

**Module: Optimization 1**  
–Tutorial 05 –

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## Algorithms for Optimization without constraint

In all the following exercises,  $A$  denotes a symmetric positive definite  $N \times N$  matrix with eigenvalues  $0 < \lambda_1 \leq \dots \leq \lambda_N$ , and  $b \in \mathbb{R}^N$ . For  $v \in \mathbb{R}^N$ , let  $J(v) = \frac{1}{2}(Av, v) - (b, v)$ . Let  $u$  be the point of minimum of  $J$  in  $\mathbb{R}^N$ .

### Exercise 1 (Fixed Step-Size Gradient Algorithm)

#### 1. Convergence of Fixed Step-Size Gradient Algorithm

Show that the fixed step-size gradient algorithm

$$u^{n+1} = u^n - \rho \nabla J(u_n)$$

converges for  $0 < \rho < \frac{2}{\lambda_N}$ .

#### 2. Maximum Convergence Rate

Show that the maximum convergence rate is achieved for  $\rho = \frac{2}{\lambda_1 + \lambda_N}$  and specify its value.

### Exercise 2 (Optimal Step-Size Gradient Algorithm)

Consider the optimal step-size gradient algorithm:

$$u^{n+1} = u^n + \rho^n d^n,$$

where  $d^n = -\nabla J(u^n)$ , and  $\rho^n$  minimizes  $\inf_{\rho \in \mathbb{R}} J(u^n + \rho d^n)$ .

1. Show that for all  $n$ ,  $(d_{n+1}, d_n) = 0$ . Calculate  $\rho_n$ .
2. Let  $e^n = u^n - u$ . Show that:

$$\|e^{n+1}\|_A^2 = \left(1 - (Ad^n, d^n) \left(\frac{(d^n, d^n)}{(e^n, d^n)}\right) \frac{(e^n, Ad^n)}{(e^n, d^n)}\right) \|e^n\|_A^2,$$

where  $\|v\|_A^2 = (Av, v)$ . Deduce that:

$$\|e^{n+1}\|_A \leq \left(\lambda_N - \frac{\lambda_1}{\lambda_N + \lambda_1}\right) \|e^n\|_A.$$

### Exercise 3 (Conjugate Gradient Algorithm)

This involves the optimal step-size gradient descent algorithm:

$$u^{n+1} = u^n + \rho^n d^n \quad (2)$$

with

$$\begin{aligned} d^0 &= -\nabla J(u^0), \\ d^n &= -\nabla J(u^n) + \frac{\|\nabla J(u^n)\|^2}{\|\nabla J(u^{n+1})\|^2}, \\ \rho^n &= -(\nabla J(u^n), d^n) / (A d^n, d^n) \end{aligned}$$

Recall that the principle of this method is as follows:

starting from  $u^0 \in \mathbb{R}^n$ , a sequence  $(u^n)$  is constructed such that:

$$u^{n+1} \in u^n + G^n \quad \text{and} \quad J(u^{n+1}) = \inf_{v \in u^n + G^n} J(v) \quad (3)$$

where  $G^n = \text{vect}(\nabla J(u^0), \dots, \nabla J(u^n))$ .

1. Verify that

$$J(v) - J(u) = \frac{1}{2} \|v - u\|_A^2 \quad \forall v \in \mathbb{R}^N. \quad (4)$$

2. Show that the space  $G^n$  is generated by  $\{\nabla J(u^0), A\nabla J(u^0), \dots, A^n \nabla J(u^0)\}$ .

3. For  $k \geq 0$ , set  $e^k = u^k - u$ .

(a) For  $v \in u^n + G^n$ , show that there exists a polynomial  $Q$  of degree  $n$  such that  $v = u^0 + Q(A)\nabla J(u^0)$ .

(b) Using 3 and (??), demonstrate that

$$\|e^{k+1}\|_A = \min_{P \in P_{n+1}} \|P(A)e^0\|_A. \quad (5)$$

4. Show that

$$\|P(A)e^0\|_A^2 \leq \|e^0\|_A^2 \max_i P^2(\lambda_i) \quad \forall P \in P_k. \quad (6)$$

Conclude that

$$\|e_k\|_A \leq 2 \left( \frac{\sqrt{\kappa} - 1}{\sqrt{\kappa} + 1} \right)^n \|e^0\|_A, \quad \text{where } \kappa = \frac{\lambda_N}{\lambda_1} \quad (7)$$

5. Hint.  $\min_{p \in P_k} \|p\|_{L^1([a,b])} = \frac{1}{T_k(\frac{a+b}{a-b})}$ , where  $T_k$  denotes the Chebyshev polynomial of degree  $k$ . Moreover, we have

$$T_k\left(\frac{b}{a}\right) = T_k\left(\frac{a+b}{a-b}\right) = T_k\left(\frac{\frac{b}{a}-1}{\frac{b}{a}+1}\right) > \frac{1}{2} \left( \frac{\sqrt{\frac{b}{a}}+1}{\sqrt{\frac{b}{a}}-1} \right)^k, \text{ for } b > a.$$