

Lecture 24

Q: How can we get $\|x\|_2$ from CG w/o computing it if we use the Lanczos CG.

- Google says CG isn't broken and stable, why don't we care about this?
- Why doesn't "breakdown" in Lanczos (due to PR) not cause a problem for CG?
- CG vs. Steepest Descent. (3x!)
- Does Lanczos need pos. def? Where does CG use this?
- How does $\bar{T}_k$ "approximate" A?

Why is CG called CG?

Is Cholesky's diag. still $O(n^3)$?

$A V_k = V_{k+1} T_{k+1}$
 $T_{k+1} = \begin{bmatrix} \bar{T}_k & \\ & \beta_{k+1} e_k^T \end{bmatrix}$
 $\| \begin{bmatrix} \|b\| e_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} - \begin{bmatrix} \bar{T}_k \\ 0 \end{bmatrix} y \|$
 $\Rightarrow \| \begin{bmatrix} \|b\| e_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} - \begin{bmatrix} \bar{T}_k \\ 0 \end{bmatrix} y \| = 0 \iff \text{CG}$
 $\bar{T}_k y = \|b\| e_1$
 $\| \beta_{k+1} y_k \|$
 $\min_y \frac{1}{2} y^T \bar{T}_k y - \|b\| e_1^T y$
 $\Leftrightarrow \bar{T}_k y = \|b\| e_1$
 So $\|b - A V_k y\| = \| \|b\| \dot{V}_k e_1 - \dot{V}_{k+1} T_{k+1} y \|$

Lanczos doesn't need pos. def! $A V_k = V_{k+1} T_{k+1}$ is good for any symmetric A.

$A V_k = V_{k+1} T_{k+1}$
 $V_k^T A V_k = V_k^T V_{k+1} T_{k+1}$
 $= \begin{bmatrix} I & 0 \end{bmatrix} \begin{bmatrix} \bar{T}_k \\ \beta_{k+1} e_k^T \end{bmatrix}$
 $= \bar{T}_k$

SD
 $\min_x \frac{1}{2} x^T A x - x^T b$
 st $x_k = x_k + \alpha (b - A x_k)$
 $-g_k$
 $x_0 = 0$

CG
 $\min_x \frac{1}{2} x^T A x - x^T b$
 st $x_k \in K_k(A, b)$
 $= \text{span}\{b, A b, \dots, A^{k-1} b\}$
 $= V_k y$
 Note $K_k(A, b)$ contains $x_k, b - A x_k$

If you work with CG at a vector $\underline{s}$
 $A(z + s) = \underline{b}$
 or $Az = \underline{b} - A s$
 and let $z_0 = 0$

BW Stale $f = \text{Math function}$
 alg = compute inner product
 BW states if $\text{alg}(x) = f(x + \delta)$ for "small" δ

f for CG is
 $\text{find } x_k \text{ st } \|b - A x_k\| \leq \tau$