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### Iterative methods for solutions of linear systems of equations

Lets say we have an equation of the form  $Ax = b$ , we can decompose the matrix  $A$  into the following form:

$$L = \begin{bmatrix} 0 & 0 & \dots & 0 \\ a_{21} & 0 & \dots & 0 \\ a_{31} & a_{32} & \dots & 0 \\ \dots & \dots & \dots & \dots \\ a_{n-1} & \dots & a_{n,n-1} & 0 \end{bmatrix} \quad D = \begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \dots & 0 \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & a_{nn} \end{bmatrix}$$

...and an upper triangular matrix; that I am not going to write but you get the idea.

*Theorem:* If  $A$  is a diagonally dominant matrix then jacobi iteration converges to the solution of  $Ax = b$ .

$$\begin{aligned} x_1 &= D^{-1}(b - (L + U)x_0) \\ x_{k+1} &= D^{-1}(b - (L + U)x_k) \\ x_{k+m} &= D^{-1}(b - (L + U)x_k) = D^{-1}(b - (L + D + U)x_k + Dx_k) \\ &\Rightarrow D^{-1}(b - Ax_k + Dx_k) = D^{-1}r_k + x_k = x_k + D^{-1}r_k \end{aligned}$$

**Note:**  $b - Ax_k$  is the residual vector. Recall the residual is defined as:  
 $Ax = b \rightarrow r = b - Ax$

We can use this to create the conditional: if  $\|r_k\|_2 \leq \text{tol} \rightarrow \text{STOP}$ .

### Jacobi Iteration

**Inputs:**  $A, x_0$

#### Loop:

$$\begin{aligned} r &= b - Ax \quad \text{for all } k = 0, 1, 2, \dots \\ x_{k+1} &= x_k + D^{-1}r_k \\ \text{error} &= \|b - Ax_k\| \\ x_k &= x_{k+1} \end{aligned}$$

Since  $D$  is a diagonal matrix,  $D^{-1}$  will just be  $\frac{1}{D}$ , which gives us our  $x_{k+1}$  modified matrix.

Lets talk more about some tricks with the residual:

$$\begin{aligned} r_{k+1} &= b - Ax_{k+1} = b - A(x_k + D^{-1}r_k) \\ &\Rightarrow (b - Ax_k) - AD^{-1}r_k \end{aligned}$$

### Guass-Seidel

$$\begin{aligned} A &= (L + D + U) \\ Ax &= b \\ (D + U)x &= b - Lx \\ (D + U)x_{k+1} &= (b - Lx) \rightarrow (D + U)x_{k+1} = (b - Lx_k) \\ x &= (D + U)^{-1}(b - Lx) \\ x^{(k+1)} &= (D + U)^{-1}(b - Lx^k) \end{aligned}$$

We can use our Back-substitution routine to find  $(D + U)^{-1}$

*November 20*

### *Shifted Power Method*

Power Method finds  $\lambda_1$ , inverse power method finds  $\lambda_n$ , and shift on the eigenvalue.

Lets say:  $Av = \lambda v$

$$Av - \mu v = \lambda v - \mu v = (\lambda - \mu)v$$

$$(A - \mu I)v = (\lambda - \mu)v$$

From this, we can find  $\lambda - \mu$  is an eigenvalue of  $(A - \mu I)$

**Example:** Let  $\lambda_1 = 10, \lambda_n = 0.1$

Find  $\lambda_n$  apply the power method to  $A^{-1}$

let  $\lambda - \mu \approx \lambda_{\frac{n}{2}}$

Now, apply the power method to  $(A - \mu I) \rightarrow \lambda - \mu$

This technique allows us to find eigenvalues between  $\lambda_n$  and  $\lambda_1$  by iteratively checking values between the min and max eigenvalue.

*eigenvalue review*

### **Properties:**

- $Ax = \lambda x$  for all  $x \neq 0$
  - lol that is it.
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