

Numerical linear algebra

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1 Some key concepts

- Norms
- Singular values
- triangular matrix
- SVD and QR Decomposition
- stability

2 Expanding the definition of Norm from vectors to matrix

Three properties of vector norm:

- $\|x\| \geq 0$, and zero value achieved iff $x = 0$
- $\|x + y\| \leq \|x\| + \|y\|$
- $\|\alpha x\| = |\alpha| \|x\|$, α is constant

Besides the Euclidean norm we are familiar with, we may expand the definition of norm suppose p is a real number, for $\vec{x} \in \mathbf{C}^m$:

$$\|x\|_p = \left(\sum_i^m |x_i|^p \right)^{1/p}$$

Based on the general definition of vector norm, we can define the induced matrix norm as:

$$\|A\|_{m,n} = \sup_{x \in \mathbf{C}^n, x \neq 0} \frac{\|Ax\|_m}{\|x\|_n}$$

The induced matrix norm would also have three properties of vector norm.

2.1 Frobenius norm

Frobenius norm is a commonly used matrix norm, but it is not a induced norm. we define the a_j to be the j th col of matrix A , we have

$$\|A\|_F = \left(\sum_{j=1}^n \|a_j\|^2 \right)^{.5}$$

or the Frobenius norm can also be found with trace

$$\|A\|_F = \sqrt{\text{tr}(A^\dagger A)} = \sqrt{\text{tr}(AA^\dagger)}$$

2.2 usage of matrix norms

We may use the matrix norm to bound $\|AB\|$ with either induced norm or Frobenius norm. Recall the Cauchy-Schwarz inequality

$$|x^\dagger y| \leq \|x\|_p \|y\|_q$$

where $1/p + 1/q = 1$.

For induced norm we have:

$$\|AB\|_{l,n} \leq \|A\|_{l,m} \|B\|_{m,n}$$

and for Frobenius norm

$$\|AB\|_F^2 = \|A\|_F^2 \|B\|_F^2$$

3 singular values

We define the eigen values of a square matrix, and we are trying to expand this concept into a general $m \times n$ matrix. We define $\{\sigma_i\}$ to be the singular values of matrix A , and σ_i would be the i th eigenvalue of $A^\dagger A$, which is guaranteed to be a square matrix. (Assuming $m \geq n$) Denote Σ to be a $n \times n$ diagonal matrix, and U to be a $m \times n$ matrix with orthonormal columns, and V to be a $n \times n$ matrix with orthonormal columns. we have a SVD decomposition of matrix A such that:

$$A = U\Sigma V$$

, the SVD would be unique up to any unitary operations applied to col space of U and V . One may also expand the reduced SVD into full SVD by adding more zeros to make U to be $m \times m$, Σ to be $m \times n$, and V to be $n \times n$.

Numerical linear algebra, Lloyd N. Trefethen; David Bau III

QR factorization:

pg 51 CGS;

pg 58 MGS, more stable;
pg 73 Householder, need to construct Q separately;
pg 98 Householder reduction to Hessenberg matrix;