# A NOTE ON THE APPROXIMATED INVERSE OF A NON-NEGATIVE SYMMETRIC MATRIX

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**Abstract:** This paper studies the issue of the approximated inverse of nonnegative symmetric matrices. By using the matrix  $S = (s_{i,j})$  to approximate its inverse, an explicit bound on the approximation error is obtained, and one conclusion that the inverse is well approximated to the order  $1/(n-1)^2$  uniformly for large n is also proved.

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#### 1 Introduction

When solving the solution for a large system of linear equations, a good approximate inverse of the coefficient matrix is crucially important in establishing fast convergence rates for iterative algorithms. See the extensive reviews [1, 5, 7, 20]. Here, we are concerned with a  $n \times n$  symmetric diagonally dominant matrices  $T = (t_{i,j})$  with positive elements, i.e.,

$$t_{i,j} = t_{j,i} > 0 \text{ and } t_{i,i} \ge \sum_{j=1, j \neq i}^{n} t_{i,j}, \ i = 1, \cdots, n.$$
 (1.1)

It is easy to show that T must be positive definite. This kind of diagonally dominant nonnegative matrices has received wide attention [6, 8, 10]. In [2, 7, 9], the problems on inverses of nonnegative matrices have been investigated. Markham [13] and Martínez et al. [14] studied the sufficient conditions that the inverses of nonnegative matrices are Mmatrices. We propose to approximate the inverse of T,  $T^{-1}$ , by the matrix  $S = (s_{i,j})$ , where

$$s_{i,j} = \frac{\delta_{i,j}}{t_{i,i}} - \frac{1}{t_{..}},$$

and  $t_{..} = \sum_{i,j=1}^{n} (1 - \delta_{i,j}) t_{i,j}$ . In a special case that  $t_{i,i} = \sum_{j \neq i} t_{i,j}$  for all *i*, Yan and Xu [19] have obtained the upper bound of the approximation errors when using *S* to approximate the

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inverse of T, which is crucially used to establish the asymptotical normality of an estimated vector in the  $\beta$ -model for undirected random graphs with a growing number of nodes. In this paper, we derive an explicit upper bound on the approximation error for general cases (1.1).

## 2 An Explicit Bound on the Approximation Error

Let  $m := \min_{1 \le i < j \le n} t_{i,j}$ ,  $\Delta_i := t_{i,i} - \sum_{j \ne i} t_{i,j}$ ,  $M := \max\{\max_{1 \le i < j \le n} t_{i,j}, \max_{1 \le i \le n} \Delta_i\}$ , and for a matrix  $A = (a_{i,j})$ , define  $||A|| := \max_{i,j} |a_{i,j}|$ . We have the following theorem.

Theorem 2.1 If

$$C(m,M) = \frac{2(n-2)m}{nM + (n-2)m} - \frac{M}{m(n-1)} - \frac{(n-2)Mm}{[(n-2)m + M][(n-2)m + 2M]} > 0, \quad (2.1)$$

then

$$||T^{-1} - S|| \le \frac{1}{(n-1)^2} \times \left[\frac{1}{C(m,M)} \left(\frac{M}{m^2} + \frac{4M}{m^2n}\right) + \frac{n-1}{mn}\right].$$

**Proof** Let  $I_n$  be the  $n \times n$  identity matrix. Define  $F = (f_{ij}) = T^{-1} - S$ ,  $V = (v_{ij}) = I_n - TS$  and  $W = (w_{ij}) = SV$ . We have the recursion

$$F = T^{-1} - S = (T^{-1} - S)(I_n - TS) + S(I_n - TS) = FV + W.$$
(2.2)

Note that

$$\begin{aligned}
v_{i,j} &= \delta_{i,j} - \sum_{k=1}^{n} t_{i,k} s_{k,j} \\
&= \delta_{i,j} - \sum_{k=1}^{n} t_{i,k} \left( \frac{\delta_{k,j}}{t_{k,k}} - \frac{1}{t_{..}} \right) \\
&= \left( \delta_{i,j} - 1 \right) \frac{t_{i,j}}{t_{j,j}} + \frac{2t_{i,i} - \Delta_i}{t_{..}},
\end{aligned} \tag{2.3}$$

and

$$w_{i,j} = \sum_{k=1}^{n} s_{i,k} v_{k,j} = \sum_{k=1}^{n} \left( \frac{\delta_{i,k}}{t_{i,i}} - \frac{1}{t_{..}} \right) \left[ \left( \delta_{k,j} - 1 \right) \frac{t_{k,j}}{t_{j,j}} + \frac{2t_{k,k} - \Delta_k}{t_{..}} \right]$$

$$= \sum_{k=1}^{n} \frac{\delta_{i,k}}{t_{i,i}} \left[ \left( \delta_{k,j} - 1 \right) \frac{t_{k,j}}{t_{j,j}} + \frac{2t_{k,k} - \Delta_k}{t_{..}} \right] - \frac{1}{t_{..}} \sum_{k=1}^{n} \left[ \left( \delta_{k,j} - 1 \right) \frac{t_{k,j}}{t_{j,j}} + \frac{2t_{k,k} - \Delta_k}{t_{..}} \right]$$

$$= \left[ \frac{\left( \delta_{i,j} - 1 \right)}{t_{i,i}} \left( \frac{t_{i,j}}{t_{j,j}} \right) + \frac{2t_{i,i} - \Delta_i}{t_{i,i}t_{..}} \right] - \frac{1}{t_{..}} \left[ \frac{-\left( t_{j,j} - \Delta_j \right)}{t_{j,j}} + 2 + \frac{\sum_k \Delta_k}{t_{..}} \right]$$

$$= \frac{\left( \delta_{i,j} - 1 \right) t_{i,j}}{t_{i,i} t_{j,j}} + \frac{1}{t_{..}} - \frac{\Delta_i}{t_{i,i}t_{..}} - \frac{\Delta_j}{t_{j,j} t_{..}} - \frac{\sum_k \Delta_k}{t_{..}^2} \right]$$

$$(2.4)$$

Furthermore, when  $i \neq j$ ,

$$0 < \frac{1}{t_{..}} \le \frac{1}{mn(n-1)},$$
  

$$0 < \frac{t_{i,j}}{t_{i,i}t_{j,j}} \le \frac{M}{m^2(n-1)^2},$$
  

$$0 < \frac{\Delta_i}{t_{i,i}t_{..}} \le \frac{M}{m^2n(n-1)^2},$$

and it is easy to show, when i, j, k are different from each other,

$$\begin{aligned} |w_{i,i}| &\leq \max\{\frac{1}{mn(n-1)}, \frac{3M}{m^2n(n-1)^2}\},\\ |w_{i,j}| &\leq \max\{\frac{1}{mn(n-1)}, \frac{M}{m^2(n-1)^2} + \frac{3M}{m^2n(n-1)^2}\},\\ |w_{i,j} - w_{i,k}| &\leq \frac{M}{m^2(n-1)^2} + \frac{M}{m^2n(n-1)^2},\\ |w_{i,i} - w_{i,k}| &\leq \frac{M}{m^2(n-1)^2} + \frac{M}{m^2n(n-1)^2}. \end{aligned}$$

It follows that

$$\max(|w_{i,j}|, |w_{i,j} - w_{i,k}|) \le \frac{M}{m^2(n-1)^2} + \frac{3M}{m^2n(n-1)^2} \quad \text{for all } i, j, k.$$
(2.5)

Next we use the recursion (2.2) to obtain a bound of the approximate error ||F||. By (2.2) and (2.3), for any *i*, we have

$$f_{i,j} = \sum_{k=1}^{n} f_{i,k} [(\delta_{k,j} - 1) \frac{t_{k,j}}{t_{j,j}} + \frac{2t_{k,k} - \Delta_k}{t_{..}}] + w_{i,j}, \quad j = 1, \cdots, n.$$
(2.6)

Thus, to prove Theorem 2.1, it is sufficient to show that  $|f_{i,j}| \leq C(M,m)/(n-1)^2$  for any i, j. Fixing any i, let  $f_{i,\alpha} = \max_{1 \leq k \leq n} f_{i,k}$  and  $f_{i,\beta} = \min_{1 \leq k \leq n} f_{i,k}$ .

First, we will show that  $f_{i,\beta} \leq 1/t_{..} \leq 1/(m(n-1)^2)$ . A direct calculation gives that

$$\sum_{k=1}^{n} f_{i,k} t_{k,i} = \sum_{k=1}^{n} (T_{i,k}^{-1} - (\frac{\delta_{i,k}}{t_{i,i}} - \frac{1}{t_{..}})) t_{k,i}$$
$$= 1 - (1 - \sum_{k=1}^{n} \frac{t_{k,i}}{t_{..}}) = \sum_{k=1}^{n} \frac{t_{k,i}}{t_{..}}.$$
(2.7)

Thus,  $f_{i,\beta} \sum_{k=1}^{n} t_{k,i} \leq \sum_{k=1}^{n} f_{i,k} t_{k,i} = \sum_{k=1}^{n} \frac{t_{k,i}}{t_{..}}$ . It follows that  $f_{i,\beta} \leq 1/t_{..}$  and, similarly,  $f_{i,\alpha} \geq 1/t_{..}$ .

 $1/t_{..}$ . Note that  $(1 - \Delta_{\alpha}/t_{\alpha,\alpha})f_{i,\beta} = -\sum_{k=1}^{n} f_{i,\beta}(\delta_{k,\alpha} - 1)\frac{t_{k,\alpha}}{t_{\alpha,\alpha}}$ . Thus,

$$f_{i,\alpha} + (1 - \frac{\Delta_{\alpha}}{t_{\alpha,\alpha}})f_{i,\beta} = \sum_{k=1}^{n} (f_{i,k} - f_{i,\beta})(\delta_{k,\alpha} - 1)\frac{t_{k,\alpha}}{t_{\alpha,\alpha}} + \sum_{k=1}^{n} f_{i,k}(\frac{2t_{k,k} - \Delta_{k}}{t_{\ldots}}) + w_{i,\alpha}.$$
 (2.8)

Similarly, by 
$$(1 - \Delta_{\beta}/t_{\beta,\beta})f_{i,\beta} = -\sum_{k=1}^{n} f_{i,\beta}(\delta_{k,\beta} - 1)\frac{t_{k,\beta}}{t_{\beta,\beta}}$$
, we have that

$$f_{i,\beta} + (1 - \frac{\Delta_{\beta}}{t_{\beta,\beta}})f_{i,\beta} = \sum_{k=1}^{n} (f_{i,k} - f_{i,\beta})(\delta_{k,\beta} - 1)\frac{t_{k,\beta}}{t_{\beta,\beta}} + \sum_{k=1}^{n} f_{i,k}(\frac{2t_{k,k} - \Delta_{k}}{t_{..}}) + w_{i,\beta}.$$
 (2.9)

Combining the above two equations, it yields

$$f_{i,\alpha} - f_{i,\beta} + \left(\frac{\Delta_{\beta}}{t_{\beta,\beta}} - \frac{\Delta_{\alpha}}{t_{\alpha,\alpha}}\right) f_{i,\beta}$$
  
=  $\sum_{k=1}^{n} (f_{i,k} - f_{i,\beta}) [(\delta_{k,\alpha} - 1) \frac{t_{k,\alpha}}{t_{\alpha,\alpha}} - (\delta_{k,\beta} - 1) \frac{t_{k,\beta}}{t_{\beta,\beta}}] + w_{i,\alpha} - w_{i,\beta}.$  (2.10)

Let  $\Omega = \{k : (1 - \delta_{k,\beta})t_{k,\beta}/t_{\beta,\beta} \ge (1 - \delta_{k,\alpha})t_{k,\alpha}/t_{\alpha,\alpha}\}$  and let  $|\Omega| = \lambda$ . Note that  $1 \le \lambda \le n-1$ . Then,

$$\sum_{k=1}^{n} (f_{i,k} - f_{i,\beta}) [(\delta_{k,\alpha} - 1) \frac{t_{k,\alpha}}{t_{\alpha,\alpha}} - (\delta_{k,\beta} - 1) \frac{t_{k,\beta}}{t_{\beta,\beta}}]$$

$$\leq \sum_{k\in\Omega} (f_{i,k} - f_{i,\beta}) [(1 - \delta_{k,\beta}) \frac{t_{k,\beta}}{t_{\beta,\beta}} - (1 - \delta_{k,\alpha}) \frac{t_{k,\alpha}}{t_{\alpha,\alpha}}]$$

$$\leq (f_{i,\alpha} - f_{i,\beta}) [\frac{\sum_{k\in\Omega} t_{k,\beta}}{t_{\beta,\beta}} - \frac{\sum_{k\in\Omega} (1 - \delta_{k,\alpha}) t_{k,\alpha}}{t_{\alpha,\alpha}}]$$

$$\leq (f_{i,\alpha} - f_{i,\beta}) [\frac{\lambda M}{\lambda M + (n - 1 - \lambda)m} - \frac{(\lambda - 1)m}{(\lambda - 1)m + (n - \lambda)M + M}].$$
(2.11)

Let

$$f(\lambda) = \frac{\lambda M}{\lambda M + (n-1-\lambda)m} - \frac{(\lambda-1)m}{(\lambda-1)m + (n-\lambda)M}.$$

There are two cases to consider the maximum of  $f(\lambda)$  in the range of  $\lambda \in [1, n-1]$ .

**Case I** When M = m, it is easy to show  $f(\lambda) = 1/(n-1)$ . **Case II** If  $M \neq m$ , since

$$f'(\lambda) = \frac{(n-1)Mm}{[\lambda M + (n-1-\lambda)m]^2} - \frac{(n-1)Mm}{[(\lambda-1)m + (n-\lambda)M]^2}$$
$$= \frac{(n-1)Mm[(n-2\lambda)(M-m)][\lambda M + (n-1-\lambda)m + (\lambda-1)m + (n-\lambda)M]}{[\lambda M + (n-1-\lambda)m]^2[(\lambda-1)m + (n-\lambda)M]^2}$$

and

$$f''(\lambda) = -2(M-m)Mm(n-1)\left(\frac{1}{[\lambda M + (n-1-\lambda)m]^3} + \frac{1}{[(\lambda-1)m + (n-\lambda)M]^3}\right),$$

 $f(\lambda)$  takes its maximum at  $\lambda = n/2$  when  $1 \le \lambda \le n-1$ . A direct calculation gives that

$$f(\frac{n}{2}) = \frac{nM - (n-2)m}{nM + (n-2)m}.$$
(2.12)

Moreover, denote

$$g(\lambda) = \frac{(\lambda - 1)m}{(\lambda - 1)m + (n - \lambda)M} - \frac{(\lambda - 1)m}{(\lambda - 1)m + (n - \lambda)M + M},$$

therefore

$$g'(\lambda) = \frac{Mm[M^2((n-\lambda)^2 + 2(n-\lambda)(\lambda-1) + n-1) + (2Mm-m^2)(\lambda-1)^2]}{[(\lambda-1)m + (n-\lambda)M]^2[(\lambda-1)m + (n-\lambda)M + M]^2},$$

 $g'(\lambda) > 0$  when  $1 \le \lambda \le n - 1$  such that for  $1 \le \lambda \le n - 1$ ,

$$0 \le g(\lambda) \le g(n-1) = \frac{(n-2)Mm}{[(n-2)m+M][(n-2)m+2M]}.$$
(2.13)

By (2.12) and (2.13), we have

$$\max_{1 \le \lambda \le n-1} \left[ \frac{\lambda M}{\lambda M + (n-1-\lambda)m} - \frac{(\lambda-1)m}{(\lambda-1)m + (n-\lambda)M + M} \right] \\
\le \max_{1 \le \lambda \le n-1} f(\lambda) + \max_{1 \le \lambda \le n-1} g(\lambda) \\
\le \frac{1}{n-1} I(M=m) + \frac{nM - (n-2)m}{nM + (n-2)m} I(M \ne m) + \frac{(n-2)Mm}{[(n-2)m + M][(n-2)m + 2M]} \\
= \frac{nM - (n-2)m}{nM + (n-2)m} + \frac{(n-2)Mm}{[(n-2)m + M][(n-2)m + 2M]},$$
(2.14)

where  $I(\cdot)$  is an indictor function. Since  $f_{i,\alpha} - f_{i,\beta} + \frac{1}{t_{\cdot}} \ge |f_{i,\beta}|$ , we have

$$f_{i,\alpha} - f_{i,\beta} + \left(\frac{\Delta_{\beta}}{t_{\beta,\beta}} - \frac{\Delta_{\alpha}}{t_{\alpha,\alpha}}\right) f_{i,\beta}$$

$$\geq f_{i,\alpha} - f_{i,\beta} - \left(f_{i,\alpha} - f_{i,\beta} + \frac{1}{t_{..}}\right) \left| \frac{\Delta_{\beta}}{t_{\beta,\beta}} - \frac{\Delta_{\alpha}}{t_{\alpha,\alpha}} \right|$$

$$\geq \left(1 - \frac{M}{m(n-1)}\right) \left(f_{i,\alpha} - f_{i,\beta}\right) - \frac{M}{m^2 n(n-1)^2}.$$
(2.15)

Combining (2.10), (2.11), (2.14) and (2.15), it yields

$$(1 - \frac{M}{m(n-1)})(f_{i,\alpha} - f_{i,\beta})$$

$$\leq \left(\frac{nM - (n-2)m}{nM + (n-2)m} + \frac{(n-2)Mm}{[(n-2)m + M][(n-2)m + 2M]}\right) \times (f_{i,\alpha} - f_{i,\beta}) + \frac{M}{m^2(n-1)^2} + \frac{4M}{m^2n(n-1)^2},$$

so that

$$C(m, M)(f_{i,\alpha} - f_{i,\beta}) \le \frac{M}{m^2(n-1)^2} + \frac{4M}{m^2n(n-1)^2}$$

where C(m, M) is defined in (2.1). Consequently, if the condition (2.1) holds, then

$$\begin{aligned} \max_{j=1,\cdots,n} |f_{i,j}| &\leq f_{i,\alpha} - f_{i,\beta} + \frac{1}{t_{\cdots}} \\ &\leq \frac{1}{C(m,M)} \times \left[ \frac{M}{m^2(n-1)^2} + \frac{4M}{m^2n(n-1)^2} \right] + \frac{1}{mn(n-1)} \\ &= \frac{1}{(n-1)^2} \times \left[ \frac{1}{C(m,M)} \left( \frac{M}{m^2} + \frac{4M}{m^2n} \right) + \frac{n-1}{mn} \right], \end{aligned}$$

where the first inequality holds by  $\max_{j} |f_{i,j}| \leq f_{i,\alpha} - f_{i,\beta} + f_{i,\beta}I(f_{i,\beta} > 0)$  and  $0 \leq f_{i,\beta}I(f_{i,\beta} > 0) \leq 1/t_{..}$ . This completes the proof.

**Remark** If M and m are constants,  $C(m, M) \approx 2m/(M+m)$  when n is large enough. Therefore the condition (2.1) is very week.

#### **3** Discussion

Our proposed matrices S could be used as preconditioners for solving linear systems with diagonally dominant and non-negative matrices just as [18, 20] concerned with M-matrices. The bound on the approximation error in Theorem 2.1 depends on m, M and n. When m and M are bounded by a constant, all the elements of  $T^{-1} - S$  are of order  $O(1/(n-1)^2)$  as  $n \to \infty$ , uniformly.

Finally, we illustrate by an example that the bound on the approximation error in Theorem 2.1 is optimal in the sense that any bound in the form of C(m, M)/f(n) requires  $f(n) = O((n-1)^2)$  as  $n \to \infty$ . Assume that the matrix T consists of the elements:  $t_{i,i} = (n-1)M, i = 1, \dots, n-1; t_{n,n} = (n-1)m$  and  $t_{i,j} = m, i, j = 1, \dots, n; i \neq j$ , which satisfies (1.1). By the Sherman-Morrison formula, we have

$$(T^{-1})_{i,j} = \frac{\delta_{i,j}}{(n-1)M - m} - \frac{m}{[(n-1)M - m]^2}, i, j = 1, \cdots, n-1$$
  
$$(T^{-1})_{n,j} = \frac{\delta_{n,j}}{(n-2)m} - \frac{1}{(n-2)[(n-1)M - m]}, \quad j = 1, \cdots, n.$$

In this case, the elements of S are

$$S_{i,j} = \frac{\delta_{i,j}}{(n-1)M} - \frac{1}{n(n-1)m}, \ i, j = 1, \cdots, n-1; i \neq j,$$
  
$$S_{n,j} = \frac{\delta_{n,j}}{(n-1)m} - \frac{1}{n(n-1)m}, \ j = 1, \cdots, n.$$

It is easy to show that the bound of  $||T^{-1} - S||$  is  $O(\frac{1}{(n-1)^2m})$ . This suggests that the rate  $1/(n-1)^2$  is optimal. On the other hand, if M and m are constants, the upper bound of  $||T^{-1} - S||$  approximately equal to  $(1 + \frac{M}{m})\frac{M}{2(n-1)^2m^2}$ . Therefore there is a gap between these two bounds that implies there might be space for improvement. It is interesting to see if the bounds in Theorem 2.1 can be further relaxed.

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# 关于非负对称矩阵的近似逆矩阵

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**摘要:** 本文研究了非负对称矩阵的近似逆矩阵问.利用矩阵*S* = (*s<sub>i,j</sub>*) 去近似它的逆矩阵方法,获得 了近似误差的一个显式上界,并且证明了近似逆的误差对于很大的*n*一致地具有阶1/(*n* – 1)<sup>2</sup>. 关键词: 近似误差;逆;对称;非负元

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