
NUMERICAL LINEAR ALGEBRA: Solutions for Sheet 3

1. Suppose $Ax = b$ and $(A + \delta A)(x + \delta x) = b$, which together imply $(A + \delta A)(x + \delta x) = Ax$. Multiplying this out gives $Ax + A\delta x + \delta A(x + \delta x) = Ax$ and consequently $A\delta x = -\delta A(x + \delta x)$. Assuming A is invertible, $\delta x = -A^{-1}(\delta A)(x + \delta x)$. We now apply norm inequalities to obtain

$$\|\delta x\| = \|A^{-1}(\delta A)(x + \delta x)\| \leq \|A^{-1}\| \|\delta A\| \|x + \delta x\| = \frac{\|A\|}{\|A\|} \|A^{-1}\| \|\delta A\| \|x + \delta x\|.$$

Dividing by $\|x + \delta x\|$ gives the desired result,

$$\frac{\|\delta x\|}{\|x + \delta x\|} \leq \|A\| \|A^{-1}\| \frac{\|\delta A\|}{\|A\|}.$$

This bounds the relative perturbation of the solution by the relative perturbation in the matrix, magnified by the condition number $\kappa(A) = \|A\| \|A^{-1}\|$. Small errors in A might lead to large errors in the computed solution.

2. We have

$$\kappa(Q) = \|Q\| \|Q^{-1}\| = \|Q\| \|Q^T\|$$

and since $\|Z\| = 1$ for an orthogonal matrix Z we get $\kappa(Q) = 1$. The Cholesky factorisation of $A^T A$ is given by $A^T A = LL^T$ where L is a lower triangular matrix. Now consider the QR factorisation of A , i.e. $A = QR$. Then we get

$$A^T A = (QR)^T QR = R^T Q^T QR = R^T R.$$

If

$$R = \begin{bmatrix} \hat{R} \\ 0 \end{bmatrix}$$

with $\hat{R} \in \mathbb{R}^{n \times n}$ and $\hat{R}^T \hat{R} = R^T R$. So \hat{R}^T could be L^T . In fact this precisely depends on the choices of sign of the diagonal entries in the Cholesky and QR algorithm.

3. Hilbert matrix

size	condition number
4	1.5514e + 04
8	1.5258e + 10
12	1.8032e + 16

Result for $x = A \backslash b$ where $b = \text{sum}(A, 2)$ with solution $x = [1, \dots, 1]^T$ gives

$$x = \begin{bmatrix} 0.999999976736699 \\ 1.000002910963504 \\ 0.999909292281920 \\ 1.001227445722276 \\ 0.991048902442979 \\ 1.039167445001900 \\ 0.891228720678228 \\ 1.196361444783960 \\ 0.770302875862095 \\ 1.167911583742747 \\ 0.930297943531136 \\ 1.012541473056602 \end{bmatrix}.$$

4. One Gauss–Seidel iteration applied to the system $Ax = b$ is

for $i = 1, \dots, n$

$$x_i^{(k+1)} = \left(b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^n a_{ij} x_j^{(k)} \right) / a_{ii}.$$

Since A is lower triangular, $a_{ij} = 0$ if $j > i$. Thus, the Gauss–Seidel iteration reduces to

for $i = 1, \dots, n$

$$x_i^{(k+1)} = \left(b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} \right) / a_{ii}.$$

Notice that this is identical to forward substitution:

$$x_1 = b_1/a_{11}, \quad x_2 = (b_2 - x_1 a_{21})/a_{22}, \quad \dots, \quad x_n = \left(b_n - \sum_{j=1}^{n-1} a_{nj} x_j \right) / a_{nn}$$

Thus, Gauss–Seidel will solve a lower triangular matrix *exactly* in only one iteration.

Suppose A is nearly upper triangular. Then we would want an iterative method for which one step is similar to backwards substitution. Such a method executes the Gauss–Seidel iteration in reverse order:

for $i = n, \dots, 1$

$$x_i^{(k+1)} = \left(b_i - \sum_{j=i+1}^n a_{ij} x_j^{(k+1)} - \sum_{j=1}^{i-1} a_{ij} x_j^{(k)} \right) / a_{ii}.$$

5. The *spectral radius* is defined as $\rho(A) := \max_{\lambda \in \Lambda(A)} |\lambda|$, where $\Lambda(A)$ is the spectrum of A . Let λ_{\max} be an eigenvalue of A such that $\rho(A) = |\lambda_{\max}|$. Then there must be a non-trivial eigenvector v associated with λ_{\max} . As eigenvectors are not scale-dependent, we can pick v such that $\|v\| = 1$, where $\|\cdot\|$ is the vector norm which induces the operator norm. Then

$$\rho(A) = |\lambda_{\max}| = |\lambda_{\max}| \|v\| = \|\lambda_{\max} v\| = \|Av\| \leq \max_{\|x\|=1} \|Ax\| = \|A\|.$$

Thus $\rho(A) \leq \|A\|$ for any operator norm A .

Does this hold if $\|\cdot\|$ is not an operator norm (e.g., the Frobenius norm)? Can you prove this?

6. Let λ be an eigenvalue of $A \in \mathbb{R}^{n \times n}$ such that $(A - \lambda I)x = 0$ for some $x \neq 0$. Take k such that $|x_k| = \max_i |x_i|$. Then the k th row of the equation $(A - \lambda I)x = 0$ gives

$$(a_{kk} - \lambda)x_k = - \sum_{j=1, j \neq k}^n a_{kj} x_j.$$

Thus,

$$|a_{kk} - \lambda| |x_k| = \left| \sum_{j=1, j \neq k}^n a_{kj} x_j \right| \leq \sum_{j=1, j \neq k}^n |a_{kj}| |x_j|.$$

Dividing through by $|x_k|$ (which must be greater than zero since $x \neq 0$), we have

$$|a_{kk} - \lambda| \leq \sum_{j=1, j \neq k}^n |a_{kj}| \frac{|x_j|}{|x_k|} \leq \sum_{j=1, j \neq k}^n |a_{kj}|.$$

Having proved Gerschgorin's Theorem, we will apply it to bound the spectral radius of the Jacobi iteration matrix. Recall that the Jacobi splitting of $A = L + D + U$ (L strictly lower triangular, D diagonal, U strictly upper triangular) yields the iteration matrix $M^{-1}N = D^{-1}(-L - U)$. Thus,

$$A = \begin{pmatrix} 9 & 1 & 1 \\ 1 & 5 & 0 \\ 1 & 1 & 3 \end{pmatrix}, \quad M = \begin{pmatrix} 9 & 0 & 0 \\ 0 & 5 & 0 \\ 0 & 0 & 3 \end{pmatrix}, \quad N = \begin{pmatrix} 0 & -1 & -1 \\ -1 & 0 & 0 \\ -1 & -1 & 0 \end{pmatrix}.$$

The Jacobi iteration matrix is thus

$$M^{-1}N = \begin{pmatrix} 0 & -\frac{1}{9} & -\frac{1}{9} \\ -\frac{1}{5} & 0 & 0 \\ -\frac{1}{3} & -\frac{1}{3} & 0 \end{pmatrix}$$

Applying Gerschgorin's Theorem to each row yields three bounds:

$$|\lambda| \leq 2/9, \quad |\lambda| \leq 1/5, \quad |\lambda| \leq 2/3.$$

From this we deduce that $\rho(M^{-1}N) \leq 2/3$.

But we can get a tighter bound by using the result from Question 2. Calculate that $\|M^{-1}N\|_1 = 8/15 < 2/3$. Thus, $\rho(M^{-1}N) \leq 8/15$. In fact, we find $\rho(M^{-1}N) \approx 0.2910$.

7. Define A to be the upper triangular matrix with $1/2$ on the main diagonal ones in all superdiagonal entries. This is an example of a *non-normal* matrix. All the eigenvalues are $1/2$, so $\rho(A) = 1/2$. Thus, we must have $\|A^k\|_\infty \rightarrow 0$ as $k \rightarrow \infty$. But there is growth of A^k for transient values of k , and this can have wide-ranging implications. For example, it warns us that iterative methods may not converge very nicely, since $\|x^{(k)} - x\| \leq \|(M^{-1}N)^k\| \|x^{(0)} - x\|$.

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A = triu(ones(10)) - 1/2*eye(10);
Ak = A;
for k=1:50,
    normAk(k) = norm(Ak,'inf');
    Ak=Ak*A;
end

semilogy([1:50], normAk);
xlabel('power, k', 'fontsize', 14)
ylabel('||A||_\infty', 'fontsize', 14)
title('infinity norms of matrix powers', 'fontsize', 14)
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8. Let $Ax = b$, and let $A = D + L + U$. Define the SOR method by:

$$(D + \omega L) x^{(k+\frac{1}{2})} = \omega b + ((1 - \omega) D - \omega U) x^{(k)}, \quad (1)$$

$$(D + \omega U) x^{(k+1)} = \omega b + ((1 - \omega) D - \omega L) x^{(k+\frac{1}{2})} \quad (2)$$

From the 1st equation, we obtain that:

$$x^{(k+\frac{1}{2})} = \omega (D + \omega L)^{-1} b + (D + \omega L)^{-1} ((1 - \omega) D - \omega U) x^{(k)}, \quad (3)$$

Substituting the value of $x^{(k+\frac{1}{2})}$ into the 2nd equation we have

$$x^{(k+1)} = (D + \omega U)^{-1} ((1 - \omega) D - \omega L) \left((D + \omega L)^{-1} ((1 - \omega) D - \omega U) \right) x^{(k)} \quad (4)$$

$$+ \omega (D + \omega U)^{-1} ((1 - \omega) D - \omega L) (D + \omega L)^{-1} b + \omega (D + \omega U)^{-1} b \quad (5)$$

It is clear that

$$G = (D + \omega U)^{-1} ((1 - \omega) D - \omega L) (D + \omega L)^{-1} ((1 - \omega) D - \omega U)$$

is the iteration matrix. One can show that if

$$A = M - N \iff N = M - A \iff M^{-1}N = I - M^{-1}A \quad (6)$$

Hence, it is enough to show that $G = M^{-1}N$ as given above can be expressed as $G = I - M^{-1}A$ where M is the symmetric matrix to be determined. It follows that

$$(1 - \omega)D - \omega L = -(D + \omega L) + (2 - \omega)D$$

and similarly

$$(1 - \omega)D - \omega U = -(D + \omega U) + (2 - \omega)D$$

Then G can be transformed to:

$$G = (D + \omega U)^{-1} ((1 - \omega)D - \omega L) (D + \omega L)^{-1} ((1 - \omega)D - \omega U) \quad (7)$$

$$= -(D + \omega U)^{-1} ((D + \omega L) ((D + \omega L)^{-1})) ((1 - \omega)D - \omega U) \quad (8)$$

$$+ (2 - \omega) (D + \omega U)^{-1} D (D + \omega L)^{-1} ((1 - \omega)D - \omega U) \quad (9)$$

$$= (D + \omega U)^{-1} (D + \omega U) - (2 - \omega) (D + \omega U)^{-1} D \quad (10)$$

$$+ (2 - \omega) (D + \omega U)^{-1} D (D + \omega L)^{-1} ((1 - \omega)D - \omega U) \quad (11)$$

$$= I - (2 - \omega) (D + \omega U)^{-1} D (I - (D + \omega L)^{-1} ((1 - \omega)D - \omega U)) \quad (12)$$

$$= I - (2 - \omega) (D + \omega U)^{-1} D (D + \omega L)^{-1} (D + \omega L - (1 - \omega)D + \omega U) \quad (13)$$

$$= I - \omega (2 - \omega) (D + \omega U)^{-1} D (D + \omega L)^{-1} A \quad (14)$$

So M for this splitting is given by

$$M = \frac{1}{\omega(2 - \omega)} (D + \omega L) D^{-1} (D + \omega U).$$

9. (or see solution for question 10) Writing $J = M^{-1}N$ the Jacobi iteration matrix, we have $M = \text{diag}(A)$ and $N = M - A$ so

$$M = \begin{pmatrix} a_{11} & & & \\ & a_{22} & & \\ & & \ddots & \\ & & & a_{nn} \end{pmatrix}, \quad M^{-1} = \begin{pmatrix} \frac{1}{a_{11}} & & & \\ & \frac{1}{a_{22}} & & \\ & & \ddots & \\ & & & \frac{1}{a_{nn}} \end{pmatrix}, \quad N = \begin{pmatrix} 0 & a_{12} & \cdots & a_{1n} \\ a_{21} & 0 & \ddots & \vdots \\ \vdots & \ddots & \ddots & a_{nn-1} \\ a_{n1} & \cdots & a_{n-1n} & 0 \end{pmatrix}$$

and therefore

$$J = \begin{pmatrix} 0 & \frac{a_{12}}{a_{11}} & \cdots & \frac{a_{1n}}{a_{11}} \\ \frac{a_{21}}{a_{22}} & 0 & \ddots & \vdots \\ \vdots & \ddots & \ddots & \frac{a_{nn-1}}{a_{n-1n-1}} \\ \frac{a_{n1}}{a_{nn}} & \cdots & \frac{a_{n-1n}}{a_{nn}} & 0 \end{pmatrix}$$

We now apply Gershgorin's theorem as in question 6, and we have that for all eigenvalues λ there exists some i , such that

$$|\lambda| = |\lambda - 0| \leq \sum_{j=1, j \neq i}^n \left| \frac{a_{ij}}{a_{ii}} \right| < 1 \quad (15)$$

by SRDD of A . Therefore all eigenvalues of J have modulus strictly less than one, and hence the Jacobi iteration converges.

10. We want to prove that if A is SRDD then both the Jacobi and Gauss-Seidel iterations converge. It is clear that if a matrix is SRDD then it is non-singular. By Gershgorin Circle Theorem its spectral radius does not contain the origin hence it is non-singular (see Question 6).

Let λ be the eigenvalue for both the Jacobi and Gauss-Seidal iteration method. For the Jacobi iteration method it means that:

$$-D^{-1} (L + U) \underline{x} = \lambda \underline{x}$$

$$(L + U) \underline{x} = \lambda D \underline{x}$$

$$(L + U + \lambda D) \underline{x} = 0$$

Similarly for the G-S iteration method:

$$(U + \lambda L + \lambda D) \underline{x} = 0.$$

Now let us assume in both cases that there exists λ such that $|\lambda| \geq 1$. As λ is greater than equal to 1 then A becomes even more diagonally dominant as the term λD increases. Hence $(L + U + \lambda D)$ or $(U + \lambda L + \lambda D)$ are singular. But this contradicts the fact that A is non-singular. Therefore $|\lambda| < 1$. Therefore all eigenvalues of the Jacobi and Gauss-Seidal lie within the unit disc, hence the spectral radius is strictly less than 1, and both methods converge independent of b and $x^{(0)}$.