



Computational Science:  
Computational Methods in Engineering

## Preconditioning



## Condition Number

Condition number of a matrix  $[A]$  measures how much the answer to a linear algebra problem changes due to small changes in  $[A]$ .

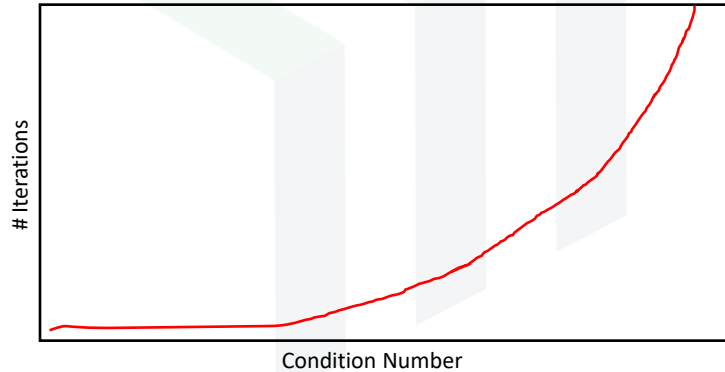
The condition number can be calculated from the norms of the matrix.

The condition number of a matrix  $[A]$  is defined as  $\text{cond}(A) = \|A\| \|A^{-1}\|$

A large condition number means the system is **ill conditioned** and a small condition number means the system is **well conditioned**.

## Problem with ill-Conditioned Matrices

The number of iterations needed for an iterative solver to find a solution to  $[A][x] = [b]$  depends heavily on the condition number of the matrix  $[A]$ .



## Preconditioning

Suppose  $[A][x] = [b]$  is to be solved iteratively, but  $[A]$  has a high condition number.

A preconditioner  $[P]$  is a nonsingular matrix such any of the following have a smaller condition number than  $[A]$  alone.

$$[P]^{-1}[A] \quad \text{Left Preconditioned}$$

$$[A][P]^{-1} \quad \text{Right Preconditioned}$$

$$[P]^{-1}[A]([P]^{-1})^T \quad \text{Split Preconditioned}$$

The ideal preconditioner transforms  $[A]$  into an identity matrix and makes any iterative method converge in one iteration.

## Right Preconditioning

### Formulation

Original Equation

$$[A][x] = [b]$$

Insert Preconditioner

$$[A][P]^{-1}[P][x] = [b]$$

Derive Preconditioned Equation

$$[A'][y] = [b]$$

$$[A'] = [A][P]^{-1} \quad [y] = [P][x]$$

### Implementation

1. Solve  $[y] = [A']^{-1}[b]$
2. Calculate  $[x] = [P]^{-1}[y]$

### Iterative Solvers

1. Least-squares (LSQR)
2. Transpose-free quasi-minimal residual (TFQMR)
3. Stabilized biconjugate gradients (BICGSTAB)
4. Stabilized biconjugate gradients (I) (BICGSTABL)
5. Conjugate gradient square (CGS)

## Left Preconditioning

### Formulation

Original Equation

$$[A][x] = [b]$$

Insert Preconditioner

$$[P]^{-1}[A][x] = [P]^{-1}[b]$$

Derive Preconditioned Equation

$$[A'][x] = [b']$$

$$[A'] = [P]^{-1}[A] \quad [b'] = [P]^{-1}[b]$$

### Implementation

1. Solve  $[x] = [A']^{-1}[b']$

### Iterative Solvers

1. Generalized minimum residual (GMRES)
2. Quasi-minimal residual (QMR)
3. Biconjugate gradients (BICG)

## Split Preconditioning

### Formulation

Original Equation

$$[A][x] = [b]$$

Insert Preconditioner

$$[P]^{-1}[A][P]^{-T}[P]^T[x] = [P]^{-1}[b]$$

Derive Preconditioned Equation

$$[A'] [y] = [b']$$

$$\left. \begin{aligned} [A'] &= [P]^{-1}[A][P]^{-T} \\ [b'] &= [P]^{-1}[b] \\ [y] &= [P]^T[x] \end{aligned} \right\}$$

### Implementation

1. Solve  $[y] = [A']^{-1}[b']$
2. Calculate  $[x] = [P]^{-T}[y]$

### Iterative Solvers

1. Preconditioned conjugate gradients (PCG)
2. Minimum residual (MINRES)
3. Symmetric LQ (SYMMLQ)

## Jacobi (or Diagonal) Preconditioner

Perhaps the simplest preconditioner is the Jacobi preconditioner. In this case, the preconditioner  $[P]$  is simply the diagonal of the matrix  $[A]$ .

$$[P] = \text{diag}([A]) \quad [A] = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ a_{N1} & a_{N2} & \cdots & a_{NN} \end{bmatrix} \quad [P] = \begin{bmatrix} a_{11} & & & \\ & a_{22} & & \\ & & \ddots & \\ & & & a_{NN} \end{bmatrix}$$

Jacobi preconditioning is particularly effective for diagonally dominant matrices.

# Notes About Preconditioning

- The preconditioner  $[P]$  is chosen to accelerate the convergence of an iterative method.
- Preconditioning is needed when little progress is made between iterations or the residual error of an iterative solution stagnates.
- When the numbers in  $[A]$  have massively different magnitudes (i.e. 2-3 orders of magnitude or more),  $[A]$  will typically require a preconditioner.
- Incomplete factorizations, such as the incomplete LU factorization or the incomplete Cholesky factorization, are often used as preconditioners when natural preconditioners are absent.