Optimal Bounds for the Largest Eigenvalue of a 3 × 3 Correlation Matrix

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Abstract
A new approach that bounds the largest eigenvalue of 3 × 3 correlation matrices is presented. Optimal bounds by given determinant and trace of the squared correlation matrix are derived and shown to be more stringent than the optimal bounds by Wolkowicz and Styan in specific cases.

Keywords
Correlation Matrix, Positive Semi-Definite Matrix, Extreme Point, Eigenvalue, Inequality

1. Introduction
The topic of bounds on eigenvalues of symmetric matrices has a long history (e.g. [1], Chap. III). In some situations optimal bounds have been found. For the set of complex matrices $A = \{a_{ij}\}, 1 \leq i, j \leq n$, with real eigenvalues, Wolkowicz and Styan [2] obtained optimal bounds by given $\text{Tr}(A)$ and $\text{Tr}(A^2)$. For the same set of matrices with positive eigenvalues, Merikoski and Virtanen [3] [4] have studied optimal bounds by given $\text{Tr}(A)$ and $\text{det}(A)$. Zhan [5] obtained the optimal bounds for the smallest and largest eigenvalues of real symmetric matrices whose entries belong to a fixed finite interval. However, when restricted to the set of real 3 × 3 correlation matrices, these bounds collapse to useless or trivial bounds, as argued in the Remarks 2.1. Moreover, for correlation matrices $R = \{r_{ij}\}, 1 \leq i, j \leq n$, with unit diagonal elements, one has always $\text{Tr}(R) = n$. Therefore, the separate knowledge of $\text{Tr}(R^2)$ and $\text{det}(R)$ does not exhaust the complete information about a correlation matrix, even in the case of 3 × 3 correlation matrices. It is therefore justified to search for further possibly optimal bounds on eigenvalues for correlation matrices.

The present study is devoted to a new approach for bounding the largest eigenvalue of 3 × 3 correlation matrices. In Theorem 2.1 we derive some new optimal bounds by given determinant and trace of the squared correlation matrix. They are compared in Theorem 3.1 to the optimal bounds in [2] and found to be more stringent in some specific cases. Section 4 illustrates with some numerical comparisons.

2. Bounds by Given Determinant and Trace of the Squared Correlation Matrix

Starting point is a real $3 \times 3$ matrix $A = (a_{ij}), 1 \leq i, j \leq 3$, with characteristic polynomial

$$q_s(\lambda) = \lambda^3 - s_1 \lambda^2 + \frac{1}{2}(s_2^3 - s_2) \lambda - d,$$

where $d = \det(A)$ is the determinant, and $s_i = \text{Tr}(A^i), i = 1, 2,$ are the traces of the matrix and its square. Each zero of this polynomial is called an eigenvalue (EV). Expressed in terms of the variable $z = \lambda - 1$ one finds the polynomial

$$p_s(z) = z^3 - (s_1 - 3) z^2 - \left(2s_1 + \frac{1}{2}(s_2 - s_2^3) - 3 \right) z - \left( d + \frac{1}{2}(s_2^3 - s_2) + s_1 - 1 \right).$$

Restricting the attention to correlation matrices $R = (r_{ij}), 1 \leq i, j \leq 3$, with unit diagonal elements, one has $s_i = \text{Tr}(R) = 3$ and the polynomial simplifies to the "depressed cubic"

$$p_s(z) = z^3 - Pz - 2Q, \quad P = \frac{1}{2}(s_2 - 3), \quad Q = \frac{1}{2}d + \frac{1}{4}(s_2 - 5).$$ (2.2)

The set of correlation matrices is uniquely determined by the set of $3$ upper diagonal elements $r = (r_{ij}), 1 \leq i < j \leq 3$, denoted by $E_3$. For convenience, we use throughout the algebraic notation $r = (r_{13}, r_{32}, r_{12}) = (\alpha, \beta, \gamma) \in [-1,1]^3$. It is known that, up to permutations, an element $r \in E_3$ if, and only if, one has $\gamma \in [\gamma^-, \gamma^+], \quad \gamma^+ = \alpha\beta \pm \sqrt{(1-\alpha^2)(1-\beta^2)}, \quad$ where the interval bounds characterize the extreme points of the elliptope $E_3$ (e.g. [6], Theorem 3.1, and [7], Theorem 3.1). Since $s_2 = \text{Tr}(R^2) = 3 + 2(\alpha^2 + \beta^2 + \gamma^2)$ and $d = \det(R) = 1 - (\alpha^2 + \beta^2 + \gamma^2) + 2\alpha\beta\gamma$, the coefficients of the cubic in (2.2) are given by

$$P = \alpha^2 + \beta^2 + \gamma^2, \quad Q = \alpha\beta\gamma.$$ (2.3)

We ask for possibly optimal bounds for the largest EV (LEV) of a correlation matrix by given $d = \det(R)$ and $s_2 = \text{Tr}(R^2)$, or equivalently by given $(P, Q)$, which is the maximum available information. The following sharp inequality, which characterizes the semi-definite property of a $3 \times 3$ correlation matrix, is trivial but an essential ingredient of the analysis.

**Lemma 2.1.** For all $r \in E_3$, one has the inequality $d = \det(R) = 1 - P + 2Q \geq 0$. It is attained at the extreme points of $E_3$ consisting of all $r^+ = (\alpha, \beta, \gamma) \in [-1,1]^3$.\[\]

In the following, we assume first that $r = (\alpha, \beta, \gamma) \neq (0,0,0)$, that is, the EVs $\lambda_1 \geq \lambda_2 \geq \lambda_3 \geq 0$ are not all one, that is $(\lambda_1, \lambda_2, \lambda_3) \neq (1,1,1)$, and in particular $\lambda_1 > 1$. Therefore, one searches for the positive zero $z = \lambda - 1 > 0$ of the depressed cubic (2.2). Making use of the identity $d = 1 - P + 2Q$ we rewrite the latter in two different ways:

$$p_r(z) = (z^2 - z - 2Q)(z + 1) + dz = (z^2 - z - (P - 1))(z + 1) - d.$$ (2.4)

Using that $d \geq 0, z > 0$, one sees that a positive zero of these two cubic polynomials necessarily satisfy the two quadratic inequalities

$$(I) \quad z^2 - z - 2Q \leq 0, \quad (II) \quad z^2 - z - (P - 1) \geq 0.$$ (2.5)

In terms of $(P, Q)$ the possible ranges of validity of these inequalities are as follows:

**Inequality (I)**

$$0 \leq \frac{1}{2}(1 - \sqrt{1 + 8Q}) \leq z \leq \frac{1}{2}(1 + \sqrt{1 + 8Q})$$ (2.6)

By Lemma 2.2 below the square root is always real. The lower bound is non-negative provided $Q \leq 0$, a restriction assumed in this case.
Inequality (II)

\[ 0 \leq z \leq \frac{1}{2} (1 - \sqrt{4P - 3}) \text{, or } z \geq \frac{1}{2} (1 + \sqrt{4P - 3}) > 0 \]  

(2.7)

The square root is real provided \( 4P - 3 \geq 0 \), which is assumed in this case. If \( 4P - 3 < 0 \) the inequality (II) is always satisfied and no information, besides \( 1 < \lambda \leq 3 \), about the LEV is gained herewith. The upper bound is non-negative provided \( \frac{3}{4} \leq P < 1 \), a restriction assumed in this situation.

**Lemma 2.2.** For all \( r \in E_3 \) one has the inequality \( 1 + 8QQ \geq 0 \).

**Proof.** Clearly, one has \( 1 + 8QQ \geq 1 + 8Q \left( \alpha \beta - \sqrt{1 - \alpha^2} \right) \geq 0 \) if, and only if, one has \( f(\alpha, \beta) = 64(\alpha \beta)^2 (\alpha^2 + \beta^2) - 48(\alpha \beta)^2 + 1 \geq 0 \).

**Case 1:** \( \alpha^2 + \beta^2 \geq 1 \)

One has \( f(\alpha, \beta) \geq 16(\alpha \beta)^2 + 1 \geq 0 \) and the inequality is fulfilled.

**Case 2:** \( \alpha^2 + \beta^2 = c^2 < 1 \), \( \alpha^2 \leq c^2 < 1 \)

One has \( f(\alpha, \beta) = 16\alpha^2 (c^2 - \alpha^2) (4c^2 - 3) + 1 \). Set \( \delta = c^2 \in [0, 1], \alpha = \sqrt{x} \in [0, \delta] \). Then, for fixed \( \delta \) one must have \( g(x, \delta) = 16x(x - \delta) (4\delta - 3) + 1 \geq 0 \). If \( \delta \geq \frac{3}{4} \) then the inequality is always fulfilled. If \( 0 \leq \delta < \frac{3}{4} \) then rewrite \( g(x, \delta) = 16x(\delta - x)(3 - 4\delta) + 1 \geq 0 \). Since \( 3 - 4\delta > 0 \) this is minimum for \( x = \frac{\delta}{2} \), and

\[ g(x, \delta) \geq h(\delta) = g\left(\frac{\delta}{2}, \delta\right) = 1 - 4(3 - 4\delta)^2. \]  

Since the function \( h(\delta) \) is minimum at \( \delta = \frac{1}{2} \), one gets

\[ g(x, \delta) \geq h\left(\frac{1}{2}\right) = 0. \]

How are the feasible inequalities (I) and (II) linked? Lemma 2.1 implies the inequalities

\[ \frac{1}{2} (1 - \sqrt{1 + 8Q}) \leq \frac{1}{2} (1 - \sqrt{4P - 3}) \leq \frac{1}{2} (1 + \sqrt{4P - 3}) \leq \frac{1}{2} (1 + \sqrt{1 + 8Q}), \]  

(2.8)

where the first and third inequalities are attained at the extreme points \( r^\pm = (\alpha, \beta, \gamma^\pm) \in [-1, 1]^3 \), and the middle one is attained when \( P = \frac{3}{4} \). These inequalities restrict the number of LEV bounds to the meaningful combinations stated in the main result below. For convenience, we parameterize elements \( r \in E_3 \) as univariate functions \( r(x) = (\alpha, \beta, x) \in [\gamma^-, \gamma^+] \). Similarly, the coefficients \( (P, Q) \) in (2.3) are parameterized as \( P(x) = \alpha^2 + \beta^2 + x^2 \), \( Q(x) = \alpha \beta x \). The result depends upon \( b = \frac{1}{2} \sqrt{3 - 4P(0)} \) defined if \( P(0) \leq \frac{3}{4} \).

**Theorem 2.1.** (Optimal bounds for the LEV of a \( 3 \times 3 \) correlation matrix). The largest eigenvalue \( \lambda_1(x) \) of a \( 3 \times 3 \) correlation matrix \( r(x) = (\alpha, \beta, x) \in E_3 \) satisfies the following bounds:

**Upper bound**

Case (A): \( \lambda_1(x) \leq \frac{1}{2} (3 + \sqrt{1 + 8Q(x)}) \), \( x \in [\gamma^-, \gamma^+] \)

The upper bound is attained at the extreme points \( x = \gamma^\pm \).

**Lower bound**

Case (B): \( \lambda_1(x) \geq \frac{1}{2} (3 + \sqrt{4P(x) - 3}) \).

Sub-Case (B1): \( P(0) \geq \frac{3}{4} \), \( x \in [\gamma^-, \gamma^+] \)
Sub-Case (B2): \( P(0) < \frac{3}{4}, \quad x \in [\gamma^-, -b] \cup [b, \gamma^+] \)

The lower bound is attained at the extreme points \( x = \gamma^\pm \).

Case (C1): \( \lambda_i(x) \geq \frac{1}{2} \left( 3 - \sqrt{1 + 8Q(x)} \right) \)

Sub-Case (C11): \( P(0) < \frac{3}{4}, \quad Q(1) \geq 0, \quad x \in [-b, 0] \)

Sub-Case (C12): \( P(0) < \frac{3}{4}, \quad Q(1) \leq 0, \quad x \in [0, b] \)

The lower bound \( \lambda_1(0) = 1 \) is attained at the “zero” correlation matrix \( (\alpha, \beta, \gamma) = (0, 0, 0) \).

Case (C2): \( \lambda_1(x) \geq 1 \)

Sub-Case (C21): \( P(0) < \frac{3}{4}, \quad Q(1) > 0, \quad x \in (0, b) \)

Sub-Case (C22): \( P(0) < \frac{3}{4}, \quad Q(1) < 0, \quad x \in (-b, 0) \)

The lower bound is not attained, but in the limit as \( x \to 0^\pm \) one has \( \lambda_1(0^\pm) = 1 \).

Remarks 2.1. If the bounds are attained, that is in the cases (A), (B) and (C1), they are the best bounds by given \( (P, Q) \), or equivalently \( s_2 = \text{Tr}(R^2), \quad d = \det(R) \). It is interesting to compare the new optimal bounds with related results, which deal, however, all with larger sets of matrices. For complex matrices \( A = (a_{ij}), 1 \leq i < j \leq n \), of arbitrary dimensions with real eigenvalues, Wolkowicz and Styan [2] obtained optimal bounds by given \( \text{Tr}(A) \) and \( \text{Tr}(A^2) \), called hereafter WS bounds. Albeit this is not the available maximum information for \( 3 \times 3 \) correlation matrices, a detailed comparison with the WS bounds is instructive and provided in Section 3. In contrast to this, for the same set of matrices with positive eigenvalues, the bounds in [3] by given \( \text{Tr}(A) \) and \( \det(A) \), hereafter called MV bounds, are not optimal, that is not attained for a specific matrix with the given properties. Even more, the best possible bounds cannot in general be expressed algebraically, as shown in [4]. More recently, Zhan [5] obtains the optimal bounds for the smallest and largest eigenvalues of real symmetric matrices whose entries belong to a fixed finite interval. However, when restricted to the set of real \( 3 \times 3 \) correlation matrices, the Zhan bounds collapse to useless or trivial bounds ([5], Corollary 2 (ii), p. 854, Theorem 5 (ii), pp. 854-855). Information on further possible comparison statements are provided in Section 3.

Proof of Theorem 2.1. It is clear by (2.6) and (2.8) that the upper bound in Case (A) must hold. Equality in (I) is attained when \( d = 0 \), that is \( x = \gamma^\pm \) by Lemma 2.1. To derive the lower bounds suppose first that \( 4P(x) - 3 \geq 0 \). Then, the lower bound in (2.7) is defined and by (2.8) it must imply a lower bound for the LEV. Again, equality in (II) is attained when \( d = 0 \), that is \( x = \gamma^\pm \) by Lemma 2.1. The distinction between the Sub-Cases (B1) and (B2) follows from the analysis of the inequality \( 4P(x) - 3 < 0 \). Suppose now that \( 4P(x) - 3 < 0 \).

Since the inequality (II) does not provide any information on the LEV, a lower bound for the LEV is (2.6), which is defined when \( Q(x) \leq 0 \) and attained when \( (\alpha, \beta, \gamma) = (0, 0, 0) \). The distinction between the Sub-Cases (C11) and (C12) is obtained through analysis of the inequalities \( 4P(x) - 3 < 0 \), \( Q(x) \leq 0 \). No information is available when \( 4P(x) - 3 < 0 \), \( Q(x) > 0 \), hence \( \lambda_1(x) \geq 1 \). An analysis of the preceding inequalities yields the distinction between the Sub-Cases (C21) and (C22).

The following result is about uniform bounds, which do not depend on the given information.

Corollary 2.1. (Uniform bounds for the LEV of a \( 3 \times 3 \) correlation matrix). If \( P = \alpha^2 + \beta^2 + \gamma^2 \geq \frac{3}{4} \) the LEV of a \( 3 \times 3 \) correlation matrix \( r = (\alpha, \beta, \gamma) \in E_3 \) satisfies the absolute bounds \( \frac{3}{2} \leq \lambda_i \leq 3 \). The upper bound is attained at \( (\alpha, \beta, \gamma) = (e_1, e_2, e_3) \), \( e_i \in [-1, 1] \) and the lower bound at \( (\alpha, \beta, \gamma) = \frac{1}{2}(e_1, e_2, -e_3) \), \( e_i \in [-1, 1] \).

Proof. Clearly, the absolute maximum of value 3 in case (A) is attained when \( Q(x) = 1 \) and \( x = \gamma^\pm \), which
is only possible for \( (\alpha, \beta, \gamma) = (e_1, e_2, e_3) \), \( e_i \in \{-1,1\} \). Similarly, the absolute minimum of value \( \frac{3}{2} \) in case (B), which holds when \( P(x) \geq \frac{3}{4} \), is attained when \( P(x) = \frac{3}{4} \) and \( x = \gamma^v \), that is for 
\[
(\alpha, \beta, \gamma) = \frac{1}{2}(e_1, e_2, -e_1, e_2), \ e_i \in \{-1,1\}.
\]

**Remark 2.2.** The bounds also follow from the WS bounds in (3.1) of the next section. However, only the lower bound (B) tells us when it is attained.

### 3. Analytical Comparison Results

For correlation matrices the WS bounds are optimal conditionally on the value of \( \text{Tr}(R^2) \), or equivalently on \( P \). Since the new bounds of Theorem 2.1 depend on both \( P \) and \( Q \), it is useful to analyze the conditions under which the one bounds are more stringent than the others. It is remarkable that for \( 3 \times 3 \) correlation matrices the WS bounds yield actually contiguous bounds for all 3 EVs ([2], Equation (2.31)): 

\[
1 - \frac{2}{3} \sqrt{3P} \leq \lambda_1 \leq 1 - \frac{1}{3} \sqrt{3P} \leq \lambda_2 \leq 1 + \frac{1}{3} \sqrt{3P} \leq \lambda_3 \leq 1 + \frac{2}{3} \sqrt{3P}.
\]

When referring to the bounds in (3.1), as function of \( x \in [\gamma^-, \gamma^+] \), the notation \( \lambda_1^{\text{WS}}(x) \) and \( \lambda_1^{M,\text{WS}}(x) \) is used for the lower respectively upper bound. Similar notations are used for the bounds in Theorem 2.1, where the upper indices refer to the various cases.

**Theorem 3.1.** The WS bounds compare with the bounds of Theorem 2.1 as follows:

**Upper bound**

(Aa) \( P(x) < \frac{3}{8}(1 + 4Q(x)) \) \( \Rightarrow \lambda_1^{M,\text{WS}}(x) < \lambda_1^{M,\text{A}}(x) \)

With \( P^*(x) = \frac{3}{8}(1 + 4Q(x) + \sqrt{1 + 8Q(x)}) \), one has

(Ab) \( \frac{3}{8}(1 + 4Q(x)) \leq P(x) < P^*(x) \) \( \Rightarrow \lambda_1^{M,\text{WS}}(x) < \lambda_1^{M,\text{A}}(x) \)

(Ac) \( P(x) \geq P^*(x) \) \( \Rightarrow \lambda_1^{M,\text{WS}}(x) \geq \lambda_1^{M,\text{A}}(x) \)

**Lower bound**

(B) \( \lambda_1^{n,\text{RS}}(x) \leq \lambda_1^{n,\text{B}}(x) \)

(C1) \( \lambda_1^{n,\text{RS}}(x) \geq \lambda_1^{n,\text{C1}}(x) \)

(C2) \( \lambda_1^{n,\text{RS}}(x) \geq \lambda_1^{n,\text{C2}}(x) = 1 \)

**Proof.** A case by case analysis based on Theorem 2.1 and Equation (3.1) is required. In Case (A) one has \( \lambda_1^{M,\text{RS}}(x) \geq \lambda_1^{M,\text{A}}(x) \) if, and only if, the inequality \( 3\sqrt{1 + 8Q(x)} \leq 8P(x) - 3(1 + 4Q(x)) \) is fulfilled. In Sub-Case (Aa) this cannot be fulfilled, hence \( \lambda_1^{M,\text{RS}}(x) < \lambda_1^{M,\text{A}}(x) \). Otherwise, the preceding inequality holds if, and only if, one has

\[
4P^*(x) - 3(1 + 4Q(x))P(x) + 9Q^2(x) \geq 0.
\]

This quadratic polynomial in \( P(x) \) has the non-negative discriminant \( \Delta(x) = 9 \cdot (1 + 8Q(x)) \). Therefore, the inequality holds if \( P(x) \geq P^*(x) \), with \( P^*(x) = \frac{3}{8}(1 + 4Q(x) + \sqrt{1 + 8Q(x)}) \) the non-negative zero of the quadratic polynomial, which is Sub-Case (Ac). The remaining situation is Sub-Case (Ab). In Case (B) the inequality \( \lambda_1^{n,\text{RS}}(x) \leq \lambda_1^{n,\text{B}}(x) \) holds if, and only if, one has \( 3\sqrt{4P(x) - 3} \geq -(4P(x) - 3) \), which is obviously
fulfilled because \( 4P(x) - 3 \geq 0 \). In Case (C1) one has \( \lambda_{n,WS}^{m}(x) \geq \lambda_{n,CL}^{m}(x) \) if, and only if, one has
\[
P^2(x) - 3\{1 + 4Q(x)\} P(x) + 36Q^2(x) \leq 0.
\]

With Lemma 3.1 below, and the proof of Theorem 3.1, this is only possible if
\[
P^*(x) = \frac{3}{2}\{1 + 4Q(x) - \sqrt{1 + 8Q(x)}\} \leq P(x) < \frac{3}{4}, \quad Q(x) \leq 0, \quad x \in [-b,b],
\]
where \( P(x) \) is the smallest zero of the preceding quadratic polynomial. Indeed, let \( b = \frac{1}{2}\sqrt{3 - 4P(0)} \) as in Section 2, and consider the function
\[
f(x) = P(x) - \frac{3}{2}\{1 + 4Q(x) - \sqrt{1 + 8Q(x)}\}, \quad x \in [-b,b].
\]
One has \( \frac{1}{2} f'(x) = x - 3Q(1)\{1 - \sqrt{1 + 8Q(x)}\}^{-1} \), and \( f'(x) = 0 \) if, and only if, one has
\[
x \cdot g(x) = 0, \quad g(x) = 8Q(1)x^2 - 48Q(1)^2 - \{48Q(1)^2 - 1\}x + 6Q(1)\{12Q81^2 - 1\}.
\]
The possible zeros of \( f'(x) = 0 \) are \( x_0 = 0, x_{2,3} = \left\{48Q(1)^2 - 1 \pm \sqrt{1 + 96Q(1)^2}\right\}/16Q(1) \). In Sub-Case (C11) one has \( Q(1) \geq 0, x \in [-b,0] \), and only \( x_0, x_3 \) (with “–” sign) may belong to \( x \in [-b,0] \). Now, one has
\[
\frac{1}{2} f^*(x) = 1 - 12Q(1)^2(1 + 8Q(x))^{-3/2}.
\]
One sees that \( f^*(0) \geq 0 \) and \( x_3 \in [-b,0] \) if, and only if, one has \( Q(1) \leq \sqrt{3}/6 \). In this situation \( x_1 = 0 \) is either a relative minimum (or an inflection point when \( f^*(0) = 0 \)) and \( x_3 \) a relative maximum. This implies that \( f(x) \geq \min\{f(-b), f(0)\} \). If \( f^*(0) < 0 \) then \( x_3 \not\in [-b,0] \) and \( x_1 = 0 \) is a relative maximum, and again \( f(x) \geq \min\{f(-b), f(0)\} \). But \( f(0) = P(0) \geq 0 \) and
\[
f(-b) = \frac{3}{4} g(z), \quad g(z) = z + 2\sqrt{1-z} - 1, \quad z = 8Q(1)b.
\]
Since \( 0 \leq z < 1 \) and \( g'(z) \leq 0 \) one sees that \( g(z) > g(1) = 0 \). If follows that \( f(x) \geq 0, x \in [-b,0] \). In Sub-Case (C12) one shows similarly that \( f(x) \geq 0, x \in [0,b] \). This shows the inequality \( P\ast(x) \leq P(x), x \in [-b,b] \). The inequality in Case (C2) is trivial.

\textbf{Lemma 3.1}. For all \( r \in E_3 \) one has the inequality\( P \leq \frac{3}{2}\{1 + 4Q\} \).

\textbf{Proof}. If \( P > \frac{3}{2}\{1 + 4Q\} \) then \( d = 1 - P + 2Q \geq 0 \) implies that \( 1 + 2Q \geq P > \frac{3}{2}\{1 + 4Q\} \), hence \( 1 + 8Q < 0 \), in contradiction to Lemma 2.2.

According to Theorem 3.1 the new bounds are more stringent than the WS bounds in the following cases: (Ac) and (B), Similar comparison statements can be made for other LEV bounds. For example, one can compare Theorem 2.1 with the MV bounds in \[3\], Theorems 1, 2, 3, or with Theorem 2.1 in \[8\]. It might also be useful to compare the new lower bounds with the classical lower bound \( \lambda_{n}(x) \geq \frac{2}{3}(\alpha + \beta + x) \) and its improvement in \[9\], or with the lower bound in \[10\], Theorem 3.1. We note that these few further possibilities do certainly not exhaust the list of LEV bounds found in the literature.

\textbf{4. Some Numerical Comparisons}

To conclude this study, it might be instructive to illustrate the results numerically. Since the LEV is the largest root of a cubic polynomial, a lot of formulas exist to calculate it. A most popular one is the exact trigonometric
Table 1. Numerical comparison of LEV bounds.

<table>
<thead>
<tr>
<th>Case</th>
<th>( r = (\alpha, \beta, x) \in E_i )</th>
<th>( \lambda_i(x) )</th>
<th>New Bound</th>
<th>WS Bound</th>
</tr>
</thead>
<tbody>
<tr>
<td>(Aa)</td>
<td>(0.25, 0.25, 0)</td>
<td>1.35355</td>
<td>2</td>
<td>1.40825</td>
</tr>
<tr>
<td>(Ab)</td>
<td>(−0.5, 0.5, −0.5)</td>
<td>2</td>
<td>2.20711</td>
<td>2</td>
</tr>
<tr>
<td>(Ac)</td>
<td>(−0.5, 0.5, 0.5)</td>
<td>1.5</td>
<td>2</td>
<td>1.5</td>
</tr>
<tr>
<td></td>
<td>(−0.5, 0.5, −1)</td>
<td>2.36603</td>
<td>2.36603</td>
<td>2.41421</td>
</tr>
<tr>
<td>(B)</td>
<td>(−0.5, 0.5, −1)</td>
<td>2.36603</td>
<td>2.36603</td>
<td>1.70711</td>
</tr>
<tr>
<td></td>
<td>(−0.5, 0.5, 0.5)</td>
<td>1.5</td>
<td>1.5</td>
<td>1.5</td>
</tr>
<tr>
<td></td>
<td>(−0.5, 0.5, −0.5)</td>
<td>2</td>
<td>1.5</td>
<td>1.5</td>
</tr>
<tr>
<td>(C1)</td>
<td>(0.25, 0.25, −0.25)</td>
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<td>1.03229</td>
<td>1.25</td>
</tr>
<tr>
<td>(C2)</td>
<td>(0.5, 0.5, 0.49999)</td>
<td>1.99999</td>
<td>1</td>
<td>1.5</td>
</tr>
</tbody>
</table>

Vieta formula, also known under Chebyshev cube root’s formula. Following [11] in Section 6.1, one gets the roots of the depressed cubic Equation (2.2), which yield the trigonometric EV formulas:

\[
\lambda_1 = 1 + \frac{2}{3}\sqrt{3}P \cdot \cos \varphi, \lambda_2 = 1 - \frac{1}{3}\sqrt{3}P \cdot (\cos \varphi - \sqrt{3} \sin \varphi), \lambda_3 = 1 - \frac{1}{3}\sqrt{3}P \cdot (\cos \varphi + \sqrt{3} \sin \varphi),
\]

with \( \varphi = \frac{1}{3} \arccos \left\{ 3\sqrt{3}Q/P^{3/2} \right\} \).

Note that the first use of Vieta’s formulas for computing the eigenvalues of a 3 × 3 matrix is apparently due to [12]. Other authors making use of it include [13] and [14] among others.

Another quite recent and attractive evaluation of the LEV, which can be applied to correlation matrices of any dimension, is the limiting Bernoulli type ratio approximation formula in [15], in Theorem 2.1 and Section 3. For an arbitrary correlation matrix \( R = (r_{ij})_1 \leq i, j \leq n \), one has the limiting formula

\[
\lambda_i = \lim_{k \to \infty} \left( \frac{\text{Tr}(R^{-k+1})}{\text{Tr}(R^k)} \right).
\]

Table 1 provides a selection of numerical examples for the possible cases in Theorem 3.1.

References


