

Feb 22 - 2023 -- notes

$$d_k = x_e - x_k \quad (A x_e = b)$$

$$r_k = b - A x_k = A (x_e - x_k) = A d_k$$

$$x_{k+1} = x_k + \alpha_k r_k$$

$$r_{k+1} = r_k - \alpha_k A r_k$$

$$d_{k+1} = x_e - x_{k+1} = x_e - (x_k + \alpha_k r_k) = d_k - \alpha_k r_k$$

$$\begin{aligned} (A d_{k+1}, d_{k+1}) &= (A d_{k+1}, d_k - \alpha_k r_k) = (r_{k+1}, d_k - \alpha_k r_k) \\ &= (r_{k+1}, A^{-1} r_k) = (r_k - \alpha_k A r_k, A^{-1} r_k) \\ &= (r_k, A^{-1} r_k) - \alpha_k (r_k, r_k) \\ &= (r_k, A^{-1} r_k)[1 - \alpha_k (r_k, r_k) / (r_k, A^{-1} r_k)] \\ &= (r_k, A^{-1} r_k)[1 - (r_k, r_k)^2 / ((A r_k, r_k) (r_k, A^{-1} r_k))] \\ &\leq (r_k, A^{-1} r_k)[1 - 4 \lambda_1 \lambda_n / (\lambda_1 + \lambda_n)^2] \end{aligned}$$

$$(A d_k, d_k)$$

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convergence of steepest descent:

If $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$ then

$$\frac{\lambda_1 - \lambda_n}{\lambda_1 + \lambda_n} = \frac{\kappa - 1}{\kappa + 1} \quad \kappa = \text{condition number} = \lambda_1 / \lambda_n$$

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MR

$$\tilde{x} = x + \alpha r$$

want

$$\tilde{r} = r - \alpha A \tilde{x} \perp A \tilde{x} \quad (r - \alpha A \tilde{x}, A \tilde{x}) = 0 \rightarrow \alpha = (r, A \tilde{x}) / (A \tilde{x}, A \tilde{x})$$

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MR

assumption A is positive definite (not necessarily symmetric)

$$(Ax, x) > 0 \text{ for all } x \neq 0 \quad (\text{real})$$

$$(Ax, x) = (x, A^T x) = (A^T x, x)$$

$$(Ax, x) = \frac{1}{2} ((A + A^T) x, x)$$

$\frac{1}{2} (A + A^T) = \text{symmetric part of } A$ - it is assumed to be SPD.

CLAIM: there is a $\mu > 0$

$$(Ax, x) = \frac{1}{2} ((A + A^T) x, x) \geq \mu (x, x) \quad \text{for all } x \quad \mu > 0$$

μ = smallest eigenvalue of symm. part

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RN steepest descent

$$\tilde{x} = x + \alpha d \quad d = A^T r$$

want

$$\tilde{r} = r - \alpha Ad \perp Ad \quad (r - \alpha Ad, Ad) = 0 \rightarrow \alpha = (r, Ad)/(Ad, Ad)$$

$$\alpha = (r, Ad)/(Ad, Ad) = (A^T r, d)/(Ad, Ad) = (d, d)/(Ad, Ad)$$

normal equations:

$$A^T A x = A^T b \quad (\text{NE}) \quad \text{residual} = A^T (b - A x)$$

$d = A^T (b - A x)$ = residual for normal equations

steepest descent for NE

$$\tilde{x} = x + \alpha d$$

$$[\tilde{d} = d - \alpha (A^T A) d]$$

$$\alpha = (d, d) / (A^T A d, d) = (d, d) / (A d, Ad)$$

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Krylov methods = residual pol.

$$(1 - \alpha_k t)(1 - \alpha_{k-1} t) \dots (1 - \alpha_0 t)$$

$$p_{k+1}(0) = 0 \rightarrow p_{k+1}(t) = 1 - t q_k(t)$$

min degree for which :

$$p(A) v_1 = 0 \quad (p \text{ of degree } k)$$

GRADE

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Arnoldi - step j

$$w = A v_j$$

$$w = Av_j - h_{1j} v_1 - h_{2j} v_2 \dots - h_{jj} v_j$$
$$h_{j+1} = \dots$$

$$v_{j+1} = w/h_{j+1} \rightarrow h_{j+1} v_{j+1} = Av_j - h_{1j} v_1 - h_{2j} v_2 \dots - h_{jj} v_j$$

$$\rightarrow h_{1j} v_1 + h_{2j} v_2 \dots + h_{jj} v_j + h_{j+1} v_{j+1} = A v_j$$

$$A v_j = \sum_{i=1}^{j+1} h_{ij} v_i$$