

# Chapter 15

## 15.1 Conjugate Gradients (CG)

Minimize

$$\|r_k\|_{A^{-1}}, \quad (15.1)$$

where  $r_k = b - Ax_k$ . Also means  $\|e_k\|_A$  is minimized, where  $e_k = x_* - x_k$ .

$$x_0 = 0 \quad (15.2)$$

$$r_0 = b \quad (15.3)$$

$$p_0 = r_0 \quad (15.4)$$

For  $k = 1, 2, 3, \dots$

$$\text{step length } \alpha_k = \frac{r_{k-1}^T r_{k-1}}{p_{k-1}^T A p_{k-1}} \quad (15.5)$$

$$\text{improve solution } x_k = x_{k-1} + \alpha_k p_{k-1} \quad (15.6)$$

$$\text{residual } r_k = r_{k-1} - \alpha_k A p_{k-1} \quad (15.7)$$

$$\text{improvement } \beta_k = \frac{r_k^T r_k}{r_{k-1}^T r_{k-1}} \quad (15.8)$$

$$\text{search dir. } p_k = r_k + \beta_k p_{k-1} \quad (15.9)$$

**Theorem:** Let  $A$  be s.p.d. If  $r_{n-1} \neq 0$  (no convergence yet), then:

1.

$$K_k = \langle x_1, \dots, x_k \rangle$$

$$\begin{aligned}
&= \langle p_0, \dots, p_{k-1} \rangle \\
&= \langle r_0, \dots, r_{k-1} \rangle \\
&= \langle b, Ab, \dots, A^{k-1}b \rangle
\end{aligned} \tag{15.10}$$

2.

$$r_k^T r_j = 0 \quad k > j \tag{15.11}$$

3.

$$p_k^T A p_j = 0 \quad k > j \tag{15.12}$$

(A-conjugate search dir.)

**Proof:**

1.

$$x_0 = 0 \tag{15.13}$$

$$x_k = x_{k-1} + \alpha_k p_{k-1} \tag{15.14}$$

therefore,

$$x_k \in \langle p_0, \dots, p_{k-1} \rangle \tag{15.15}$$

and vice versa.

$$p_k = r_k + p_k p_{k-1} \tag{15.16}$$

therefore,

$$\langle p_0, \dots, p_{k-1} \rangle = \langle r_0, \dots, r_{k-1} \rangle \tag{15.17}$$

$$r_k = r_{k-1} - \alpha_k A p_{k-1} \tag{15.18}$$

therefore,

$$\langle r_0, \dots, r_{k-1} \rangle = \langle b, Ab, \dots, A^{k-1}b \rangle \tag{15.19}$$

2.

$$r_k = r_{k-1} - \alpha_k A p_{k-1} \tag{15.20}$$

$$r_k^T r_j = r_{k-1}^T r_j - \alpha_k p_{k-1}^T A r_j \tag{15.21}$$

If  $j < k - 1$  then  $r_k^T r_j = 0$  by reduction. If  $j = k - 1$  we need

$$\alpha_k = \frac{r_{k-1}^T r_{k-1}}{p_{k-1}^T A r_{k-1}} \quad (15.22)$$

We have

$$\alpha_k = \frac{r_{k-1}^T r_{k-1}}{p_{k-1}^T A p_{k-1}} \quad (15.23)$$

$$p_{k-1} = r_{k-1} + \beta_{k-1} p_{k-2} \quad (15.24)$$

$$p_{k-1}^T A p_{k-1} = p_{k-1}^T A r_{k-1} + p_{k-1} \underbrace{p_{k-1}^T A p_{k-2}}_{=0} \quad (15.25)$$

3. Use

$$p_k = r_k + \beta_k p_{k-1} \quad (15.26)$$

$$p_k^T A p_j = r_k^T A p_j + \beta_k p_{k-1}^T A p_k \quad (15.27)$$

If  $j < k - 1$  then  $p_k^T A p_j = 0$  by reduction. If  $j = k - 1$  we need

$$\begin{aligned} \beta_k &= -\frac{r_k^T A p_{k-1}}{p_{k-1}^T A p_{k-1}} \\ &= -\frac{\alpha_k r_k^T A p_{k-1}}{\alpha_k p_{k-1}^T A p_{k-1}} \end{aligned} \quad (15.28)$$

Questions:

$$r_k^T (-\alpha A p_{k-1}) \stackrel{?}{=} r_k^T r_k \quad \text{yes} \quad (15.29)$$

$$r_{k-1}^T r_{k-1} \stackrel{?}{=} p_{k-1}^T (\alpha A p_{k-1}) \quad \text{yes} \quad (15.30)$$