## **Iterative Linear Algebra Methods**

The following are outlines of the best-known iterative methods for approximately solving Ax = b, where  $A \in \mathbb{R}^{n \times n}$ ,  $x \in \mathbb{R}^n$ , and  $b \in \mathbb{R}^n$ .

#### 1. The classical iterative methods: Jacobi, Gauss-Seidel, and SOR.

Notation: In describing the matrix-vector forms of the classical iterative methods, we use the decomposition A=L+D+U, in which L, D, and U are the strict lower-triangular, diagonal, and strict upper-triangular parts of A, respectively. In describing the componentwise forms, we denote the ith components of x and y and the y and y are the strict lower-triangular, diagonal, and strict upper-triangular parts of y and y are the strict lower-triangular, diagonal, and strict upper-triangular parts of y and y and y and y are the strict lower-triangular, diagonal, and strict upper-triangular parts of y and y and y are the strict lower-triangular, diagonal, and strict upper-triangular parts of y and y are the strict lower-triangular, diagonal, and strict upper-triangular parts of y and y are the strict lower-triangular, diagonal, and strict upper-triangular parts of y and y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lower-triangular parts of y and y are the strict lo

#### JACOBI ITERATION: (matrix-vector form)

Given A, b, and initial x.

Until "stop":

$$\mathsf{Update}\ x \leftarrow D^{-1}\big[b - (L+U)x\big].$$

### JACOBI ITERATION: (componentwise form)

Given A, b, and initial x.

Until "stop":

For 
$$i = 1, \ldots, n$$
:

Set 
$$x_i^+ = (b_i - \sum_{j \neq i} A_{ij} x_j) / A_{ii}$$
.

 $\mathsf{Update}\ x \leftarrow x^+.$ 

## Gauss-Seidel Iteration: (matrix-vector form)

Given A, b, and initial x.

Until "stop":

Update  $x \leftarrow (L+D)^{-1}(b-Ux)$ .

### GAUSS-SEIDEL ITERATION: (componentwise form)

Given A, b, and initial x.

Until "stop":

For 
$$i = 1, ..., n$$
:

Update 
$$x_i \leftarrow (b_i - \sum_{j \neq i} A_{ij} x_j) / A_{ii}$$
.

### SUCCESSIVE OVERRELAXATION (SOR): (matrix-vector form)

Given A, b, initial x, and  $\omega$ .

Until "stop":

Update 
$$x \leftarrow (\omega L + D)^{-1} \{ \omega b + [(1 - \omega)D - \omega U]x \}.$$

### SUCCESSIVE OVERRELAXATION (SOR): (componentwise form)

Given A, b, initial x, and  $\omega$ .

Until "stop":

For 
$$i = 1, \ldots, n$$
:

Update 
$$x_i \leftarrow (1 - \omega)x_i + \omega \left(b_i - \sum_{j \neq i} A_{ij}x_j\right) / A_{ii}$$
.

#### 2. Krylov subspace methods: GMRES(m), CG, and PCG.

Notation: In GMRES(m),  $e_1 = (1, 0, \dots, 0)^T \in \mathbb{R}^{m+1}$  and  $w_{k+1}$  is the (k+1)st component of the vector  $w \in \mathbb{R}^{m+1}$ . Generally, subscripted quantities may denote scalars, vectors, or matrices, depending on the context.

 $\mathrm{GMRES}(m)$ : (standard Gram–Schmidt implementation)

Given A, b, x, tol, itmax.

INITIALIZE: Set  $r \equiv b - Ax$ ,  $v_1 \equiv r/||r||_2$ ,  $w \equiv ||r||_2 e_1 \in \mathbb{R}^{m+1}$ .

ITERATE: For k = 1, ..., m, do:

Initialize  $v_{k+1} = Av_k$ .

For  $i = 1, \ldots, k$ , do:

Set  $h_{ik} = v_i^T v_{k+1}$ .

Update  $v_{k+1} \leftarrow v_{k+1} - h_{ik}v_i$ .

Set  $h_{k+1,k} = ||v_{k+1}||_2$ .

If k > 1, apply  $J_{k-1} \cdots J_1$  to  $(h_{1,k}, \dots, h_{k,k}, h_{k+1,k}, 0, \dots)^T \in \mathbb{R}^{m+1}$ .

Determine a Givens rotation  $J_k$  such that

$$J_k \cdots J_1 \begin{pmatrix} h_{1,k} \\ \vdots \\ h_{k,k} \\ h_{k+1,k} \\ 0 \\ \vdots \end{pmatrix} \equiv \begin{pmatrix} r_{1,k} \\ \vdots \\ r_{k,k} \\ 0 \\ 0 \\ \vdots \end{pmatrix}.$$

If k=1, form  $R_1\equiv (r_{11})$ ; else form  $R_k\equiv \begin{pmatrix} r_{1,k} & r_{1,k} \\ 0\cdots & r_{k,k} \end{pmatrix}$ .

Update  $w \leftarrow J_k w$ . If  $|w_{k+1}| \le tol$  or k = m, go to Solve; else update  $v_{k+1} \leftarrow v_{k+1}/h_{k+1,k}$ .

Solve: Let k be the final iteration number from ITERATE.

Solve  $R_k y = \bar{w}$  for y, where  $\bar{w} \equiv (w_1, \dots, w_k)^T$ .

Update  $x \leftarrow x + (v_1, \dots, v_k)y$ .

If  $|w_{k+1}| \leq tol$ , accept x and stop; otherwise, return to Initialize.

# CONJUGATE GRADIENT METHOD (CG):

Given A, b, x, and tol.

Set 
$$r = b - Ax$$
,  $\rho^2 = ||r||_2^2$ ,  $z = 0$ ,  $\beta = 0$ .

Until "stop":

If  $\rho \leq tol$ , update  $x \longleftarrow x + z$  and stop.

Update  $p \longleftarrow r + \beta p$ .

Compute Ap.

Compute  $p^T A p$  and  $\alpha = \rho^2/p^T A p$ .

Update  $z \longleftarrow z + \alpha p$  and  $r \longleftarrow r - \alpha A p$ .

Update  $\beta \longleftarrow \|r\|_2^2/\rho^2$  and  $\rho^2 \longleftarrow \|r\|_2^2$ .

### PRECONDITIONED CONJUGATE GRADIENT METHOD (PCG):

Given A, b, x, tol, and a symmetric positive-definite preconditioner M.

Set 
$$r = b - Ax$$
,  $r = b - Ax$ ,  $w = M^{-1}r$ ,  $\rho^2 = r^T w$ ,  $z = 0$ ,  $\beta = 0$ .

Until "stop":

If  $\rho \leq tol$ , update  $x \longleftarrow x + z$  and stop.

Update  $p \longleftarrow w + \beta p$ .

Compute Ap.

Compute  $p^TAp$  and  $\alpha = \rho^2/p^TAp$ .

Update  $z \longleftarrow z + \alpha p$  and  $r \longleftarrow r - \alpha A p$ .

Update  $w=M^{-1}r$ ,  $\beta=r^Tw/\rho^2$ , and  $\rho^2=r^Tw$ .