

# Numerical optimization III

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The method of steepest descent

Newton-Raphson method requires the hessian matrix !

Let  $F : R^n \rightarrow R$  be differentiable.

Consider

$$\phi(t) = F(x_0 + tv)$$

where  $\|v\| = 1$ .

Q) Show that  $\phi'(0)$  is minimized when  $v = -\frac{\nabla F(x_0)}{\|\nabla F(x_0)\|}$ .

Therefore, our remaining problem is to find one-dimensional variable  $t$  for the choice of  $v$ .

Let  $t_0$  be the minimizer for  $\phi_0(t) = F(x_0 - t\nabla F(x_0))$ .

Then, set

$$x_1 = x_0 - t_0 \nabla F(x_0).$$

By repeating this process, we have

$$x_2 = x_1 - t_1 \nabla F(x_1).$$



The method of steepest descent is: Given  $x_0$ ,

$$x_{k+1} = x_k - t_k \nabla F(x_k)$$

where  $t_k$  minimizes

$$F(x_k - t \nabla F(x_k)).$$

Q) Apply the method of steepest descent to

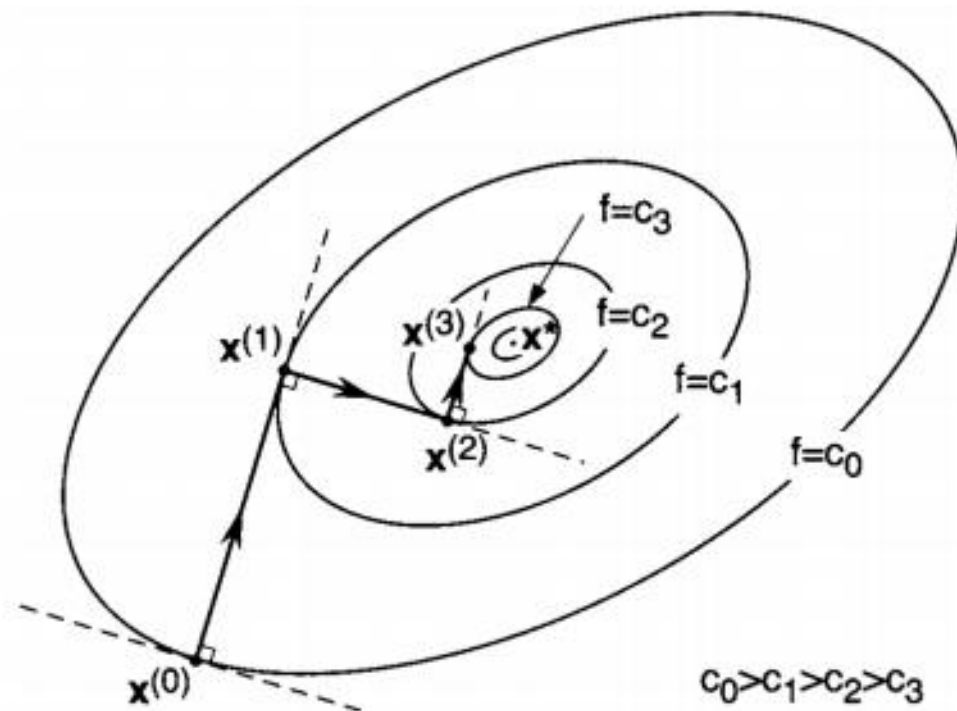
$$f(x, y) = 4x^2 - 4xy + 2y^2$$

with  $x_0 = (2, 3)$ .

Theorem. Let  $F : R^n \rightarrow R$  be continuously differentiable function. Suppose that  $x_k$  and  $x_{k+1}$  are two consecutive iterates given by the method of steepest descent. Then,

$$(\nabla F(x_k))^T \nabla F(x_{k+1}) = 0.$$

Q) Discuss the trajectory of  $(x_k)$  given by the method of steepest descent.



**Proposition 8.1** *If  $\{\mathbf{x}^{(k)}\}_{k=0}^{\infty}$  is a steepest descent sequence for a given function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , then for each  $k$  the vector  $\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}$  is orthogonal to the vector  $\mathbf{x}^{(k+2)} - \mathbf{x}^{(k+1)}$ .  $\square$*

