

Special Matrices

Diagonally Dominant

A matrix is **diagonally dominant** when

$$|a_{ii}| \geq \sum_{j=1, j \neq i}^n |a_{ij}|$$

holds for each $i = 1, 2, \dots, n$.

It is **strictly diagonally dominant** when,

$$|a_{ii}| > \sum_{j=1, j \neq i}^n |a_{ij}|$$

Symmetric Positive Definite Matrices (SPD)

A matrix A is **symmetric positive definite** if it is symmetric and if $x^t A x > 0$ for every n -dimensional vector $x \neq 0$.

If A is an $n \times n$ SPD matrix, then

- A has an inverse
- $a_{ii} > 0$ for each $i = 1, 2, \dots, n$
- $\max_{1 \leq k, j \leq n} |a_{kj}| \leq \max_{1 \leq i \leq n} |a_{ii}|$
- $(a_{ij})^2 < a_{ii} a_{jj}$, for each $i \leq j$

The matrix A is positive definite if and only if A can be factored in the form LDL^t , where L is lower triangular with 1s on its diagonal and D is a diagonal matrix with positive diagonal entries.

The matrix A is positive definite if and only if A can be factored in the form LL^t , where L is lower triangular with nonzero diagonal entries.

Cholesky (LL^T)

Tridiagonal LU Factorization

Vector Norms

A vector norm on R^n is a function, $\|\cdot\|$, from R^n into R with the following properties:

- $\|x\| \geq 0$ for all $x \in R^n$
- $\|x\| = 0$ if and only if $x = 0$
- $\|\alpha x\| = |\alpha| \|x\|$ for all $\alpha \in R, x \in R^n$
- $\|x + y\| \leq \|x\| + \|y\|$ for all $x, y \in R^n$

The l_2 norm is defined as such:

$$||x||_2 = \left\{ \sum_{i=1}^n x_i^2 \right\}^{\frac{1}{2}}$$

The l_∞ norm is defined as such:

$$||x||_\infty = \max_{1 \leq i \leq n} |x_i|$$