

Bounds on the stability number of a graph via the inverse theta function

Miklós Ujvári *

Abstract. In the paper we consider degree, spectral, and semidefinite bounds on the stability number of a graph. The bounds are obtained via reformulations and variants of the inverse theta function, a notion recently introduced by the author in a previous work.

Keywords: stability number, inverse theta number

1 Introduction

In this paper we provide several new descriptions and variants of the inverse theta function, a notion recently introduced by the author (see [10]). We also present some applications in the stable set problem, bounds on the cardinality of a maximum stable set in a graph.

We start the paper with describing sandwich theorems on the inverse theta number and its predecessor, the theta number (see [3]). First we fix some notation. Let $n \in \mathcal{N}$, and let $G = (V(G), E(G))$ be an undirected graph, with vertex set $V(G) = \{1, \dots, n\}$, and with edge set $E(G) \subseteq \{\{i, j\} : i \neq j\}$. Let $A(G)$ be the 0-1 adjacency matrix of the graph G , that is let

$$A(G) := (a_{ij}) \in \{0, 1\}^{n \times n}, \text{ where } a_{ij} := \begin{cases} 0, & \text{if } \{i, j\} \notin E(G), \\ 1, & \text{if } \{i, j\} \in E(G). \end{cases}$$

The complementary graph \overline{G} is the graph with adjacency matrix

$$A(\overline{G}) := J - I - A(G),$$

where I is the identity matrix, and J denotes the matrix with all elements equal to one. The disjoint union of the graphs G_1 and G_2 is the graph $G_1 + G_2$ with adjacency matrix

$$A(G_1 + G_2) := \begin{pmatrix} A(G_1) & 0 \\ 0 & A(G_2) \end{pmatrix}.$$

*H-2600 Vác, Szent János utca 1. HUNGARY

Let $(\delta_1, \dots, \delta_n)$ be the sum of the row vectors of the adjacency matrix $A(G)$. The elements of this vector are the degrees of the vertices of the graph G . We define similarly the values $\bar{\delta}_1, \dots, \bar{\delta}_n$ in the complementary graph \bar{G} instead of G . Let Δ_G (resp. μ_G) be the maximum (resp. the arithmetic mean) of the degrees in the graph G . Note that

$$\mu_{\bar{G}} = n - 1 - \mu_G, \mu_{G_1+G_2} = \frac{n_1\mu_{G_1} + n_2\mu_{G_2}}{n_1 + n_2}. \quad (1)$$

By Rayleigh's theorem (see [6]) for a symmetric matrix $M = M^T \in \mathcal{R}^{n \times n}$ the minimum and maximum eigenvalue, λ_M resp. Λ_M , can be expressed as

$$\lambda_M = \min_{\|u\|=1} u^T M u, \Lambda_M = \max_{\|u\|=1} u^T M u.$$

By the Perron-Frobenius theorem (see [5]) for an elementwise nonnegative symmetric matrix $M = M^T \in \mathcal{R}_+^{n \times n}$ the maximum is attained for a nonnegative unit (eigen)vector: we have $\Lambda_M = u^T M u$ for some $u \in \mathcal{R}_+^n$, $u^T u = 1$. Furthermore, if $M = M^T \in \mathcal{R}_+^{n \times n}$, then $-\lambda_M \leq \Lambda_M$.

The maximum (resp. minimum) eigenvalue of the adjacency matrix $A(G)$ is denoted by Λ_G (resp. λ_G). By Exercise 11.14 in [4], we have

$$\mu_G, \sqrt{\Delta_G} \leq \Lambda_G \leq \Delta_G, \sqrt{\mu_G(n-1)}. \quad (2)$$

The set of the n by n real symmetric positive semidefinite matrices will be denoted by \mathcal{S}_+^n , that is

$$\mathcal{S}_+^n := \{M \in \mathcal{R}^{n \times n} : M = M^T, u^T M u \geq 0 (u \in \mathcal{R}^n)\}.$$

For example, the Laplacian matrix of the graph G ,

$$L(G) := D_{\delta_1, \dots, \delta_n} - A(G) \in \mathcal{S}_+^n.$$

(Here $D_{\delta_1, \dots, \delta_n}$ denotes the diagonal matrix with diagonal elements $\delta_1, \dots, \delta_n$.)

It is well-known (see [6]), that the following statements are equivalent for a symmetric matrix $M = (m_{ij}) \in \mathcal{R}^{n \times n}$: a) $M \in \mathcal{S}_+^n$; b) $\lambda_M \geq 0$; c) M is Gram matrix, that is $m_{ij} = v_i^T v_j$ ($i, j = 1, \dots, n$) for some vectors v_1, \dots, v_n . Furthermore, by Lemma 2.1 in [8], the set \mathcal{S}_+^n can be described as

$$\mathcal{S}_+^n = \left\{ \left(\frac{a_i^T a_j}{(a_i a_j^T)_{11}} - 1 \right)_{i,j=1}^n \mid \begin{array}{l} m \in \mathcal{N}, a_i \in \mathcal{R}^m (1 \leq i \leq n) \\ a_i^T a_i = 1 (1 \leq i \leq n) \end{array} \right\}. \quad (3)$$

The stability number, $\alpha(G)$, is the maximum cardinality of the (so-called stable) sets $S \subseteq V(G)$ such that $\{i, j\} \subseteq S$ implies $\{i, j\} \notin E(G)$. The chromatic number, $\chi(G)$, is the minimum number of stable sets covering the vertex set $V(G)$.

Let us define an *orthonormal representation* of the graph G (shortly, o.r. of G) as a system of vectors $a_1, \dots, a_n \in \mathcal{R}^m$ for some $m \in \mathcal{N}$, satisfying

$$a_i^T a_i = 1 \ (i = 1, \dots, n), \ a_i^T a_j = 0 \ (\{i, j\} \in E(\overline{G})).$$

In the seminal paper [3] L. Lovász proved the following result, now popularly called *sandwich theorem*, see [1]:

$$\alpha(G) \leq \vartheta(G) \leq \chi(\overline{G}), \quad (4)$$

where $\vartheta(G)$ is the *Lovász number* of the graph G , defined as

$$\vartheta(G) := \inf \left\{ \max_{1 \leq i \leq n} \frac{1}{(a_i a_i^T)_{11}} : a_1, \dots, a_n \text{ o.r. of } G \right\}.$$

The Lovász number has several equivalent descriptions, see [3]. For example, by (3) and standard semidefinite duality theory (see e.g. [7]), it is the common optimal value of the Slater-regular primal-dual semidefinite programs

$$(TP) \quad \min \lambda, \begin{cases} x_{ii} = \lambda - 1 \ (i \in V(G)), \\ x_{ij} = -1 \ (\{i, j\} \in E(\overline{G})), \\ X = (x_{ij}) \in \mathcal{S}_+^n, \ \lambda \in \mathcal{R} \end{cases}$$

and

$$(TD) \quad \max \operatorname{tr}(JY), \begin{cases} \operatorname{tr}(Y) = 1, \\ y_{ij} = 0 \ (\{i, j\} \in E(G)), \\ Y = (y_{ij}) \in \mathcal{S}_+^n. \end{cases}$$

(Here tr stands for trace.) Reformulating the program (TD) , Lovász derived the following dual description of the theta number (Theorem 5 in [3]):

$$\vartheta(G) = \max \left\{ \sum_{i=1}^n (b_i b_i^T)_{11} : b_1, \dots, b_n \text{ o.r. of } \overline{G} \right\}. \quad (5)$$

Analogously, the *inverse theta number*, $\iota(G)$, satisfies the *inverse sandwich inequality*,

$$(\alpha(G))^2 + n - \alpha(G) \leq \iota(G) \leq n\vartheta(G), \quad (6)$$

see [10]. Here the inverse theta number, defined as

$$\iota(G) := \inf \left\{ \sum_{i=1}^n \frac{1}{(a_i a_i^T)_{11}} : a_1, \dots, a_n \text{ o.r. of } G \right\},$$

equals the common attained optimal value of the primal-dual semidefinite programs

$$(TP^-) \quad \inf \operatorname{tr}(Z) + n, \ z_{ij} = -1 \ (\{i, j\} \in E(\overline{G})), \ Z = (z_{ij}) \in \mathcal{S}_+^n,$$

$$(TD^-) \quad \sup \operatorname{tr}(JM), \begin{cases} m_{ii} = 1 \ (i = 1, \dots, n), \\ m_{ij} = 0 \ (\{i, j\} \in E(G)), \\ M = (m_{ij}) \in \mathcal{S}_+^n. \end{cases}$$

Moreover, rewriting the feasible solution M of the program (TD^-) as the Gram matrix $M = (b_i^T b_j)$ for some vectors $b_1, \dots, b_n \in \mathcal{R}^m$, we obtain the following analogue of (5):

$$\iota(G) = \max \left\{ \sum_{i,j=1}^n b_i^T b_j : b_1, \dots, b_n \text{ o.r. of } \overline{G} \right\}. \quad (7)$$

The structure of the paper is as follows: In Section 2 we will describe a refinement of (7) and also several new descriptions of the inverse theta function (with well-known analogues in the theory of the theta function). Some of these results will be applied in Section 3, where we present two new lower bounds for the stability number of a graph, and examine their additivity properties. Finally, in Section 4 we study two variants of the inverse theta function, and derive further bounds in the stable set problem.

2 New descriptions of $\iota(G)$

In this section we will describe three reformulations of the inverse theta number of a graph G . The results have analogues in the theory of the theta function, which we will mention in chronological order.

Let us denote by \mathcal{A}_G the following set of matrices:

$$\mathcal{A}_G := \left\{ A = (a_{ij}) \in \mathcal{R}^{n \times n} \left| \begin{array}{l} a_{ii} = 0 \ (i = 1, \dots, n), \\ a_{ij} = 0 \ (\{i, j\} \in E(G)), \\ a_{ij} = a_{ji} \ (\{i, j\} \in E(\overline{G})) \end{array} \right. \right\}.$$

We will describe bounds for the minimum eigenvalue λ_A with $A \in \mathcal{A}_G$.

First, we have for $A \in \mathcal{A}_G$ the lower bounds

$$\lambda_A \geq \lambda_{|A|} \geq -\Lambda_{|A|} \geq -\Lambda_{\overline{G}} \cdot \max_{i,j} |a_{ij}|,$$

by Rayleigh's theorem and the Perron-Frobenius theorem. (Here $|A| \in \mathcal{R}^{n \times n}$ denotes the elementwise maximum of the matrices A and (0) .)

On the other hand, using an equivalent form of the reformulation

$$\vartheta(G) = \max \left\{ \Lambda_M \left| \begin{array}{l} m_{ii} = 1 \ (i = 1, \dots, n), \\ m_{ij} = 0 \ (\{i, j\} \in E(G)), \\ M = (m_{ij}) \in \mathcal{S}_+^n \end{array} \right. \right\},$$

(see for example [1], [10]), L. Lovász proved in Theorem 6 of [3] the upper bound

$$\lambda_A \leq \frac{\Lambda_A}{1 - \vartheta(G)} \quad (A \in \mathcal{A}_G). \quad (8)$$

Analogously, as a consequence of the next theorem, we have also the upper bound

$$\lambda_A \leq \frac{\operatorname{tr}(JA)}{n - \iota(G)} \quad (A \in \mathcal{A}_G). \quad (9)$$

(Note that by Rayleigh's theorem $\operatorname{tr}(JA) \leq n\lambda_A$, and by the inverse sandwich theorem $\iota(G) - n \leq n(\vartheta(G) - 1)$ so there is no obvious dominance relation between the bounds in (8) and (9).)

THEOREM 2.1. *The program*

$$(P_1) : \quad \sup n + \frac{\operatorname{tr}(JA)}{-\lambda_A}, \quad A \in \mathcal{A}_G$$

has attained optimal value $\iota(G)$.

Proof. The variable transformations

$$M_A := I + \frac{1}{-\lambda_A}A, \quad A_M := M - I$$

show that programs (TD^-) and (P_1) are equivalent: if A and M are feasible solutions of (P_1) and (TD^-) , respectively, then M_A and A_M are feasible solutions of the other program such that between the corresponding values the inequalities

$$\operatorname{tr}(JM_A) \geq n + \frac{\operatorname{tr}(JA)}{-\lambda_A}, \quad n + \frac{\operatorname{tr}(JA_M)}{-\lambda_{A_M}} \geq \operatorname{tr}(JM)$$

hold. Hence, the two programs have the same (attained) optimal value. \square

A different approach leads to another description of the inverse theta number.

Karger, Motwani, and Sudan proved the reformulation

$$\frac{1}{1 - \vartheta(G)} = \min \left\{ \nu \left| \begin{array}{l} n_{ii} = 1 \quad (i = 1, \dots, n), \\ n_{ij} = \nu \quad (\{i, j\} \in E(\overline{G})), \\ N = (n_{ij}) \in \mathcal{S}_+^n, \nu \in \mathcal{R} \end{array} \right. \right\},$$

and used a variant of this theorem in their graph colouring algorithm. (See [2] for a summary of related results.) By the inverse sandwich theorem we have the lower bound

$$\frac{1}{1 - \vartheta(G)} \geq \frac{n}{n - \iota(G)}; \quad (10)$$

we will show that this latter value can be obtained as the optimal value of a semidefinite program, too.

Let us consider the primal-dual semidefinite programs

$$(P_2) : \quad \sup -\text{tr } B, \quad \begin{cases} b_{11} - b_{ii} = 0 \ (i = 2, \dots, n), \\ b_{ij} = 0 \ (\{i, j\} \in E(G)), \\ \text{tr}((J - I)B) = 1, \ B = (b_{ij}) \in \mathcal{S}_+^n, \end{cases}$$

$$(D_2) : \quad \inf \gamma, \quad \begin{cases} \text{tr } C = n, \\ c_{ij} = \gamma \ (\{i, j\} \in E(\overline{G})), \\ C = (c_{ij}) \in \mathcal{S}_+^n, \ \gamma \in \mathcal{R} \end{cases}$$

(see also Theorem 4.1 for a variant). The programs have common attained optimal value by standard semidefinite duality theory, see for example [7].

THEOREM 2.2. *The programs (P_2) and (D_2) have (common attained) optimal value $n/(n - \iota(G))$.*

Proof. Similarly as in the proof of Theorem 2.1, the variable transformations

$$M_B := \frac{n}{\text{tr } B} B, \quad B_M := \frac{1}{\text{tr}(JM) - n} M$$

show the equivalence of programs (P_2) and $n/(n - (TD^-))$, where the latter program can be obtained from (TD^-) formally exchanging its value function $\text{tr}(JM)$ for $n/(n - \text{tr}(JM))$ and adding the extra constraint $\text{tr}(JM) > n$. \square

Now, we turn to the third description of the inverse theta number.

We will use the following lemma, a slight modification of (7).

LEMMA 2.1. *For any graph G ,*

$$\iota(G) = \sup \left\{ \sum_{i,j=1}^n \hat{b}_i^T \hat{b}_j \mid \begin{array}{l} \hat{b}_1, \dots, \hat{b}_n \text{ o.r. of } \overline{G} \\ e_1^T \hat{b}_i > 0 \text{ for } i = 1, \dots, n \end{array} \right\},$$

with e_1 denoting the m -vector $(1, 0, \dots, 0)^T$.

Proof. Let (b_i) be an orthonormal representation of \overline{G} such that

$$\iota(G) = \sum_{i,j=1}^n b_i^T b_j$$

(that is an optimal solution in (7)). For $0 < \varepsilon < 1$, let us define an orthonormal representation $(\hat{b}_i(\varepsilon))$ of \overline{G} the following way:

$$(\hat{b}_i(\varepsilon)) := \begin{pmatrix} \sqrt{1 - \varepsilon^2} \cdot O \\ \varepsilon b_1, \dots, \varepsilon b_n \end{pmatrix},$$

where $O \in \mathcal{R}^{n \times n}$ is an orthogonal matrix satisfying $e_1^T O > 0$. Note that then $e_1^T \hat{b}_i(\varepsilon) > 0$ holds for all i . On the other hand, it can easily be verified that

$$\sum_{i,j=1}^n \hat{b}_i^T(\varepsilon) \hat{b}_j(\varepsilon) \rightarrow \iota(G) \quad (\varepsilon \rightarrow 1).$$

Hence, we have proved

$$\iota(G) \leq \sup \left\{ \sum_{i,j=1}^n \hat{b}_i^T \hat{b}_j \mid \begin{array}{l} \hat{b}_1, \dots, \hat{b}_n \text{ o.r. of } \bar{G} \\ e_1^T \hat{b}_i > 0 \text{ for } i = 1, \dots, n \end{array} \right\},$$

which is the nontrivial part of the lemma. \square

Applying the variable transformation described in (3) to the program in Lemma 2.1, as an immediate consequence we obtain an analogue of Theorem 2.2 in [8].

THEOREM 2.3. *The optimal value of the program*

$$(P_3) : \quad \sup \sum_{i,j=1}^n \frac{d_{ij} + 1}{\sqrt{(d_{ii} + 1) \cdot (d_{jj} + 1)}}, \quad \begin{cases} d_{ij} = -1 \quad (\{i, j\} \in E(G)), \\ D = (d_{ij}) \in \mathcal{S}_+^n \end{cases}$$

equals $\iota(G)$. \square

We will apply Theorem 2.3 in the next section for obtaining lower bounds in the stable set problem.

3 Lower bounds on $\alpha(G)$

In this section we will describe two lower bounds on the stability number of a graph G , and examine their additivity properties.

Note that the

$$Z_1 := L(\bar{G}), \quad Z_2 := \Lambda_{\bar{G}} I - A(\bar{G})$$

feasible solutions in (TP^-) give the inequalities

$$\sqrt{\iota(G)} \leq \sqrt{n(\mu_{\bar{G}} + 1)}, \quad \sqrt{n(\Lambda_{\bar{G}} + 1)}.$$

By Exercises 11.20 and 11.14 in [4], we have

$$\chi(G) \leq \Lambda_G + 1 \leq \sqrt{\mu_G(n-1)} + 1, \quad \mu_G \leq \Lambda_G.$$

On the other hand, easy calculation verifies

$$\sqrt{\mu_G(n-1)} + 1 \leq \sqrt{n(\mu_G + 1)}.$$

Hence, we have

$$\chi(\overline{G}) \leq \sqrt{n(\mu_{\overline{G}} + 1)} \leq \sqrt{n(\Lambda_{\overline{G}} + 1)}.$$

On the dual side instead of $\sqrt{\iota(G)}, \chi(\overline{G})$ we can approximate $\iota(G)/n, \alpha(G)$. Note that

$$D_1 := L(G), D_2 := \Lambda_G I - A(G)$$

are feasible solutions of the program (P_3) in Theorem 2.3. This fact implies the version of the following theorem, where $\alpha(G)$ is exchanged for $\iota(G)/n$. (For analogous results with $\vartheta(G)$, see [8].)

THEOREM 3.1. *For any graph G ,*

a)

$$\alpha'(G) := 1 + \sum_{\{i,j\} \in E(\overline{G})} \frac{2/n}{\sqrt{(\delta_i + 1) \cdot (\delta_j + 1)}} \leq \alpha(G);$$

b)

$$\alpha''(G) := 1 + \frac{\mu_{\overline{G}}}{\Lambda_G + 1} \leq \alpha(G).$$

Proof. By Exercise 11.14 in [4] we have $\mu_G \leq \Lambda_G$. Using this relation it is immediate that

$$\frac{n}{\Lambda_G + 1} \leq \alpha''(G) \leq \frac{n}{\mu_G + 1}.$$

We will show that the inequalities

$$\frac{n}{\mu_G + 1} \leq \alpha'(G) \leq \sum_{i=1}^n \frac{1}{\delta_i + 1}$$

hold also, from which the theorem follows, as

$$\sum_{i=1}^n \frac{1}{\delta_i + 1} \leq \alpha(G)$$

by the Caro-Wei theorem (see e.g. [8]).

First, using the obvious inequality

$$\frac{2}{\sqrt{\delta_i + 1} \cdot \sqrt{\delta_j + 1}} \leq \frac{1}{\delta_i + 1} + \frac{1}{\delta_j + 1}, \quad (11)$$

we obtain

$$\begin{aligned} \alpha'(G) &\leq 1 + \frac{1}{n} \cdot \sum_{i=1}^n \frac{1}{\delta_i + 1} \cdot (n - 1 - \delta_i) \\ &= \sum_{i=1}^n \frac{1}{\delta_i + 1}. \end{aligned}$$

On the other hand, we will verify the relation

$$\alpha'(G) \geq \frac{n}{\mu_G + 1}. \quad (12)$$

Using the arithmetic mean-harmonic mean inequality, it is easy to show that

$$\begin{aligned} \alpha'(G) &\geq 1 + \frac{4}{n} \cdot \sum_{\{i,j\} \in E(\overline{G})} \frac{1}{\delta_i + 1 + \delta_j + 1} \\ &\geq 1 + \frac{1}{n} (n\mu_{\overline{G}})^2 \bigg/ \sum_{\{i,j\} \in E(\overline{G})} (\delta_i + 1 + \delta_j + 1). \end{aligned}$$

Hence, to prove (12), it is enough to verify that

$$n\mu_{\overline{G}}(\mu_G + 1) \geq \sum_{\{i,j\} \in E(\overline{G})} (\delta_i + 1 + \delta_j + 1)$$

holds. This inequality can be rewritten as

$$\sum_{i=1}^n (n-1-\delta_i) \cdot \sum_{i=1}^n (\delta_i + 1) \geq n \cdot \sum_{i=1}^n (\delta_i + 1)(n-1-\delta_i),$$

and thus is a consequence of the Cauchy-Schwarz inequality. The proof of (12) is complete, as well. \square

The following theorem describes additivity properties of the bounds α' , α'' . (For analogous results, see [9].)

THEOREM 3.2. *With the lower bounds $\ell = \alpha'$, α'' we have*

- a) $\ell(G_1 + G_2) \leq \ell(G_1) + \ell(G_2)$,
- b) $\ell(\overline{G_1 + G_2}) \leq \max\{\ell(\overline{G_1}), \ell(\overline{G_2})\}$,

for any graphs G_1, G_2 .

Proof. Case 1: $\ell = \alpha'$. a) Rewriting the statement, we have to verify

$$\sum_{i \in V(G_1), j \in V(G_2)} \frac{2}{\sqrt{(\delta_i + 1)(\delta_j + 1)}} \leq \alpha'(G_1)n_2 + \alpha'(G_2)n_1,$$

that is (without loss of generality assuming $G_1 = G_2 = G$)

$$\left(\sum_{i=1}^n \frac{1}{\sqrt{\delta_i + 1}} \right)^2 \leq \alpha'(G)n.$$

In other words, we have to prove the inequality

$$\sum_{i=1}^n \frac{1}{\delta_i + 1} + \sum_{\{i,j\} \in E(G)} \frac{2}{\sqrt{(\delta_i + 1)(\delta_j + 1)}} \leq n,$$

which follows immediately applying (11).

b) is obvious, as

$$\begin{aligned} \alpha'(G_1 + G_2) &\leq \frac{\alpha'(\overline{G_1})n_1 + \alpha'(\overline{G_2})n_2}{n_1 + n_2} \\ &\leq \max\{\alpha'(\overline{G_1}), \alpha'(\overline{G_2})\} \end{aligned}$$

hold.

Case 2: $\ell = \alpha''$. With

$$u(G) := \Lambda_G + 1,$$

it is shown in [9] that the inequalities

$$\begin{aligned} u(G_1 + G_2) &\geq \max\{u(G_1), u(G_2)\}, \\ u(\overline{G_1 + G_2}) &\geq u(\overline{G_1}) + u(\overline{G_2}) \end{aligned}$$

hold. The statements a) and b), respectively, are straightforward consequences of these inequalities, after applying (1): For example, a) can be reduced this way to the inequality

$$\frac{n_2\mu_{G_1} + n_1\mu_{G_2}}{n_1 + n_2} \leq \max\{u(G_1), u(G_2)\} - 1,$$

which holds true, as $\mu_G \leq \Lambda_G$ for any graph G , by Exercise 11.14 in [4]. \square

See [9] for an application of this type of results in strengthening the bounds when the graph or its complement is not connected.

4 Upper bounds on $\alpha(G)$

In this section we introduce two variants of the inverse theta number. They constitute bounds for the stability numbers of G and \overline{G} .

Let us consider the primal-dual semidefinite programs

$$\begin{aligned} (P') : \quad & \inf n + \text{tr } Z', \quad \begin{cases} z'_{ij} \leq -1 \quad (\{i, j\} \in E(\overline{G})), \\ Z' = (z'_{ij}) \in \mathcal{S}_+^n, \end{cases} \\ (D') : \quad & \sup \text{tr } (JM'), \quad \begin{cases} m'_{ii} = 1 \quad (i = 1, \dots, n), \\ m'_{ij} = 0 \quad (\{i, j\} \in E(G)), \\ M' = (m'_{ij}) \in \mathcal{S}_+^n \cap \mathcal{R}_+^{n \times n}. \end{cases} \end{aligned}$$

The programs have common attained optimal value by standard semidefinite duality theory (see for example [7]), we will denote this value by $\iota'(G)$.

Obviously, $\iota'(G) \leq n\vartheta'(G)$, where $\vartheta'(G)$ is a sharpening of the theta number, due to McEliece, Rodemich, Rumsey, and Schrijver ($\alpha(G) \leq \vartheta'(G) \leq \vartheta(G)$, see for example [2]), defined as

$$\vartheta'(G) := \sup \left\{ \operatorname{tr}(JY') \mid \begin{array}{l} \operatorname{tr} Y' = 1, \\ y'_{ij} = 0 \ (\{i, j\} \in E(G)), \\ Y' = (y'_{ij}) \in \mathcal{S}_+^n \cap \mathcal{R}_+^{n \times n} \end{array} \right\}.$$

In other words, we have the inequality

$$\frac{n}{n - \iota'(G)} \leq \frac{1}{1 - \vartheta'(G)}, \quad (13)$$

an analogue of (10).

The reformulation

$$\frac{1}{1 - \vartheta'(G)} = \min \left\{ \nu' \mid \begin{array}{l} n'_{ii} = 1 \ (i = 1, \dots, n), \\ n'_{ij} \leq \nu' \ (\{i, j\} \in E(\bar{G})), \\ N' = (n'_{ij}) \in \mathcal{S}_+^n, \nu' \in \mathcal{R} \end{array} \right\}$$

is well-known, see for example [2]. Moreover, applying the technique used in the proof of Theorem 2.2 we obtain the analogous

THEOREM 4.1. *The primal-dual semidefinite programs*

$$(P'_2) : \quad \sup -\operatorname{tr} B', \quad \begin{cases} b'_{11} - b'_{ii} = 0 \ (i = 2, \dots, n), \\ b'_{ij} = 0 \ (\{i, j\} \in E(G)), \\ \operatorname{tr}((J - I)B') = 1, B' = (b'_{ij}) \in \mathcal{S}_+^n \cap \mathcal{R}_+^{n \times n}, \end{cases}$$

$$(D'_2) : \quad \inf \gamma', \quad \begin{cases} \operatorname{tr} C' = n, \\ c'_{ij} \leq \gamma' \ (\{i, j\} \in E(\bar{G})), \\ C' = (c'_{ij}) \in \mathcal{S}_+^n, \gamma' \in \mathcal{R} \end{cases}$$

have common attained optimal value $n/(n - \iota'(G))$. □

Besides the mentioned relations

$$\vartheta'(G) \geq \iota'(G)/n, \alpha(G),$$

we have also

$$\frac{1}{2} \left(1 + \sqrt{4(\iota'(G) - n) + 1} \right) \geq \iota'(G)/n, \alpha(G)$$

as the following theorem shows. (For analogous results with $\iota(G)$, see [10].)

THEOREM 4.2. *For any graph G ,*

$$\alpha(G) \leq \frac{1}{2} \left(1 + \sqrt{4(\iota'(G) - n) + 1} \right)$$

holds.

Proof. Let S be a stable set in G with cardinality $\#S = \alpha(G)$. Let us define the matrix $M' := (m'_{ij}) \in \mathcal{R}^{n \times n}$ the following way: let $m'_{ij} := 1$ if $i, j \in S$ or $i = j$, and let $m'_{ij} := 0$ otherwise. Then, the matrix M' is a feasible solution of the program (D') with corresponding value

$$(\#S)^2 + n - (\#S) \leq \iota'(G).$$

Hence, the statement follows. \square

The bound in Theorem 4.2 implies

$$\alpha(G) \leq \sqrt{\iota'(G)},$$

and also, by $\iota'(G) \leq \iota(G)$, the relations

$$\alpha(G) \leq \frac{1}{2} \left(1 + \sqrt{4(\iota(G) - n) + 1} \right) \leq \sqrt{\iota(G)}$$

from [10]. It is an open problem whether any of these bounds can be less than $\vartheta(G)$ or even $\vartheta'(G)$ for some graphs.

Another variant of the inverse theta number leads to new bounds on the stability numbers of G and \overline{G} .

Let us define $\iota''(G)$ as the common attained optimal value of the primal-dual semidefinite programs

$$(P'') : \quad \inf \operatorname{tr}(JM''), \quad \begin{cases} m''_{ii} = 1 \quad (i = 1, \dots, n), \\ m''_{ij} = 0 \quad (\{i, j\} \in E(G)), \\ M'' = (m''_{ij}) \in \mathcal{S}_+^n, \end{cases}$$

$$(D'') : \quad \sup n - \operatorname{tr} Z'', \quad \begin{cases} z''_{ij} = 1 \quad (\{i, j\} \in E(\overline{G})), \\ Z'' = (z''_{ij}) \in \mathcal{S}_+^n. \end{cases}$$

(See, for example, [7] for standard semidefinite duality theory.)

Both formulations constitute bounds in the stable set problem. On the primal side, we have

THEOREM 4.3. *For any graph G , the inequalities*

a) $\iota''(G) \leq \alpha(\overline{G})$,

b) $\iota''(G) \leq n - \alpha(G)$

hold.

Proof. Let us introduce the notation

$$M_S := (m_{ij}) \in \mathcal{R}^{n \times n}, \quad \text{where } m_{ij} := \begin{cases} 1 & \text{if } i, j \in S, i = j, \\ -\frac{1}{\#S-1} & \text{if } i, j \in S, i \neq j, \\ 0 & \text{otherwise,} \end{cases}$$

for $S \subseteq V(G)$.

a) Let S_1, \dots, S_k be a stable set partition of $V(G)$ such that the cardinality of the index set $\{i : \#S_i \geq 2\}$ is maximal. Then,

$$\bar{S} := \cup_{i=1}^k \{S_i : \#S_i = 1\}$$

is a stable set in \bar{G} . Furthermore, the matrix

$$\sum_{i=1}^k M_{S_i}$$

is feasible in (P'') with corresponding value $\#\bar{S} \leq \alpha(\bar{G})$, which completes the proof of statement a).

b) Let S_0 be a stable set in G with cardinality $\alpha(G)$. Then, similarly as in the proof of statement a), the matrix

$$M_{S_0} + \sum_{i \notin S_0} M_{\{i\}}$$

is a feasible solution of the program (P'') with corresponding value $n - \alpha(G)$. This finishes the proof of statement b), too. \square

The dual version of Theorem 4.3 is immediate, now.

THEOREM 4.4. *Let the matrix $Z'' = (z''_{ij}) \in \mathcal{R}^{n \times n}$ be a feasible solution of the program (D'') , that is let $Z'' \in \mathcal{S}_+^n$ such that $z''_{ij} = 1$ for $\{i, j\} \in E(\bar{G})$. Then, $\text{tr } Z'' \geq n - \alpha(\bar{G}), \alpha(G)$ hold.* \square

For example, for the cherry graph

$$G_0 := (\{1, 2, 3\}, \{\{1, 2\}, \{1, 3\}\}),$$

the optimal solutions of (P'') and (D'') , respectively,

$$M_0'' := \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & -1 & 1 \end{pmatrix}, \quad Z_0'' := \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

show that by Theorems 4.3 and 4.4 $\alpha(G_0) \leq 2$ and $\alpha(\bar{G}_0) \geq 1$.

On the negative side, $\iota''(G) = 0$ if $V(G)$ has a partition $S_1 \cup \dots \cup S_k$ where all the sets S_i are stable with cardinality at least 2 (see the proof of Theorem 4.3). In this case the bounds described in Theorems 4.3 and 4.4 are trivial. It is an open problem to characterize the graphs G with $\iota''(G) = 0$.

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