

# Examen final

Teoría: G.C.

• P.d. Si los vectores no nulos  $p_1, p_2, \dots, p_L$  satisfacen que:  
 $p_i^T A p_j = 0 \quad \forall i \neq j.$

Si  $A$  es simétrica y positiva definida, entonces los vectores son linealmente independientes.

Contradicción:

Supongamos  $\exists i, k \in \{1, \dots, L\} \quad p_k \neq 0$

$$p_k = \sum_{i=0}^k a_i p_i$$

para  $a_i$  fijos. Entonces sea  $h \leq i$

$$p_h^T A p_k = p_h^T A \left( \sum_{i=0}^k a_i p_i \right)$$

$$\Rightarrow \sum_{i=0}^k a_i p_h^T A p_i \stackrel{h \neq i}{=} p_h^T A p_h > 0$$

$\nabla$   $\therefore$  los  $p_i$ 's son linealmente independientes. //

• Conclusión: Como  $\mathbb{R}^n = \text{span}\{p_0, \dots, p_{n-1}\}$  y  $p_i^T A p_j = 0$   
 $\forall i \neq j$ , sea  $x^*$  el óptimo  $\& Ax^* = b$   
 $\Rightarrow \exists \{\sigma_i\}_{i=0}^{n-1} \in \mathbb{R} \& x^* - x_0 = \sum_{i=0}^{n-1} \sigma_i p_i$



por ser conjugados de H que

$$\sigma_k = \frac{p_k^T A(x^* - x_0)}{p_k^T A p_k} = \frac{p_k^T (b - A x_0)}{p_k^T A p_k}$$

del método  $x_{k+1} = x_k + \alpha_k p_k$

$$\Rightarrow x_{k+1} = x_0 + \alpha_0 p_0 + \dots + \alpha_k p_k$$

se ha que

$$p_k^T A(x_k - x_0 + x^* - x^*) = p_k^T A(x^* - x_k) + p_k^T A(x_k - x_0)$$

$$\circ \quad p_k^T A(x^* - x_0) = p_k^T A(x^* - x_k) = -p_k^T r_k$$

$$\therefore \sigma_k = \frac{p_k^T A(x^* - x_0)}{p_k^T A p_k} = -\frac{p_k^T r_k}{p_k^T A p_k} = \alpha_k$$

converge en  $n$  pasos //

Definición:

• Definir que  $B_{k+1}$  y  $H_{k+1}$  son inversas

$$p.d. \quad B_{k+1} H_{k+1} = I$$

$$\text{se ha que } H_{k+1} = (I - A_k \Delta_k \Delta_k^T) H_k (I - p_k \Delta_k \Delta_k^T)^T$$

$$B_{k+1} = B_k - \frac{B_k \Delta_k \Delta_k^T B_k}{\Delta_k^T B_k \Delta_k} + \frac{y_k y_k^T}{y_k^T \Delta_k}$$



$$y \quad B_{k+1} z_k = y_k$$

$\Rightarrow$

$$\begin{aligned} B_{k+1} H_{k+1} &= B_{k+1} (I - p_k s_k y_k^T) H_k (I - p_k y_k s_k^T) + p_k s_k s_k^T \\ &= (B_{k+1} - p_k y_k y_k^T) H_k (I - p_k y_k s_k^T) + p_k y_k s_k^T \\ &= (B_k - \underbrace{B_k s_k s_k^T B_k}_{s_k^T B_k s_k}) H_k (I - p_k y_k s_k^T) + p_k y_k s_k^T \\ &= (I - \underbrace{B_k s_k s_k^T}_{s_k^T B_k s_k}) (I - p_k y_k s_k^T) + p_k y_k s_k^T \\ &= I - p_k y_k s_k^T - \underbrace{B_k s_k s_k^T}_{s_k^T B_k s_k} (I - p_k y_k s_k^T) + p_k y_k s_k^T \\ &= I - \underbrace{B_k s_k s_k^T}_{s_k^T B_k s_k} (I - p_k y_k s_k^T) \end{aligned}$$

$$\frac{B_k s_k s_k^T}{s_k^T B_k s_k} p_k y_k s_k^T = \frac{1}{s_k^T y_k} \frac{B_k s_k (s_k^T y_k)}{s_k^T B_k s_k} s_k^T = \frac{B_k s_k s_k^T}{s_k^T B_k s_k}$$

$$\Rightarrow \frac{B_k s_k s_k^T}{s_k^T B_k s_k} (I - p_k y_k s_k^T) = \frac{B_k s_k s_k^T}{s_k^T B_k s_k} - \frac{B_k s_k s_k^T}{s_k^T B_k s_k} = 0$$

$$\begin{matrix} 0 \\ 0 \end{matrix} \quad B_{k+1} H_{k+1} = I \quad //$$