

Lecture 10

① let $V = (v_1, v_2, \dots, v_m)$ be a basis for K_m . Then

$$x^{(m)} \in x^{(0)} + K_m \Leftrightarrow x^{(m)} = x^{(0)} + Vy, \quad y \in \mathbb{R}^m$$

The residual is

$$r^{(m)} = b - Ax^{(m)} = b - A(x^{(0)} + Vy) = r^{(0)} - AVy //$$

Orthogonality $r^{(m)} \perp L_m$ means $0 = W^T r^{(m)}$

where $W = (w_1, \dots, w_m)$ is a basis for L_m .

$$\text{We get } 0 = W^T r^{(m)} = W^T (r^{(0)} - AVy) \Rightarrow$$

$$(W^T AV)y = W^T r^{(0)}$$

and

$$x^{(m)} = x^{(0)} + V(W^T AV)^{-1} W^T r^{(0)} //$$

works if $(W^T AV)$ is non-singular //

② proof Def A is pos. def if $x^T Ax > 0 \quad \forall x$.

$L_m = K_m \Rightarrow$ let $V = (v_1, \dots, v_m)$ be a basis for both. Then

$$x^T (V^T AV)x = (Vx)^T A (Vx) = y^T Ay > 0$$

for $y \neq 0$. Also $y \neq 0$ if $x \neq 0$ since V is a basis.

③ $K_1 = L_1 = \{r^{(k)}\}$ means

$$x^{(k+1)} = x^{(k)} + K_1 r^{(k)} = x^{(k)} + \alpha_k r^{(k)}$$

Orthogonality means $r^{(k+1)} \perp L_1 = \{r^{(k)}\}$ so

$$0 = (r^{(k)})^T (b - Ax^{(k+1)}) = (r^{(k)})^T (r^{(k)} - \alpha_k Ar^{(k)})$$

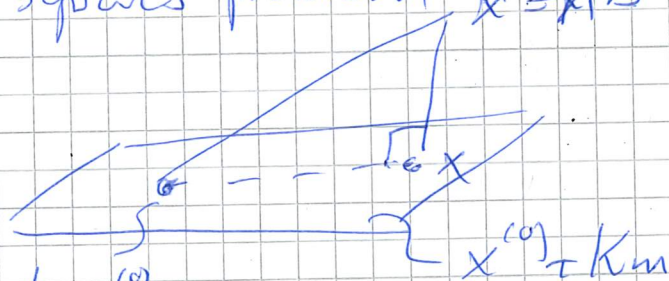
Solve for α_k gives

$$x^{(k+1)} = x^{(k)} + \frac{(r^{(k)})^T Ar^{(k)}}{\|r^{(k)}\|_2^2} r^{(k)}$$

④ Recall $(x, y)_A = x^T A y$ is a scalar product.

proof Consider the least squares problem $x^* = \bar{A}^{-1} b$

$$\min_{x \in x^{(0)} + K_m} \|x^* - x\|_A$$



so $x - x^*$ is A -orthogonal to $x^{(0)} + K_m$ or

$$0 = (x - x^{(0)})^T A (x^* - x) = \underbrace{(x - x^{(0)})^T}_{\in K_m} \underbrace{(b - Ax)}_r$$

so $r \perp K_m$. Same condition that defines the projection method.

⑤ proof Let $V = (v_1, \dots, v_m)$ be a basis for K_m then AV is a basis for $L_m = AK_m$. We get

$$r^{(m)} = r^{(0)} - AVy \perp L_m \Rightarrow 0 = (AV)^T (r^{(0)} - AVy)$$

$$\Rightarrow V^T A^T AVy = V^T A^T r^{(0)} //$$

Normal equations for

$$\min_{y \in \mathbb{R}^m} \|AVy - r^{(0)}\|_2 = \min_{y \in \mathbb{R}^m} \|A(\underbrace{x^{(0)} + Vy}_{\in K_m + x^{(0)}}) - b\|_2$$

⑥ Want to solve $Ax = b$. Can do matrix-vector multiply and linear combinations

$$\Rightarrow x = x^{(0)} + \underbrace{c_1 r^{(0)} + c_2 A r^{(0)} + \dots + c_m A^{m-1} r^{(0)}}_{\text{Krylov subspace.}}$$

⑦ proof If $\dim(K_m) = k < m$ then

$$r^{(0)} = \alpha_1 A r^{(0)} + \dots + \alpha_m A^{m-1} r^{(0)} \quad (*)$$

The residual is

$$r^{(m)} = b - Ax = b - A(x^{(0)} + c_1 r^{(0)} + \dots + c_m A^{m-1} r^{(0)})$$

$$= r^{(0)} - c_1 A r^{(0)} - \dots - c_m A^m r^{(0)}$$

Can pick $c_1, \dots, c_m$ so $r^{(m)} = 0$ due to (*) and we have the exact solution.

8

$$x = \text{pcg}(A' * A, A' * b)$$

means forming $A^T A$ and $A^T b \Rightarrow$ cgls
