

Computer Science

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Systems Modelling and Analysis

Choose yourself and new technologies

L.17. Numerical optimization methods



HUMAN CAPITAL
HUMAN – BEST INVESTMENT!



Wrocław University of Technology

EUROPEAN
SOCIAL FUND

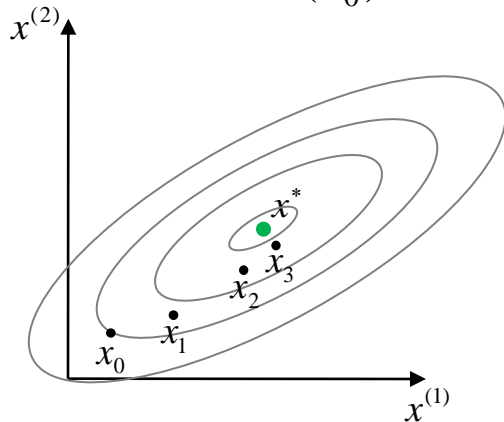
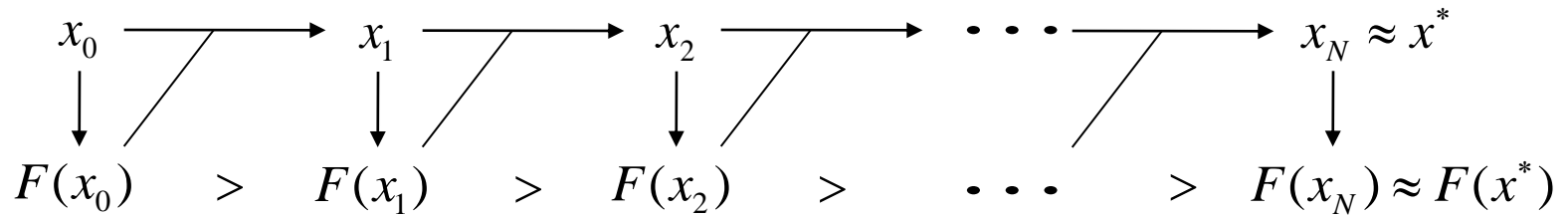


Project co-financed from the EU European Social Fund



Numerical methods

We only use information about values of objective function $F(x)$ for a given value of x .



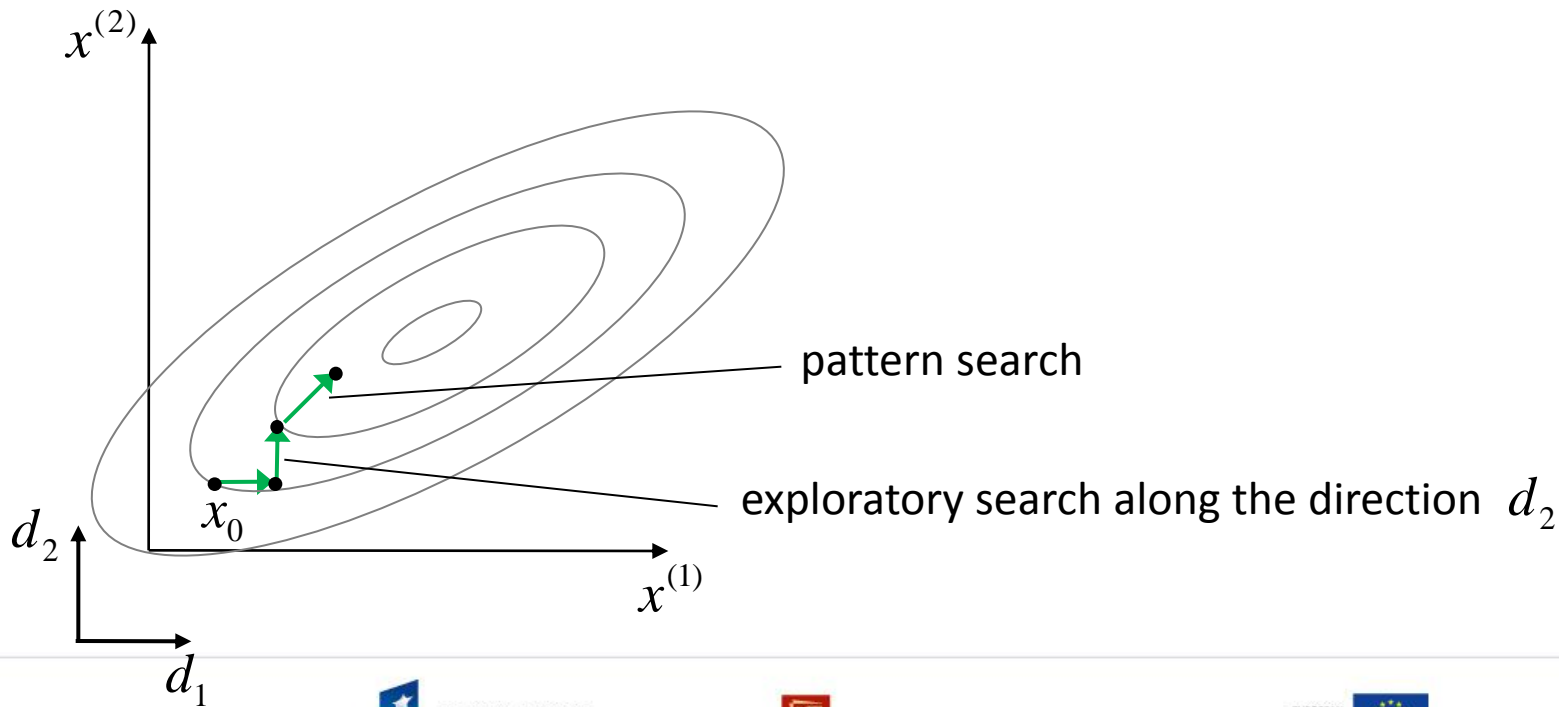
The general idea behind numerical methods.

$$x_{n+1} = G(x_n)$$



Numerical methods

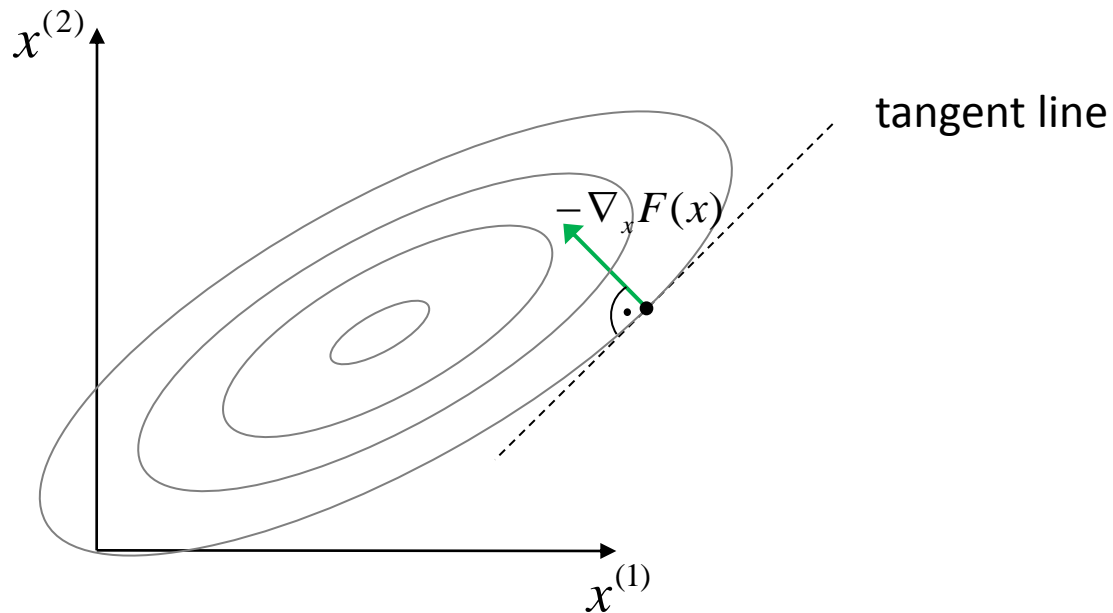
- Multidimensional search without gradient





Numerical methods

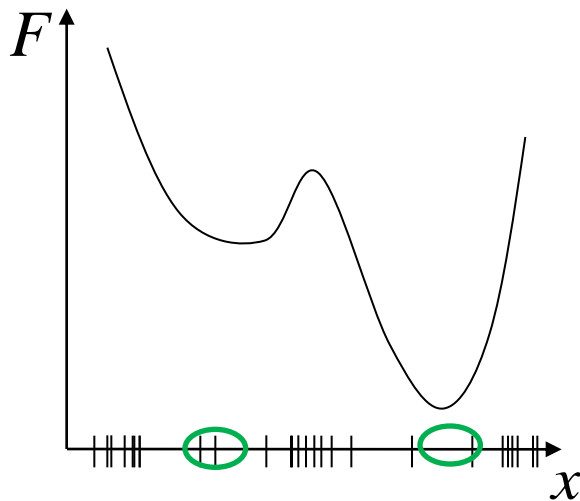
- Gradient based methods





Numerical methods

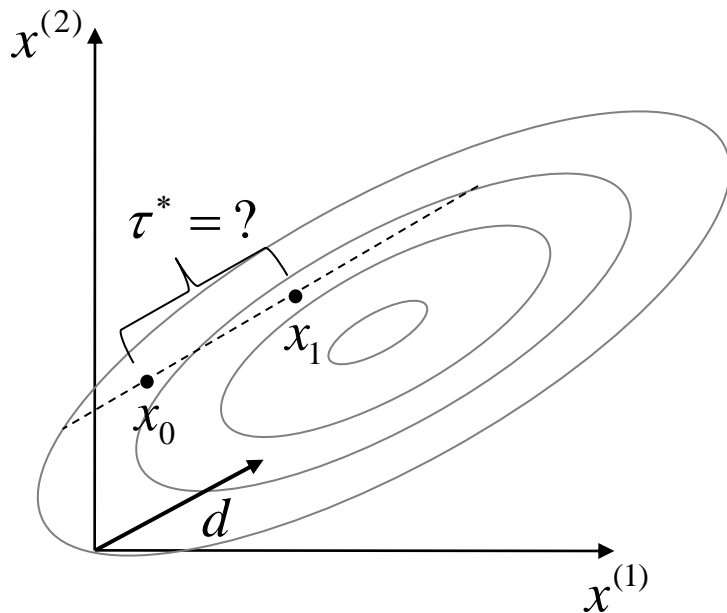
- Random search methods





Numerical methods

- Line search methods



$$x_0 \rightarrow x_1 = x_0 + \tau d$$

$$F(x_0 + \tau^* d) = \min_{\tau} F(x_0 + \tau d)$$

For given values of x_0, d :

$$F(x_0 + \tau d) \equiv f(\tau)$$

where f is one dimensional function.

Optimization task is: $\tau^* \rightarrow f(\tau^*) = \min_{\tau} f(\tau)$

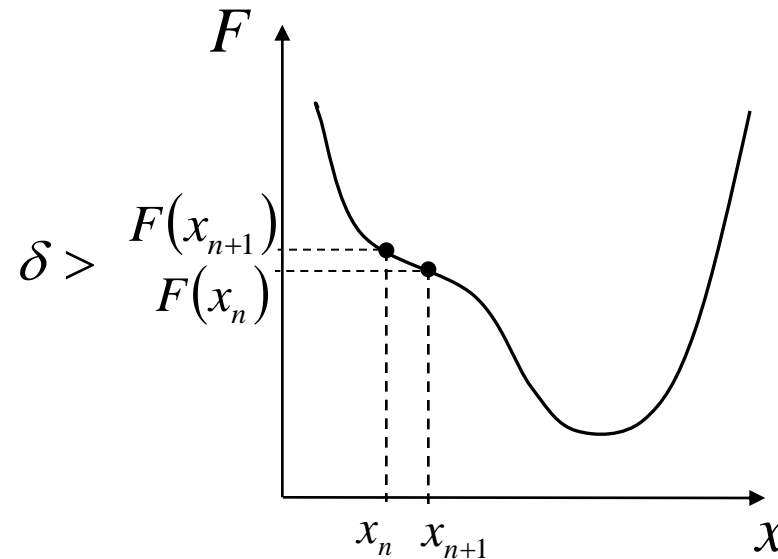


Numerical methods

- Termination criterion

$$2) \| F(x_{n+1}) - F(x_n) \| < \delta$$

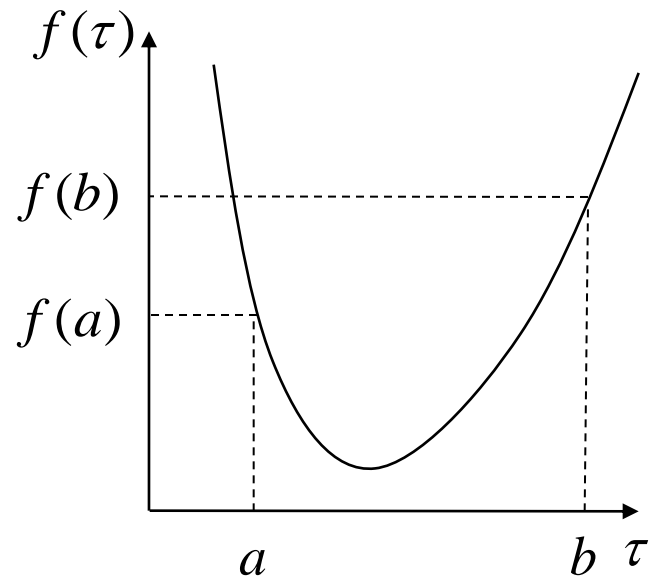
$$3) \frac{\| F(x_{n+1}) - F(x_n) \|}{\| x_{n+1} - x_n \|} < \delta$$





Numerical methods

- Line search

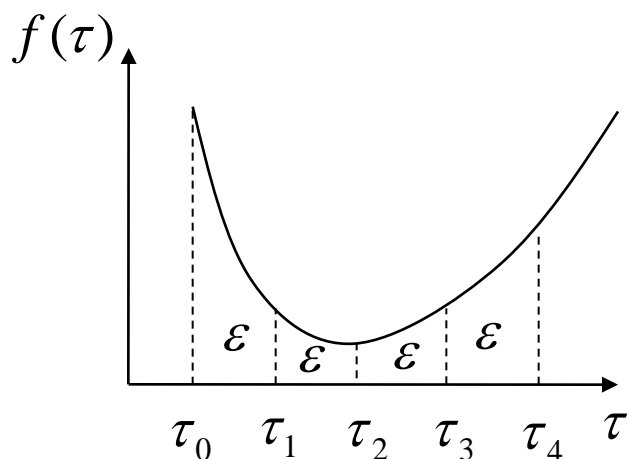


$$x \in \langle a, b \rangle \Rightarrow f(x) < \max(f(a), f(b))$$



Numerical methods

- Line search – the uniform search method



$$\tau_0 = a$$

$$\tau_n = \tau_0 + n\varepsilon$$

$$N = \left[\frac{b-a}{\varepsilon} \right] + 1$$

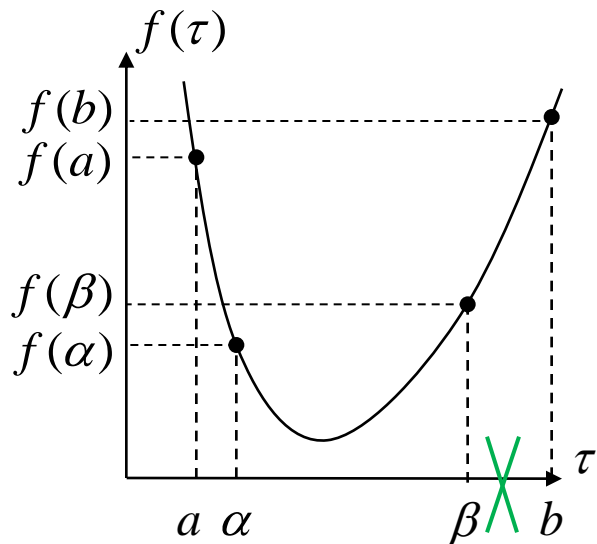
N – the number of iterations

$$\tau^* \approx \tilde{\tau} \rightarrow f(\tilde{\tau}) = \min_{1 \leq n \leq N} f(\tau_n)$$

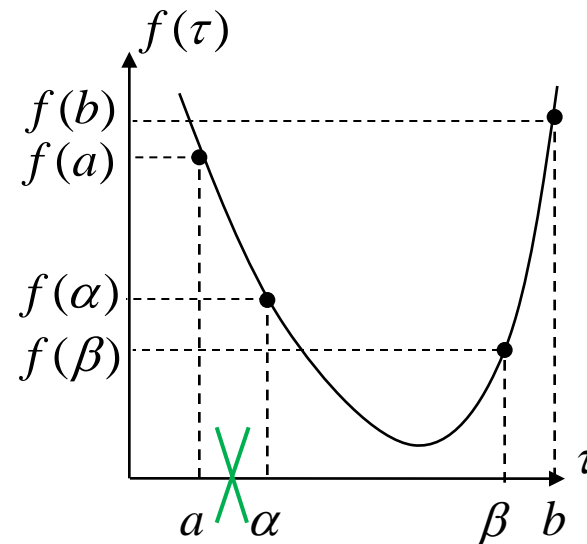


Numerical methods

- Line search – the golden section method



$$f(\alpha) < f(\beta) \Rightarrow a := a, b := \beta$$



$$f(\alpha) > f(\beta) \Rightarrow a := \alpha, b := b$$



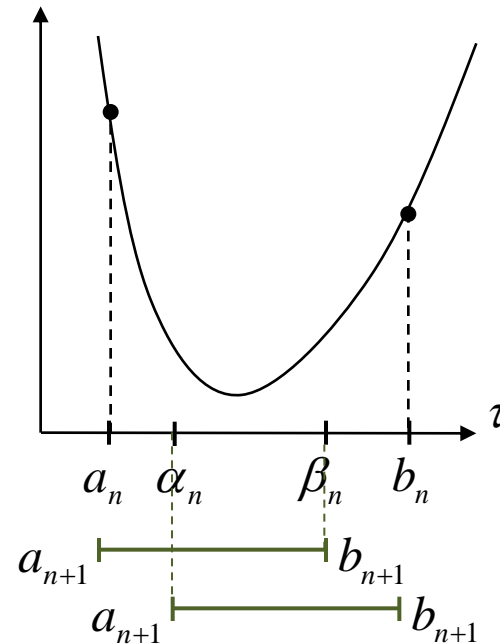
Numerical methods

- Line search – the golden section method

How to determine values of α, β ?

$$\begin{cases} \frac{\beta_n - a_n}{b_n - a_n} = \gamma \\ \frac{b_n - \alpha_n}{b_n - a_n} = \gamma \end{cases}$$

$$\begin{cases} \beta_n = a_n + \gamma(b_n - a_n) \\ \alpha_n = b_n - \gamma(b_n - a_n) \end{cases}$$





Numerical methods

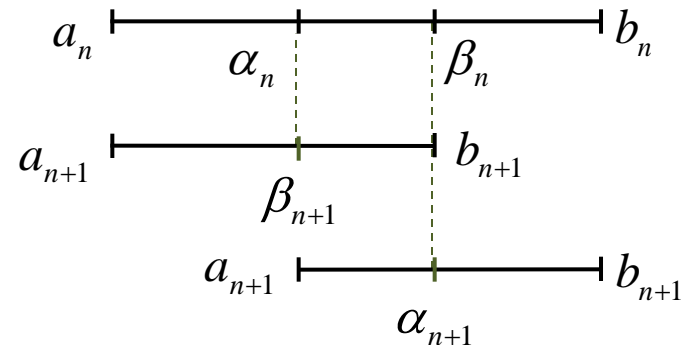
- Line search – the golden section method

How to determine values of α, β in order to reduce the number of the goal function's calculations?

$$\frac{\beta_{n+1} - a_{n+1}}{b_{n+1} - a_{n+1}} = \gamma$$

We want β_n to become α_{n+1} at the next step.

$$\frac{\alpha_n - a_n}{b_n - a_n} = \gamma$$





Numerical methods

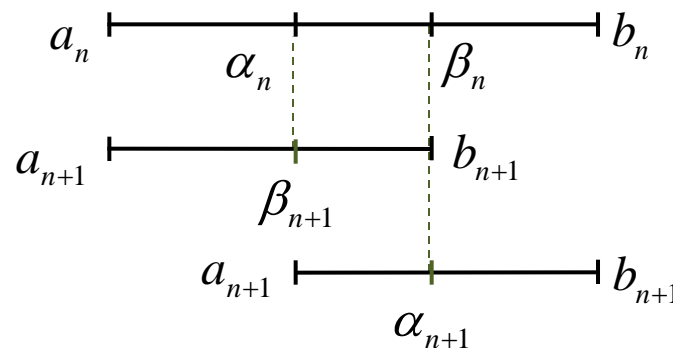
- Line search – the golden section method

$$\frac{b_n - \gamma(b_n - a_n) - a_n}{a_n + \gamma(b_n - a_n) - a_n} = \gamma$$

$$\frac{(b_n - a_n) - \gamma(b_n - a_n)}{\gamma(b_n - a_n)} = \gamma$$

$$\frac{1 - \gamma}{\gamma} = \gamma \quad \gamma^2 + \gamma - 1 = 0$$

$$\gamma = \frac{\sqrt{5} - 1}{2}$$





Numerical methods

- Line search – the golden section method

Given: $a_0, b_0, \varepsilon, \gamma = \frac{\sqrt{5}-1}{2}$

Step 0: $n = 0, \alpha_0 = b_0 - \gamma(b_0 - a_0), \beta_0 = a_0 + \gamma(b_0 - a_0)$

Step 1: IF $|b_n - a_n| < \varepsilon$ THEN STOP

Step 2: IF $f(\alpha_n) < f(\beta_n)$ THEN

$$a_{n+1} = a_n, b_{n+1} = \beta_n, \beta_{n+1} = \alpha_n, \alpha_{n+1} = b_{n+1} - \gamma(b_{n+1} - a_{n+1}), \quad n = n + 1, \text{ GOTO 1}$$

ELSE

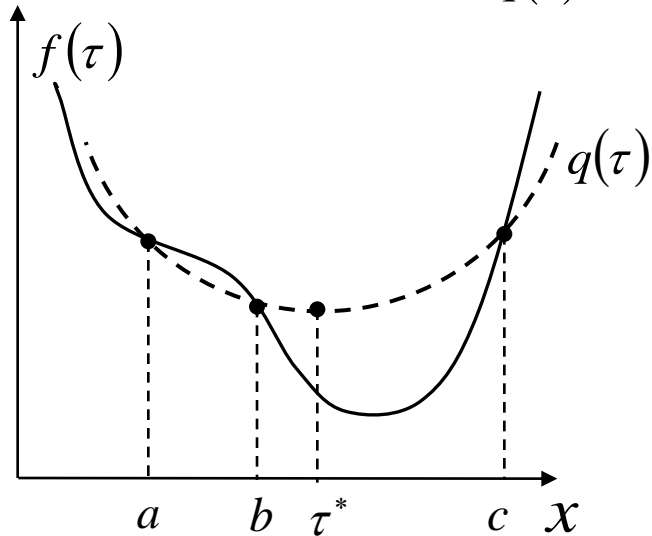
$$a_{n+1} = \alpha_n, b_{n+1} = b_n, \alpha_{n+1} = \beta_n, \beta_{n+1} = a_{n+1} + \gamma(b_{n+1} - a_{n+1}), \quad n = n + 1, \text{ GOTO 1}$$



Numerical methods

- Quadratic-fit line search method

$$q(\tau) = \frac{f(a)(\tau-b)(\tau-c)}{(a-b)(a-c)} + \frac{f(b)(\tau-a)(\tau-c)}{(b-a)(b-c)} + \frac{f(c)(\tau-a)(\tau-b)}{(c-a)(c-b)}$$



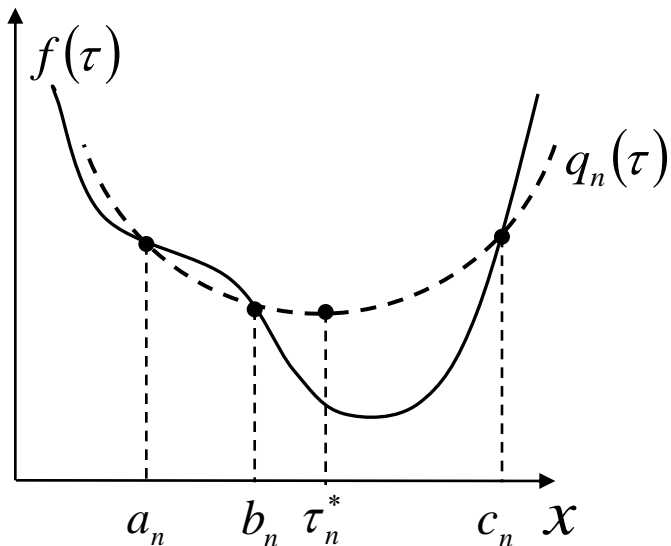
Solution of optimization task: $\tau^* \rightarrow q(\tau^*) = \min_{\tau} q(\tau)$
at the current iteration is:

$$\tau^* = \frac{1}{2} \frac{f(a)(b^2 - c^2) + f(b)(c^2 - a^2) + f(c)(a^2 - b^2)}{f(a)(b-c) + f(b)(c-a) + f(c)(a-b)}$$



Numerical methods

- Quadratic-fit line search method



IF $b_n < \tau_n^*$ THEN

IF $f(b_n) > f(\tau_n^*)$ THEN

$$a_{n+1} = b_n, b_{n+1} = \tau_n^*, c_{n+1} = c_n$$

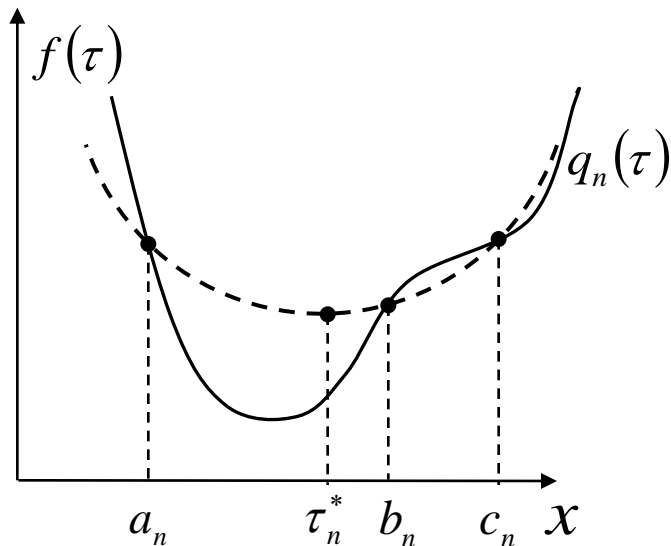
ELSE

$$a_{n+1} = a_n, b_{n+1} = b_n, c_{n+1} = \tau_n^*$$



Numerical methods

- Quadratic-fit line search method



IF $b_n > \tau_n^*$ THEN

IF $f(b_n) > f(\tau_n^*)$ THEN

$$a_{n+1} = a_n, b_{n+1} = \tau_n^*, c_{n+1} = b_n$$

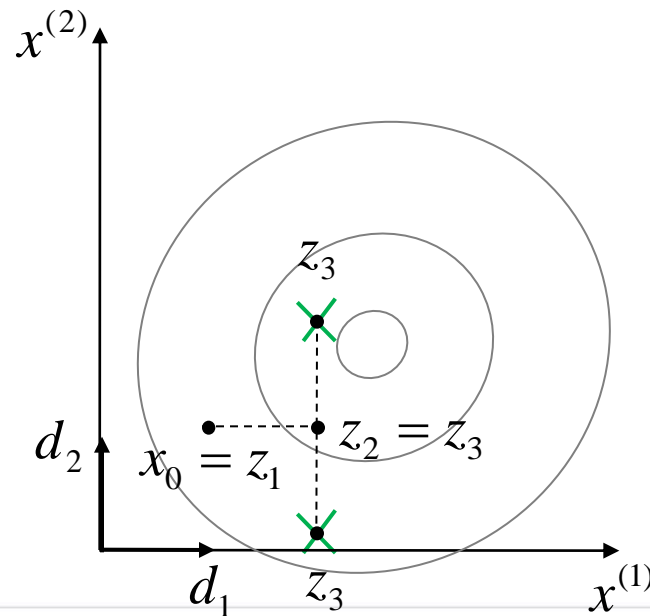
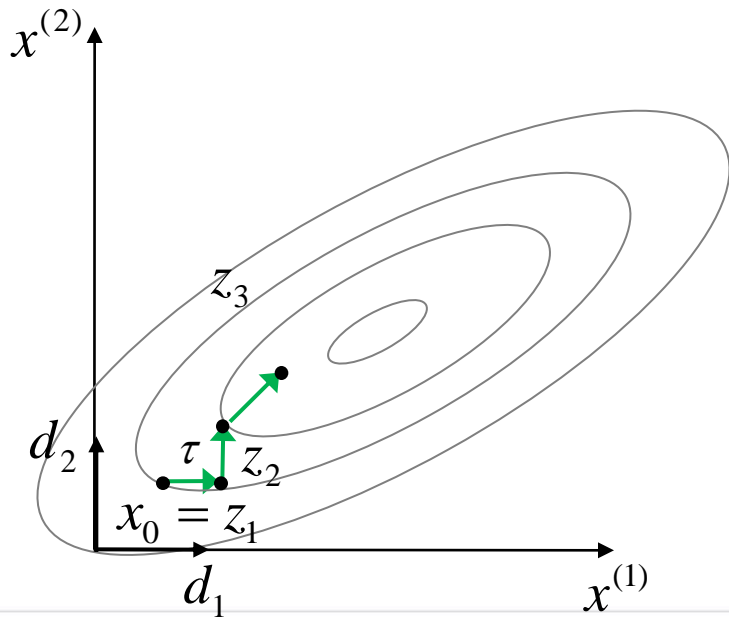
ELSE

$$a_{n+1} = \tau_n^*, b_{n+1} = b_n, c_{n+1} = c_n$$



Numerical methods

- Hooke and Jeeves method





Numerical methods

- Hooke and Jeeves method

Given: $d_1, d_2, \dots, d_s, x_0, \varepsilon, \alpha, \beta, \tau$

Step 0: $z_1 = x_0, s = 1, n = 0$

Step 1: $z_{s+1} = z_s + \tau d_s$

IF $F(z_{s+1}) < F(z_s)$ THEN GOTO 2

ELSE $z_{s+1} = z_s - \tau d_s$

IF $F(z_{s+1}) < F(z_s)$ THEN GOTO 2

ELSE $z_{s+1} = z_s, \text{ GOTO 2}$



Numerical methods

- Hooke and Jeeves method

Step 2: IF $s < S$ THEN $s := s + 1$, GOTO 2
IF $F(z_{s+1}) < F(z_1)$ THEN GOTO 3
ELSE $\tau := \beta \tau$, GOTO 1

Step 3: $x_{n+1} = z_{s+1}$
 $z_1 := z_{s+1} + \alpha(z_{s+1} - z_1)$
IF $\|x_{n+1} - x_n\| < \varepsilon$ THEN STOP
ELSE $n := n + 1, s = 1$, GOTO 1



Numerical methods

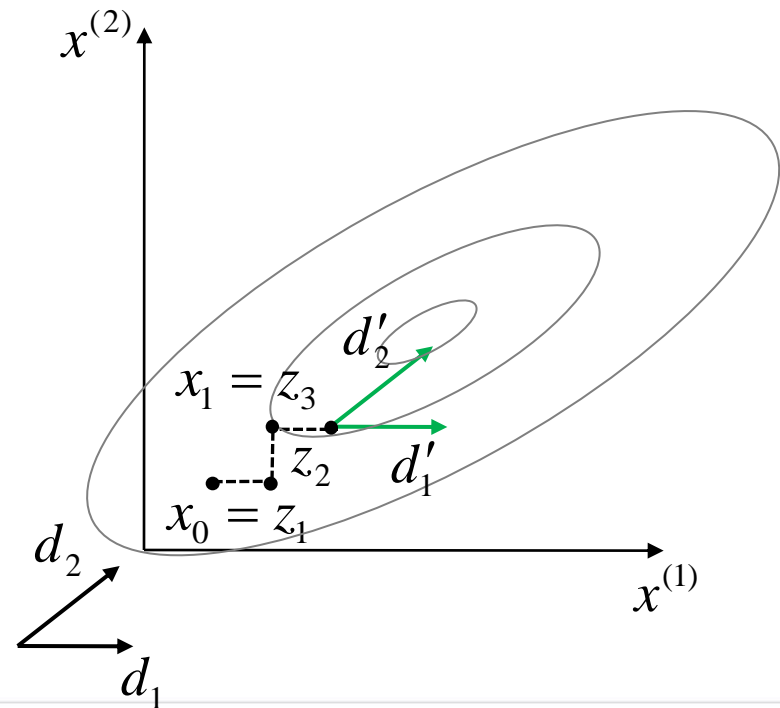
- Rosenbrock method

Conjugate directions d_1, d_2, \dots, d_s :

$$d_i^T A d_j \begin{cases} \neq 0 & \text{for } i = j \\ = 0 & \text{for } i \neq j \end{cases}$$

where A is positively defined,
symmetric matrix:

$$F(x) = x^T A x + b^T x + c$$





Numerical methods

- Rosenbrock method

Given: $d_1, d_2, \dots, d_S, x_0, \varepsilon$

Step 0: $z_1 = x_0 + \tau_s d_s$, where τ_s is minimum along the direction d_s , $s = 1, n = 0$

Step 1: $z_{s+1} = z_s + \tau_s d_s$

IF $s < S$ THEN $s := s + 1$ GOTO 1

Step 2: IF $\|x_{n+1} - x_n\| < \varepsilon$ THEN STOP



Numerical methods

- Rosenbrock method

Step 3:
$$d = \frac{z_{s+1} - z_1}{\|z_{s+1} - z_1\|},$$

$$z_s := z_{s+1} \quad (s = 1, 2, \dots, S-1)$$

$$z_s := d$$

$$z_1 := x_{n+1} + \tau d, \text{ where } \tau \text{ is minimum along the direction } d,$$

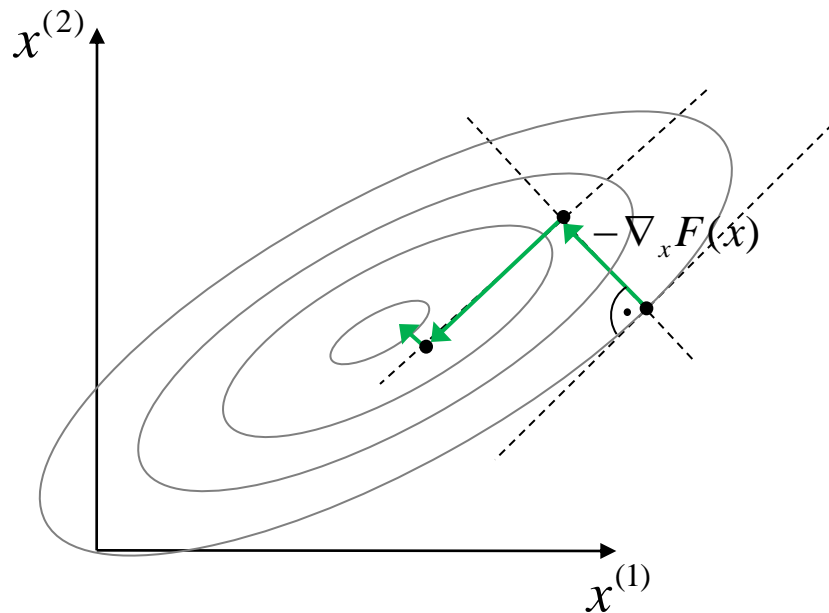
$$n := n+1, \quad s = 1$$

GOTO 1



Numerical methods

- Steepest descent method



$$x_{n+1} = x_n + \tau_n d_n \quad d_n = G(\nabla_x F(x_n))$$

$$d_n = -\nabla_x F(x_n)$$

$$x_{n+1} = x_n - \tau_n \nabla_x F(x_n), \quad x_0$$

τ_n is minimum along the direction d_n



Numerical methods

- Newton's method

Taylor's expansion:

$$F(x) = \underbrace{F(x_0) + (x-x_0)^T \nabla_x F(x_0) + \frac{1}{2} (x-x_0)^T H(x_0)(x-x_0)}_{Q(x)} + O_3(\|x-x_0\|)$$

$$\nabla_x Q(x) = \nabla_x F(x_0) + H(x_0)(x^* - x_0) = 0_S$$

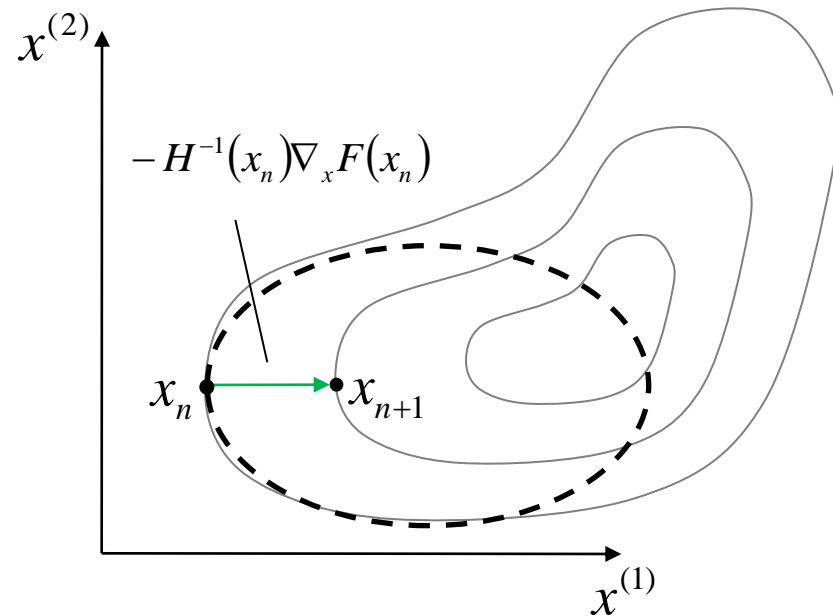
$$x^* = x_0 - H^{-1}(x_0) \nabla_x F(x_0)$$

$$x_{n+1} = x_n - H^{-1}(x_n) \nabla_x F(x_n)$$



Numerical methods

- Newton's method





Thank you for attention

