

Iteration Methods for Linear Systems

- Many linear systems arised in real-world applications are large and sparse.
 - ◊ Storage of data on a computer becomes a serious concern.
 - ◊ Overhead in a direct method, such as the Gaussian elimination, usually becomes unbearable.
 - ◊ Special techniques taking into account sparsity preservation, such as SPARSPAK, are available.
 - ◊ Would like to solve $Ax = b$ with the matrix A intact and storage of only a few vectors.
- Motiviatiion of an Iterative Method:
 - ◊ Let \tilde{x} be an approximate solution to the system $Ax = b$.
 - ◊ Define the *residual* by

$$r := b - A\tilde{x}.$$

- ◊ The error $e := x - \tilde{x}$ satisfies the equation

$$Ae = r. \tag{0.1}$$

- ◊ If we could solve (0.1) exactly, then $x := \tilde{x} + e$ would be the exact solution.
 - ▷ Note that solving (0.1) is as hard as the original problem.
 - ▷ Instead, we solve

$$Se = r \tag{0.2}$$

where S is an approximation to A .

- ▷ The difference between A and S here is that (0.2) is much easier to be solved than (0.1).

▷ Adding approximate correction to the approximate solution \tilde{x} gives what we hope is a better approximate to the the true solution.

- A Prototype Iterative Algorithm:

- ◇ x^{old} : = The current approximation to x ;

- ◇ Compute the residual $r := b - Ax^{\text{old}}$;

- ◇ Solve $Se = r$ for the unknown e ;

- ◇ Set

$$x^{\text{new}} := x^{\text{old}} + e; \quad (0.3)$$

- ◇ Repeat the cycle.

- Important Observation:

- ◇ Multiplying (0.3) by S yields

$$\begin{aligned} Sx^{\text{new}} &= Sx^{\text{old}} + Se \\ &= Sx^{\text{old}} + b - Ax^{\text{old}} \\ &= (S - A)x^{\text{old}} + b := Tx^{\text{old}} + b \end{aligned} \quad (0.4)$$

where

$$A := S - T \quad (0.5)$$

is called a *splitting* of the matrix A .

- ◇ If the iterates converges to a limit x , then $x = x^{\text{old}} = x^{\text{new}}$ and, by (0.4), we see that $Ax = b$. In other words, a limit point of the iteration scheme (0.4) is a solution of the original system.

- ◇ It remains to determine the splitting so that

- ▷ The sequence generated will converge.

- ▷ The convergence rate is faster.