Theorem) and $\nu(G) \leq \tau(G)$ for all graphs (since every matched edge must be covered by a distinct vertex). Show that $\tau(G) \leq 2\nu(G)$ for all graphs G.

Proof. Let M be a maximum matching of G. It follows that $|M| = \nu(G)$. Now we choose our vertex set V' to be all the matched vertices in G. So $|V'| = 2\nu(G)$.

Claim that V' is a vertex cover. To see that, assume there exists an edge (u, v) which is not covered by V'. It means that neither u nor v is matched. So we can add edge (u, v) to M, and thus M is not maximum, which leads to a contradiction.

So the size of minimum vertex cover $\tau(G) \leq |V'| = 2\nu(G)$.

Γ_3

Show that $\tau(G) \leq 2 \operatorname{opt}(\operatorname{VCLP}(G))$ for all graphs G (including non-bipartite graphs).

Proof. From (2) we know that $\tau(G) \leq 2\nu(G)$. Since $\nu(G) \leq \operatorname{opt}(\operatorname{MLP}(G)) \leq \operatorname{opt}(\operatorname{VCLP}(G))$, it follows that $\tau(G) \leq 2\nu(G) \leq 2\operatorname{opt}(\operatorname{VCLP}(G))$.

 Γ_4

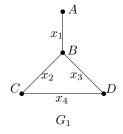
For a graph G = (V, E), let $\tau(G)$ denote the size of a minimum vertex cover, and $\nu(G)$ the size of a maximum matching. Recall the two linear programs VCLP and MLP. Let $\tau_f(G) := \operatorname{opt}(\operatorname{VCLP}(G))$ and $\nu_f(G) := \operatorname{opt}(\operatorname{MLP}(G))$. Note that

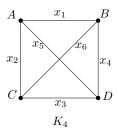
$$\nu(G) \le \nu_f(G) = \tau_f(G) \le \tau(G),$$

where the equality in the middle follows from Strong LP Duality. Also, if G is bipartite, then the equality holds throughout in (1). Let us say a graph G is VCLP exact if $\tau(G) = \tau_f(G)$, and MLP exact if $\nu(G) = \nu_f(G)$. As we already know, a bipartite graph G is both VCLP exact and MLP exact.

From now on, suppose that G is not bipartite but $\tau(G) = \tau_f(G)$.

- 1. Give an example of such a graph G that is not bipartite but still VCLP exact.
- 2. Give an example of a graph G that is MLP exact but not VCLP exact.
- 3. Suppose G is VCLP exact. Let $Y \subseteq V(G)$ be a minimum vertex cover. Let \mathbf{x} be an optimal solution of MLP(G). Show that $x_e = 0$ if $e \subseteq Y$ (i.e., if both endpoints of e are in the cover).
- 4. Show that such a graph G has a matching of size |Y|, and thus is MLP exact, too.





Solution to (1) and (2).

1. The example is G_1 in the upper left. It is not bipartite since B, C, D constitute an odd cycle. The minimum vertex cover of G_1 can be $\{B, C\}$ or $\{B, D\}$, with size 2. So $\tau(G_1) = 2$. VCLP (G_1) is

$$\begin{array}{ll} \text{minimize} & y_A+y_B+y_C+y_D\\ \text{subject to} & y_A+y_B\geq 1\\ & y_B+y_C\geq 1\\ & y_B+y_D\geq 1\\ & y_C+y_D\geq 1\\ & y_A,y_B,y_C,y_D\geq 0 \end{array}$$

Note that if we add up the constraints $y_A + y_B \ge 1$ and $y_C + y_D \ge 1$, we get $y_A + y_B + y_C + y_D \ge 2$, which gives a lower bound of the target function.

Since $\tau(G_1) = 2$, it follows that the lower bound is tight. Hence $\tau_f(G_1) = \tau(G_1) = 2$.

So G_1 is an example which is not bipartite but still VCLP exact.

2. The example is K_4 in the upper right. The minimum vertex cover can be $\{A, B, C\}, \{A, B, D\}, \{A, C, D\}$ and $\{B, C, D\}$, with size 3. So $\tau(K_4) = 3$.

However, by setting the value of each vertex to 0.5, we find that all edges are exactly covered $(y_u + y_v = 1 \text{ for edge } (u, v))$. So $\tau_f(K_4) \leq 2$, and thus K_4 is not VCLP exact.

Now consider the maximum matching and MLP. Obviously we can only match 2 pairs of vertices. So $\nu(K_4) = 2$, and $\nu_f(K_4) \geq 2$ follows. Since we already know that $\tau_f(K_4) \leq 2$ and $\nu_f(K_4) = \tau_f(K_4)$ by Strong LP Duality, we can conclude that $\nu(K_4) = \nu_f(K_4) = \tau_f(K_4) = 2$. Therefore K_4 is MLP exact.

So K_4 is an example which is MLP exact but not VCLP exact.

Suppose that Y is a minimum vertex cover of G. Note that there is no edge $(u, v) \in E$ such that both $u, v \in V \setminus Y$. Because $u, v \in V \setminus Y$ means that (u, v) is not covered by Y, which is contradict with Y being a vertex cover.

To solve (3) and (4), define a function $N(\cdot)$ relative to a VCLP exact graph G = (V, E) and a minimum vertex cover Y that:

$$N(A) = \begin{cases} \{v : v \text{ is the neighbor of some } u \in A\} \cap (V \setminus Y) &, \text{ if } A \subseteq Y \\ \{v : v \text{ is the neighbor of some } u \in A\} \cap Y &, \text{ if } A \subseteq V \setminus Y \end{cases}$$

5: Lemma

If $A \subseteq Y$, then $|A| \leq |N(A)|$. Vise versa, if $A \subseteq V \setminus Y$, then $|A| \geq |N(A)|$.

Proof. If $A \subseteq Y$ and there is |B| < |A| where B = N(A). Consider an optimal solution of VCLP relative to Y(that is, if $v \in Y$, $y_v = 1$, otherwise $y_v = 0$). Let

$$y'_{v} = \begin{cases} y_{v} - \epsilon & , v \in A \\ y_{v} + \epsilon & , v \in B \\ y_{v} & , \text{ otherwise.} \end{cases}$$

where $\epsilon < \frac{1}{2}$. To prove that (y'_v) is also a solution for VCLP, we only need to check that for edge $(u,v) \in E$:

- a) $y'_u + y'_v \ge 1$ for $u, v \in A$. Since there is $y'_u + y'_v = (1 \epsilon) + (1 \epsilon) = 2 2\epsilon > 1$;
- b) $y'_u + y'_v \ge 1$ for $u \in A, v \in Y \setminus A$. Since there is $y'_u + y'_v = (1 \epsilon) + 1 = 2 \epsilon > 1$;
- c) $y'_u + y'_v \ge 1$ for $u \in A, v \in B$. Since there is $y'_u + y'_v = (1 \epsilon) + \epsilon = 1$;
- d) $y'_u + y'_v \ge 1$ for $u \in Y \setminus A, v \in B$. Since there is $y'_u + y'_v = 1 + \epsilon > 1$;
- e) The rest of constraints remains holding since the left hand side is not changed from solution (y_v) .

So $(y_v)'$ is a solution to the VCLP, too. But it is $\epsilon(|A| - |B|)$ less than solution (y_v) , which rises a contradiction that Y is a minimum solution.

Likewise, let A = N(B) and use the same method to construct (y'_v) can prove the "vice versa" part.

Now we begin to prove (3) and (4):

Proof. (3) Consider the optimal solution of MLP be $X = (x_e)$. Assume there is $y_1, y_2 \in Y, (y_1, y_2) \in V, x_{y_1y_2} = x \ge 0$. Let $A_0 = \{y_1, y_2\}, B_0 = N(A_0), C_0 = N(B_0)$.

By the lemma above, there is $|B_0| > |C_0|$. Let all edges from C to B be $\{e_1, e_2 \dots e_m\}$, there is:

$$\sum_{i=1}^{m} y_e + 2x \le |C_0| \le |B_0|$$

By the property of vertex cover, any vertex in B can only connected to vertices in C_0 . Let $B = \{b_1, b_2 \dots b_s\}$, and $v(b_i) = \sum_{\forall e \in E: b_i \in e} x_e$, then there is:

$$\sum_{i=1}^{s} v(b_i) = \sum_{i=1}^{m} y_e \le |B_0| - 2x$$

. So there exists $I = \{v_{i_1}, v_{i_2} \dots v_{i_t}\} \subseteq B$, where $\sum_{j=1}^{t} (1 - v(b_{i_j})) \ge 2x$. We call $1 - v(b_i)$ the capacity of vertex b_i . Let I_1 be all vertices connected to y_1 and in I, and the capacity of I_1 , $c(I_1)$ is the sum of capacity of vertices in I_1 , I_2 likewise. Let $A_1 = \{y_2\}$, $B_1 = N(A_1)$, $C_1 = N(B_1)$, by the same method, we can prove that the $c(B_1)$ is no less than x. And that means the $c(I_2)$ is no less than x, since I_2 contains all those vertices with positive capacity in B_1

Without loss of generality, we can assume that $c(I_1)$ is more than x.

Now we prove that in different cases we can always modify the optimal solution X to get a better solution.

In each case, we do the same thing first: let $x_{y_1y_2} = 0$ (which gives an decreasement of x in the target). We then define a slack of (y_i, I_i) (i = 1, 2; j = 1, 2, 3, 4, 5) be that:

Increase x_e for each edge e from y_i to I_j until the constraint of at least one side of e is strict.

Let
$$I_3 = I_1 \cap I_2, I_4 = I_1 \setminus I_3, I_5 = I_2 \setminus I_3,$$

a) If the $c(I_5) \ge 0$, do a slack to (y_2, I_5) . Since the $c(I_5 \ge 0)$, the increasement is more than 0. Notice that $c(I_1)$ is not changed, so now we do the slack of (y_1, I_1) and the increasement is not less than x. Not the total increasement is more than x, so the new solution is better.

b)If $c(I_5) = 0$, there is $c(I_2) \ge 0 \to c(I_3) \ge 0$ and $c(I_1) = c(I) \ge 2x$. Now we do slack of (y_2, I_3) but stop as soon as the increasement is x. Now there is $c(I_1) \ge 2x - x = x$, so doing a slack of (y_1, I_1) can get an increasement more than x. In all, the increasement is more than x, so the new solution is better.

Since in each case there is a better solution, there is a contradiction (that (x_e) is already the optimal). So the assumption is incorrect and it is proved.

(4) This part is straightly proved by Hall's Theorem and the lemma above. Remove all edges from Y to Y, then the graph is a bipartite graph G with L = Y, $R = V \setminus Y$. By the lemma above, the condition of Hall's Theorem is satisfied. So there is a matching of size |Y|.