

Desanka Radunović – NUMERIČKE METODE

Linearna algebra; jednačine

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Matrice

$$A = (a_{ij}), \quad A^* = (\bar{a}_{ji}), \quad A^T = (a_{ji})$$

Hermite-ova

$$A^* = A \quad (A^T = A \quad \text{simetrična})$$

unitarna

$$A^* = A^{-1} \quad (A^T = A^{-1} \quad \text{ortogonalna})$$

Norme vektora i indukovane norme matrica $\|A\| = \sup_{x \neq 0} \frac{\|Ax\|}{\|x\|}$

uniformna

$$\|x\|_\infty = \max_{1 \leq i \leq n} |x_i|, \quad \|A\|_\infty = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}|$$

apsolutna

$$\|x\|_1 = \sum_{i=1}^n |x_i|, \quad \|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|$$

euklidska

$$\|x\|_2 = \sqrt{\sum_{i=1}^n |x_i|^2}, \quad \|A\|_2 = \sqrt{\max_{1 \leq i \leq n} \lambda_i(A^*A)}$$

sferna

$$\|A\|_s = \sqrt{\sum_{i=1}^n \sum_{j=1}^n |a_{ij}|^2}, \quad \|A\mathbf{x}\|_2 \leq \|A\|_s \|\mathbf{x}\|_2$$

uslovljenost

$$\begin{aligned} \text{cond}(I) &= 1 \leq \text{cond}(A) = \|A\| \cdot \|A^{-1}\| \leq \infty, \\ \text{cond}(A) &= \infty \quad \text{za} \quad \det(A) = 0 \end{aligned}$$

Sopstvene vrednosti i vektori

$$\begin{aligned} A\mathbf{x} = \lambda\mathbf{x} &\iff D(\lambda) \equiv \det(A - \lambda I) = 0 \\ |\lambda| \|\mathbf{x}\| = \|\lambda\mathbf{x}\| = \|A\mathbf{x}\| \leq \|A\| \|\mathbf{x}\| &\implies |\lambda| \leq \|A\| \end{aligned}$$

$$A^* = A, \quad \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n,$$

$$\implies \lambda_1 = \max_{\mathbf{x} \neq 0} \frac{(A\mathbf{x}, \mathbf{x})}{(\mathbf{x}, \mathbf{x})}, \quad \lambda_n = \min_{\mathbf{x} \neq 0} \frac{(A\mathbf{x}, \mathbf{x})}{(\mathbf{x}, \mathbf{x})}$$

Direktne metode

Gauss-ova metoda eliminacije

$$\boxed{Ax = b \iff UX = c}$$

$$U = \begin{pmatrix} * & \dots & * \\ & \ddots & \\ & & * \end{pmatrix}$$

$$(U; c) = L_{n-1}P_{n-1}L_{n-2}P_{n-2}\dots L_1P_1(A; b), \quad L = L_1^{-1}\dots L_{n-1}^{-1} = \begin{pmatrix} 1 & & \dots & 0 \\ & \dots & & \\ & & 1 & \\ * & \dots & & 1 \end{pmatrix}$$

$$l_{ij} = \frac{\bar{a}_{ij}^{(j-1)}}{a_{jj}^{(j-1)}}, \quad L_j = \begin{pmatrix} 1 & & & & 0 \\ & \ddots & & & \\ & & 1 & & \\ & & -l_{j+1,j} & 1 & \\ & & -l_{n,j} & 0 & \dots & 1 \end{pmatrix}, \quad P_j = \begin{pmatrix} 0 & & & & 1 \\ & \ddots & & & \\ & & 1 & & \\ & & & 0 & \dots & \\ 0 & & & & & 1 \end{pmatrix}$$

$$x_n = \frac{c_n}{u_{nn}}, \quad x_i = \frac{1}{u_{ii}} \left(c_i - \sum_{j=i+1}^n u_{ij} x_j \right), \quad i = n-1, \dots, 1.$$

Trougaona dekompozicija matrice

$$A = LU$$

$$u_{1k} = a_{1k}, \quad u_{ik} = a_{ik} - \sum_{j=1}^{i-1} l_{ij}u_{jk}, \quad k = i, \dots, n$$

$$i = 2, \dots, n.$$

$$l_{k1} = \frac{a_{k1}}{u_{11}}, \quad l_{ki} = \frac{1}{u_{ii}} \left(a_{ki} - \sum_{j=1}^{i-1} l_{kj}u_{ji} \right), \quad k = i+1, \dots, n,$$

Cholesky dekompozicija

$$A = LL^*, \quad A^* = A$$

$$l_{11} = \sqrt{a_{11}}, \quad l_{i1} = \frac{a_{i1}}{l_{11}},$$

$$l_{ii} = \sqrt{a_{ii} - \sum_{k=1}^{i-1} |l_{ik}|^2}, \quad l_{ij} = \frac{1}{l_{jj}} \left(a_{ij} - \sum_{k=1}^{i-1} l_{ik}\bar{l}_{jk} \right),$$
$$1 < j < i \leq n.$$

Householder-ova matrica

$$H = I - 2 w w^*, \quad \|w\|_2 = 1$$

osobine

$$H = H^* = H^{-1} \quad \text{iii} \quad H^2 = I$$

$$y = H x \quad \longrightarrow \quad \|y\|_2 = \|x\|_2, \quad (y, x) = (x, y)$$

$$\mathbf{x} = (x_1, \dots, x_m)^\top, \quad \sigma = \|\mathbf{x}\|_2, \quad x_1 = |x_1| e^{i\alpha}$$

$$k = -\sigma e^{i\alpha}$$

$$y = k \mathbf{e}_1 = (k, 0, \dots, 0)^\top = H x$$

$$\longrightarrow \quad w = \frac{\mathbf{x} - k \mathbf{e}_1}{\|\mathbf{x} - k \mathbf{e}_1\|_2}$$

QR dekompozicija

$$A = QR$$

$$R = \begin{pmatrix} * & \cdots & * \\ & \ddots & \\ 0 & & * \end{pmatrix} = T_{n-1} \cdots T_1 A,$$

$$T_m = \begin{pmatrix} 1 & \cdots & 0 & | & 0 & \cdots & 0 \\ & \cdots & & | & & \cdots & \\ & & 1 & | & 0 & \cdots & \\ & & & & & & \\ 0 & \cdots & & & 0 & & \\ & & \ddots & & & & \\ & & & & & & \\ & & & & & & \\ 0 & \cdots & & & 0 & & \end{pmatrix} \quad \left| \quad \begin{matrix} \hline \hline \hline \hline \hline \hline \hline \hline \hline \\ H_m \end{matrix} \right.$$

$$Q = T_1 \cdots T_{n-1}, \quad Q^* = Q^{-1} = T_{n-1} \cdots T_1$$

$$\dim(H_m) = n - m + 1, \quad \mathbf{x}_m = \begin{pmatrix} a_{m,m}^{(m-1)} & \cdots & a_{n,m}^{(m-1)} \end{pmatrix}^\top \quad (A_m = \{a_{i,j}^{(m)}\} = T_m A_{m-1}, A_0 \equiv A)$$

Singularna dekompozicija (SVD)

$$A = U W V^*$$

$$U^* = U^{-1}, \quad V^* = V^{-1}, \quad W = \begin{pmatrix} D & 0 \\ 0 & 0 \end{pmatrix}, \quad D = \begin{pmatrix} \sigma_1 & & 0 \\ & \ddots & \\ 0 & & \sigma_r \end{pmatrix}$$

singularne vrednosti

$$\sigma_k^2 = \lambda_k(A^*A), \quad \sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$$

$$\dim(A) = m \times n, \quad r \leq \min(m, n), \quad \dim(U) = m \times m, \quad \dim(V) = n \times n,$$

$$(A^*A)V = V(W^*W) \longrightarrow V = (\mathbf{v}_1, \dots, \mathbf{v}_n), \quad (A^*A)\mathbf{v}_k = \sigma_k^2 \mathbf{v}_k$$

$$k = 1, \dots, r,$$

$$(AA^*)U = U(WW^*) \longrightarrow U = (\mathbf{u}_1, \dots, \mathbf{u}_m), \quad (AA^*)\mathbf{u}_k = \sigma_k^2 \mathbf{u}_k$$

$$r = n = \dim(A) \longrightarrow \sigma_1 = \|A\|_2, \quad \sigma_n = \frac{1}{\|A^{-1}\|_2}, \quad \text{cond}(A) = \|A\|_2 \|A^{-1}\|_2 = \frac{\sigma_1}{\sigma_n}$$

sistem

$$\text{LU} \quad Ly = Pb, \quad Ux = y \quad (PAx = LUx = Pb)$$

$$\text{Cholesky} \quad Ly = b, \quad L^*x = y$$

$$\text{QR} \quad y = Q^*b, \quad Rx = y$$

$$\text{SVD} \quad x = VW^{-1}U^*b$$

determinanta

$$\text{LU} \quad \det(PA) = \pm \det(A) = \det(L) \cdot \det(U) = u_{11} \cdots u_{nn}$$

$$\text{Cholesky} \quad \det(A) = (l_{11} \cdots l_{nn})^2$$

$$\text{inverzna matrica} \quad AA^{-1} = I \quad \longrightarrow \quad Ax_i = e_i, \quad i = 1, \dots, n.$$

$$A^{-1} = (x_1, \dots, x_n), \quad I = (e_1, \dots, e_n)$$

Iterativne metode

$$F(x) = 0 \iff x = G(x)$$

$$x^{(k+1)} = G(x^{(k)}), \quad k = 0, 1, \dots, \quad \lim_{k \rightarrow \infty} x^{(k)} = \bar{x}, \quad \bar{x} = G(\bar{x})$$

kontrakcija

$$\|G(x) - G(y)\| \leq q \|x - y\|, \quad 0 \leq q < 1$$

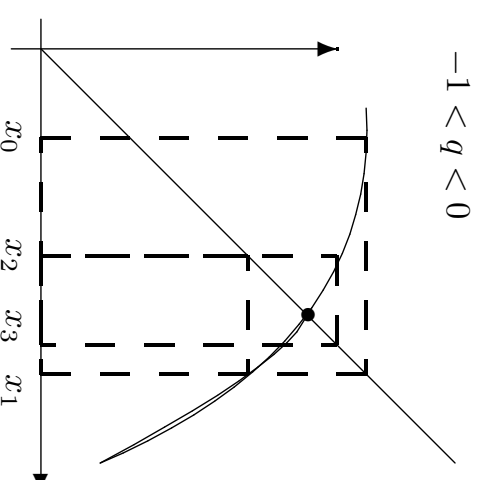
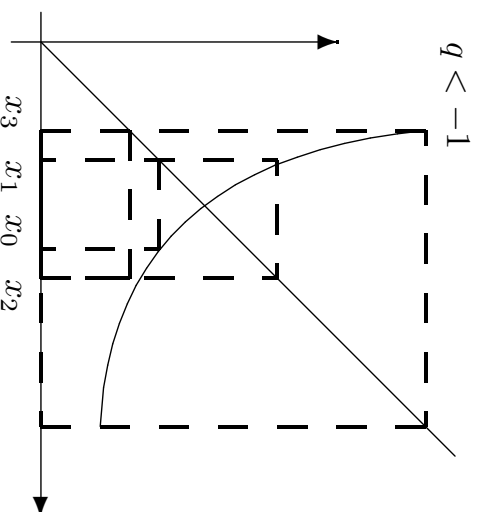
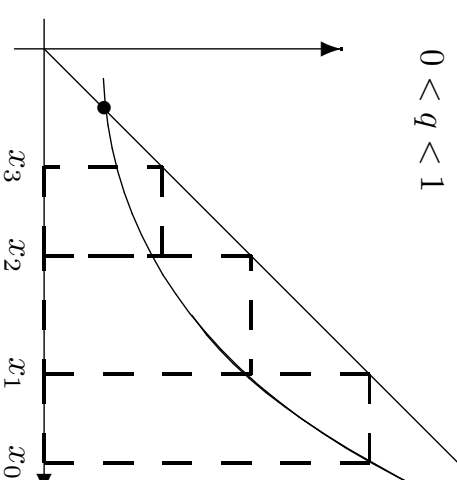
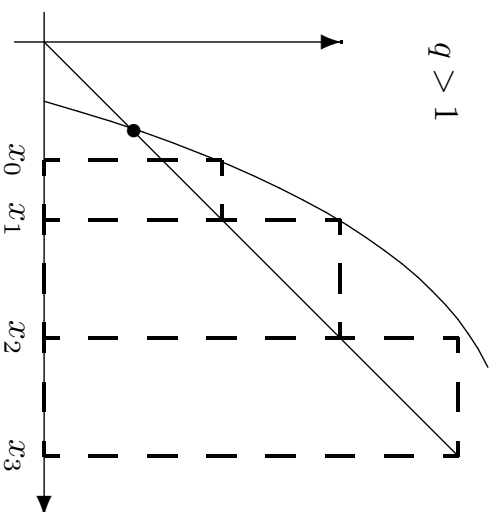
$$\mathbf{x} \in \mathcal{R}^n \quad q = \max_z \|J(G(\mathbf{z}))\|, \quad J(G(\mathbf{z})) = \left\{ \frac{\partial g_i}{\partial x_j}(\mathbf{z}) \right\}_{i,j=1}^n$$

greška

$$\text{apriorna} \quad \|\bar{x} - x^{(k)}\| \leq \frac{q^k}{1-q} \|G(x^{(0)}) - x^{(0)}\| \longrightarrow k \geq \frac{\ln \frac{\varepsilon(1-q)}{\|G(x^{(0)}) - x^{(0)}\|}}{\ln q}$$

$$\text{aposteriorna} \quad \|\bar{x} - x^{(k)}\| \leq \frac{q}{1-q} \|x^{(k)} - x^{(k-1)}\| \longrightarrow \|x^{(k)} - x^{(k-1)}\| \leq \frac{1-q}{q} \varepsilon$$

Geometrijska interpretacija metode iteracije $x^{(k+1)} = g(x^{(k)})$, ($n = 1$)



„Preconditioning“ (preduslovljavanje) matricom P

$$A \mathbf{x} = \mathbf{b} \quad \iff \quad P \mathbf{x} = N \mathbf{x} + \mathbf{b} \quad \text{za} \quad A = P - N$$

$\mathbf{x}^{(0)}$ proizvoljno

$$P \mathbf{x}^{(k+1)} = N \mathbf{x}^{(k)} + \mathbf{b}, \quad k = 0, 1, \dots,$$

$$\mathbf{x}^{(k+1)} = P^{-1} N \mathbf{x}^{(k)} + P^{-1} \mathbf{b}$$

konvergenција

$$J(G(\mathbf{x})) \equiv P^{-1} N \quad \implies \quad \|P^{-1} N\| < 1$$

za $N = P - A$

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - P^{-1}(A \mathbf{x}^{(k)} - \mathbf{b})$$

vektor greške $\mathbf{r}^{(k)} = A \mathbf{x}^{(k)} - \mathbf{b}$,

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - P^{-1} \mathbf{r}^{(k)}$$

Metode, $A = L + D + U$ (donji trougao + dijagona + gornji trougao),

Jacobi $P = D, \quad N = -(L + U)$

$$\mathbf{x}^{(k+1)} = -D^{-1}(L + U)\mathbf{x}^{(k)} + P^{-1}\mathbf{b} = \mathbf{x}^{(k)} - D^{-1}\mathbf{r}^{(k)}$$

Gauss-Seidel $P = L + D, \quad N = -U$

$$\mathbf{x}^{(k+1)} = -(L + D)^{-1}U\mathbf{x}^{(k)} + (L + D)^{-1}\mathbf{b} = \mathbf{x}^{(k)} - (L + D)^{-1}\mathbf{r}^{(k)}$$

Relaksacija $\omega A\mathbf{x} = \omega\mathbf{b}, \quad 0 < \omega < 2 \quad (\omega = 1 \quad \text{G-S})$

$$P = \omega L + D, \quad N = -(\omega U + (\omega - 1)D)$$

$$\begin{aligned}\mathbf{x}^{(k+1)} &= -(\omega L + D)^{-1}(\omega U + (\omega - 1)D)\mathbf{x}^{(k)} + \omega(\omega L + D)^{-1}\mathbf{b} \\ &= \mathbf{x}^{(k)} - \omega(\omega L + D)^{-1}\mathbf{r}^{(k)}\end{aligned}$$

Richardson-ova metoda (uopštenje uvođenjem parametra τ_k)

stacionarna $\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - \tau P^{-1} \mathbf{r}^{(k)}$ ($\tau_k \equiv \tau$)

$$P^{-1} = I \longrightarrow \tau_{\text{opt}} = \frac{2}{M+m}, \quad M = \max_i |\lambda_i(A)|, \quad m = \min_i |\lambda_i(A)|$$
$$\|\bar{\mathbf{x}} - \mathbf{x}^{(k)}\|_2 \leq \left(\frac{M-m}{M+m} \right)^k \|\bar{\mathbf{x}} - \mathbf{x}^{(0)}\|_2$$

nestacionarna $\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - \tau_k P^{-1} \mathbf{r}^{(k)}$

$$P^{-1} = I \longrightarrow \tau_{j,\text{opt}} = \frac{2}{M+m+(M-m)\cos\frac{2j-1}{2k}\pi}, \quad j = 1, \dots, k,$$
$$\|\bar{\mathbf{x}} - \mathbf{x}^{(k)}\|_2 \leq 2 \left(\frac{\sqrt{M} - \sqrt{m}}{\sqrt{M} + \sqrt{m}} \right)^k \|\bar{\mathbf{x}} - \mathbf{x}^{(0)}\|_2,$$

Gradijentne metode

$$A = A^*, \quad A \bar{\mathbf{x}} = \mathbf{b} \quad \iff \quad F(\bar{\mathbf{x}}) = \min F(\mathbf{x})$$

funkcional

$$F(\mathbf{x}) = \frac{1}{2} (A \mathbf{x}, \mathbf{x}) - (\mathbf{b}, \mathbf{x})$$

gradijent

$$\nabla F(\mathbf{x}^{(k)}) = \left(\frac{\partial F}{\partial x_1} \quad \dots \quad \frac{\partial F}{\partial x_n} \right)^T = A \mathbf{x}^{(k)} - \mathbf{b} = \mathbf{r}^{(k)}$$

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \lambda_k \mathbf{d}^{(k)}, \quad k = 0, 1, \dots, \quad \lim_{k \rightarrow \infty} \mathbf{x}^{(k)} = \bar{\mathbf{x}}$$

popravka

$$\lambda_k = - \frac{(\nabla F(\mathbf{x}^{(k)}), \mathbf{d}^{(k)})}{(A \mathbf{d}^{(k)}, \mathbf{d}^{(k)})} = - \frac{\sum_{i=1}^n \frac{\partial F(\mathbf{x}^{(k)})}{\partial x_i} d_i^{(k)}}{\sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 F(\mathbf{x}^{(k)})}{\partial x_i \partial x_j} d_i^{(k)} d_j^{(k)}}$$

pokoordinatni spust $\mathbf{d}^{(k)} = \mathbf{e}^{(k)}$, $\mathbf{e}^{(k)} = (0 \ \dots \ 0 \ 1 \ 0 \ \dots \ 0)^\top$

najstrmiji spust $\mathbf{d}^{(k)} = -\nabla F(\mathbf{x}^{(k)}) = -\mathbf{r}^{(k)}$, $\lambda_k = -\delta_k$

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \delta_k \mathbf{r}^{(k)}, \quad \delta_k = \frac{(\mathbf{r}^{(k)}, \mathbf{r}^{(k)})}{(A \mathbf{r}^{(k)}, \mathbf{r}^{(k)})}, \quad k = 0, 1, \dots,$$

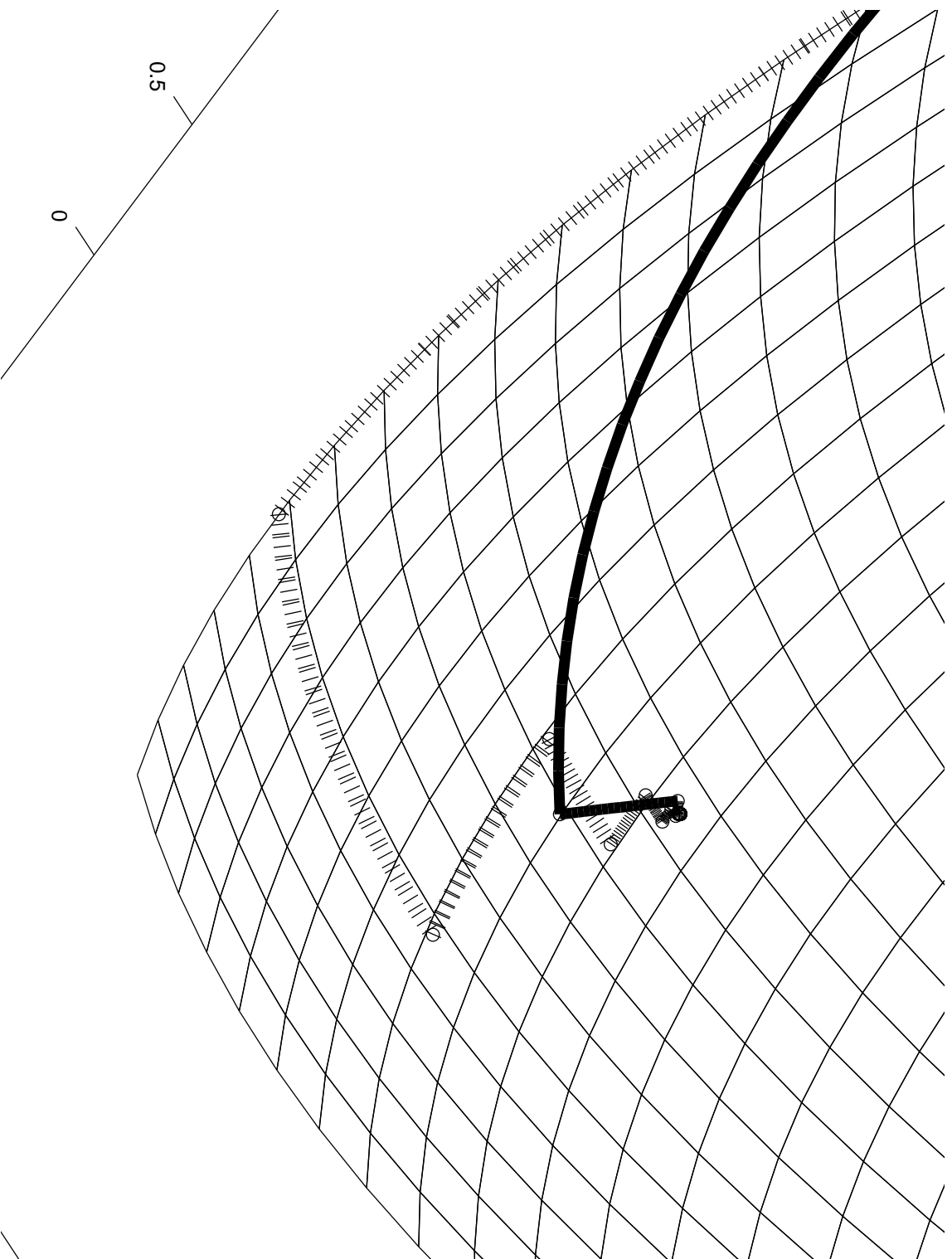
konjugovani gradient $\mathbf{d}^{(k)} = \mathbf{p}^{(k)}$, $(A \mathbf{p}^{(i)}, \mathbf{p}^{(j)}) = \delta(i - j)$

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - \lambda_k \mathbf{p}^{(k)}, \quad k = 0, 1, \dots,$$

$$\mathbf{r}^{(k)} = A \mathbf{x}^{(k)} - \mathbf{b}, \quad \mathbf{p}^{(k+1)} = \mathbf{r}^{(k+1)} - \mu_k \mathbf{p}^{(k)}, \quad \mathbf{p}^{(0)} = \mathbf{r}^{(0)}$$

$$\lambda_k = \frac{(\mathbf{r}^{(k)}, \mathbf{p}^{(k)})}{(A \mathbf{p}^{(k)}, \mathbf{p}^{(k)})}, \quad \mu_k = \frac{(A \mathbf{p}^{(k)}, \mathbf{r}^{(k+1)})}{(A \mathbf{p}^{(k)}, \mathbf{p}^{(k)})}$$

Pokoordinatni i najbrži spust



Metoda najmanjih kvadrata

$$\min_{\mathbf{x}} \|\mathbf{r}(\mathbf{x})\|_2$$

$$\|\mathbf{r}(\mathbf{x})\|_2^2 = \|\mathbf{A}\mathbf{x} - \mathbf{b}\|_2^2 = (\mathbf{A}^\top \mathbf{A}\mathbf{x}, \mathbf{x}) - 2(\mathbf{A}^\top \mathbf{b}, \mathbf{x}) + (\mathbf{b}, \mathbf{b})$$

Pridruženi problem sa samokonjugovanim operatorom

$$\mathbf{A}^\top / \mathbf{A}\mathbf{x} = \mathbf{x} \quad \longrightarrow \quad \mathbf{A}^\top \mathbf{A}\mathbf{x} = \mathbf{A}^\top \mathbf{b}$$

$$F(\mathbf{x}) = \frac{1}{2} (\mathbf{A}^\top \mathbf{A}\mathbf{x}, \mathbf{x}) - (\mathbf{A}^\top \mathbf{b}, \mathbf{x})$$

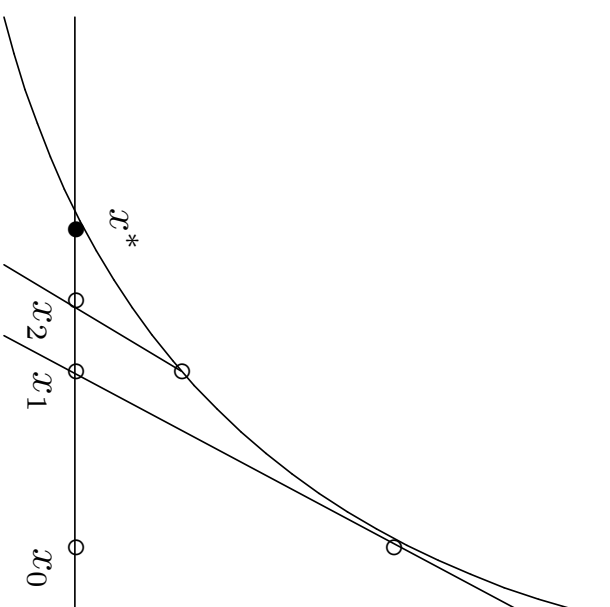
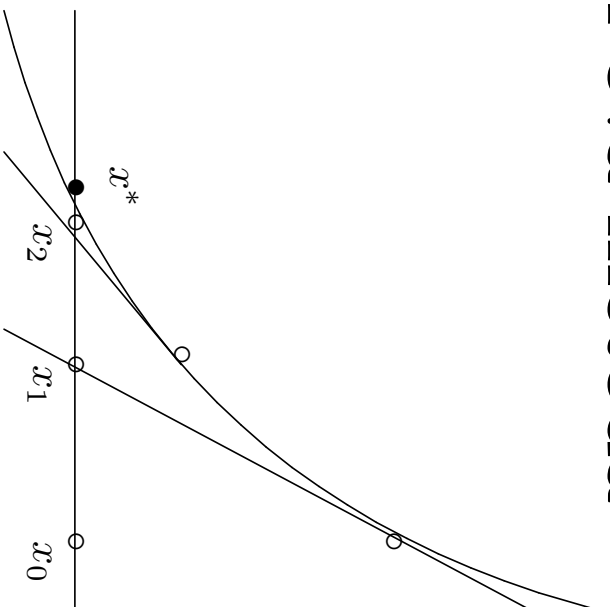
$$\begin{aligned} \mathbf{A}\bar{\mathbf{x}} = \mathbf{b} & \iff \|\mathbf{r}(\bar{\mathbf{x}})\|_2 = \min \|\mathbf{r}(\mathbf{x})\|_2, \\ & F(\bar{\mathbf{x}}) = \min F(\mathbf{x}) \end{aligned}$$

Omogućava primenu gradijentnih metoda i kada matrica sistema nije Hermite-ova

Nelinearne jednačine

$$f(x) = 0, \quad x \in \mathcal{R}^1$$

Newton-ova metoda

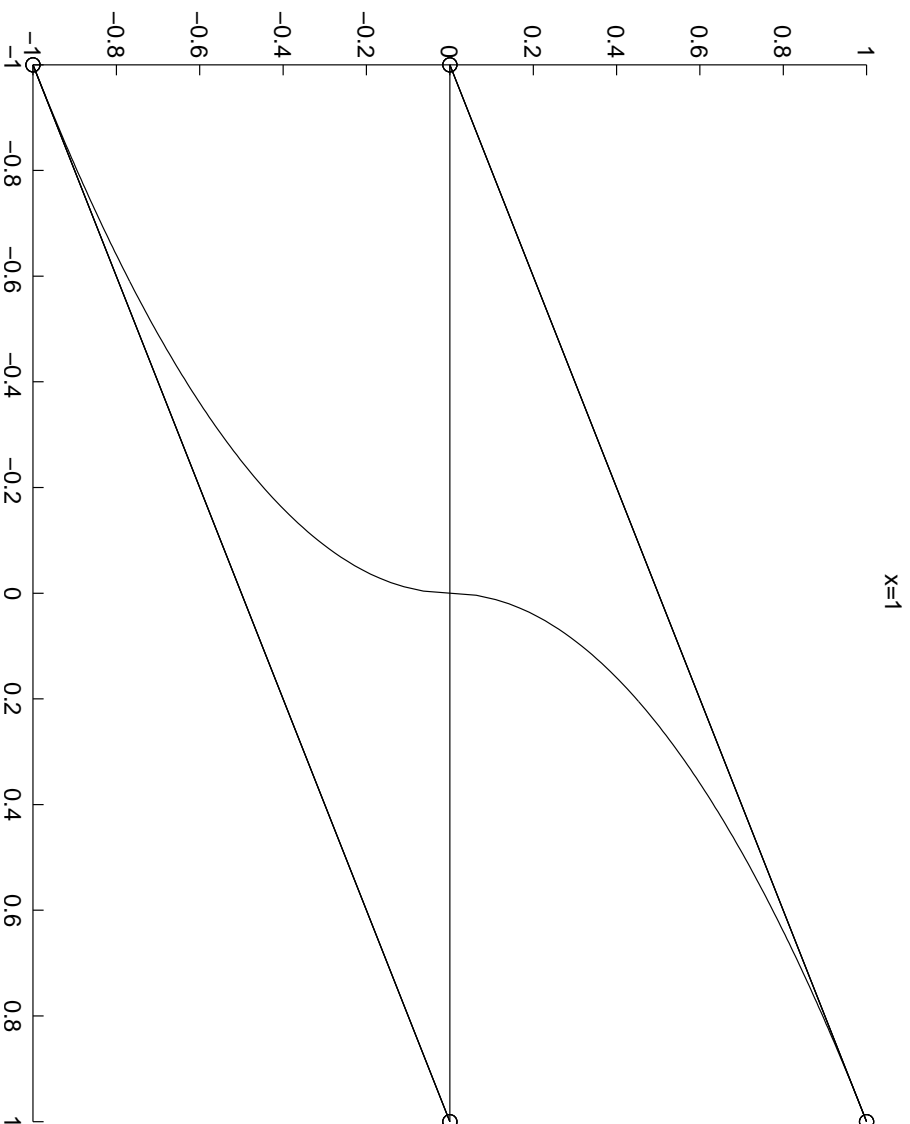


$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})},$$

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(0)})}$$

greška

$$|\bar{x} - x^{(k)}| \leq \frac{M_2}{2m_1} |x^{(k)} - x^{(k-1)}|^2$$



$$m_1 = \min |f'(x)|$$

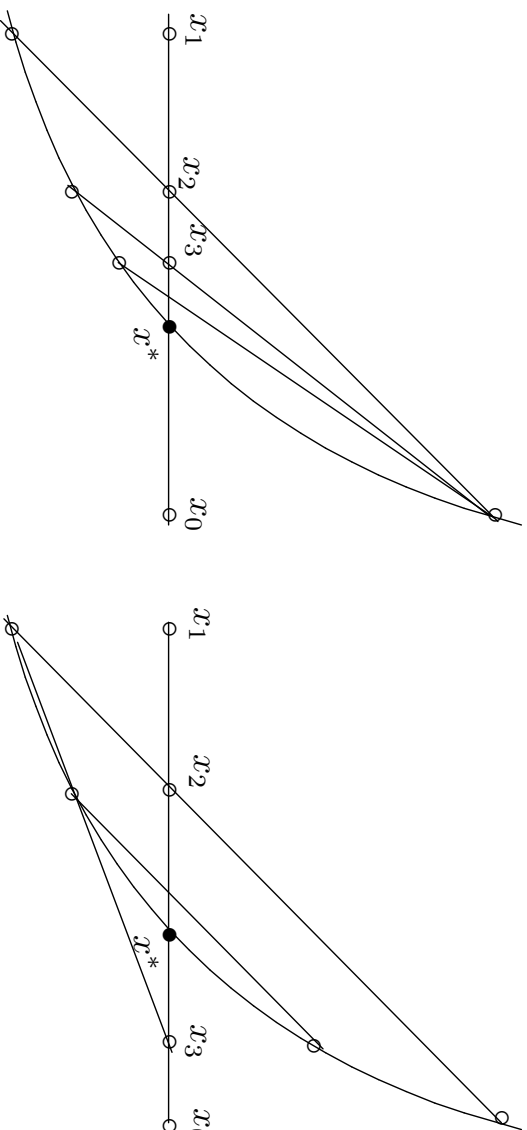
$$M_2 = \max |f''(x)|$$

Uslovi konvergenije:

- $f \in C^1[a, b]$
- $f(a)f(b) < 0$
- $\text{sign} f'(x) = \text{const}, f'(x) \neq 0$
- $\text{sign} f''(x) = \text{const}$
- $f(x_0)f''(x_0) > 0$

Regula-falsi

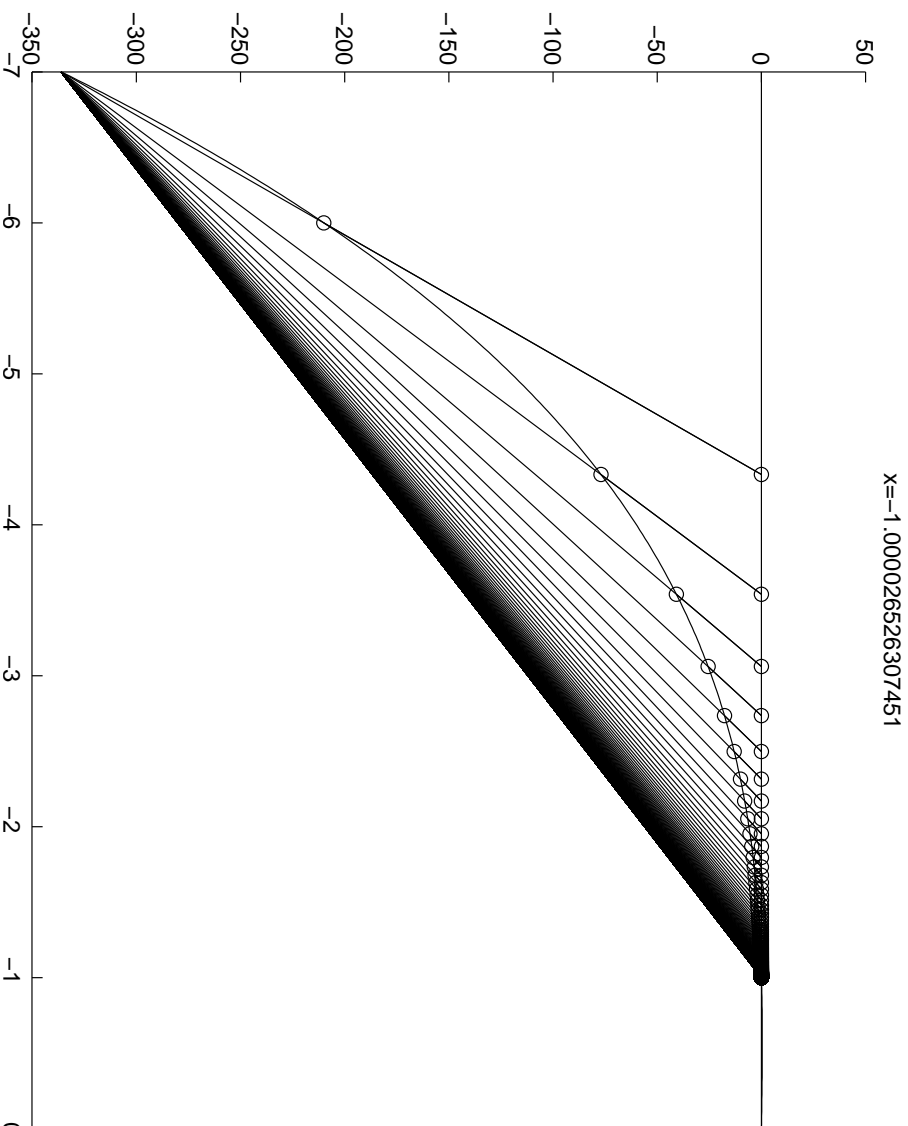
(levo)
$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f(x_F) - f(x^{(k)})} (x_F - x^{(k)})$$



Sečica

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f(x^{(k-1)}) - f(x^{(k)})} (x^{(k-1)} - x^{(k)})$$
 (desno)

Iteracije metode regula-falsi



$$f'(x) \approx \frac{f(x_F) - f(x^{(k)})}{x_F - x^{(k)}}$$

Greška:

$$|\bar{x} - x^{(k)}| \leq \frac{M_1 - m_1}{m_1} |x^{(k)} - x^{(k-1)}|$$

$$m_1 = \min |f'(x)|$$

$$M_1 = \max |f'(x)|$$

Metoda polovljenja

$$\bar{x} \approx x^{(k)} = \frac{1}{2}(a_k + b_k),$$

greška

$$|\bar{x} - x^{(k)}| \leq \frac{1}{2^{k+1}}(b - a) \leq \varepsilon \quad \text{iii} \quad k \geq \left\lceil \frac{\ln \frac{b-a}{\varepsilon}}{\ln 2} \right\rceil$$

$$[a, b] \equiv [a_0, b_0] \supset \dots \supset [a_k, b_k] \supset \dots, \quad f(a_k)f(b_k) < 0, \quad b_k - a_k = \frac{1}{2^k}(b - a)$$

Metoda Bairstow-a

$$P_n(x) = x^n + a_1x^{n-1} + \dots + a_n = 0$$

$$P_n(x) = (x^2 + px + q)(x^{n-2} + b_1x^{n-3} + \dots + b_{n-3}x + b_{n-2}) + rx + s$$

$$b_m = a_m - pb_{m-1} - qb_{m-2}, \quad m = 1, \dots, n, \quad b_{-1} = 0, \quad b_0 = 1$$

$$r = a_{n-1} - pb_{n-2} - qb_{n-3} = b_{n-1}, \quad s = a_n - qb_{n-2} = b_n + pb_{n-1}$$

$$c_m = b_m - pc_{m-1} - qc_{m-2}, \quad m = 1, \dots, n-1, \quad c_{-1} = 0, \quad c_0 = 1$$

$$c_{n-2}\Delta p + c_{n-3}\Delta q = b_{n-1} \quad \Delta p \equiv \Delta p^{(k)} = p^{(k+1)} - p^{(k)}$$

$$c'_{n-1}\Delta p + c_{n-2}\Delta q = b_n \quad \Delta p \equiv \Delta q^{(k)} = q^{(k+1)} - q^{(k)}, \quad k = 0, 1, \dots$$

Sistemi nelinearnih jednačina

$$\mathbf{F}(\mathbf{x}) = \mathbf{0}, \quad \mathbf{x} \in \mathcal{R}^n$$

$$\mathbf{F}(\mathbf{x}) = \begin{pmatrix} f_1(\mathbf{x}) \\ \vdots \\ f_n(\mathbf{x}) \end{pmatrix}, \quad \mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad \mathbf{F}'(\mathbf{x}) = J(\mathbf{F}(\mathbf{x})) = \begin{pmatrix} \frac{\partial f_1(\mathbf{x})}{\partial x_1} & \cdots & \frac{\partial f_1(\mathbf{x})}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n(\mathbf{x})}{\partial x_1} & \cdots & \frac{\partial f_n(\mathbf{x})}{\partial x_n} \end{pmatrix}$$

Newton-ova metoda

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - [\mathbf{F}'(\mathbf{x}^{(k)})]^{-1} \mathbf{F}(\mathbf{x}^{(k)})$$

$$\|\mathbf{F}'(\mathbf{x}) - \mathbf{F}'(\mathbf{y})\| \leq \gamma \|\mathbf{x} - \mathbf{y}\| \quad \lim_{k \rightarrow \infty} \mathbf{x}^{(k)} = \bar{\mathbf{x}}, \quad \mathbf{F}(\bar{\mathbf{x}}) = \mathbf{0}$$

$$[\mathbf{F}'(\mathbf{x})]^{-1} \text{ postoji i } \|\mathbf{F}'(\mathbf{x})\|^{-1} \leq \beta \quad \longrightarrow$$

$$\|[\mathbf{F}'(\mathbf{x}^{(0)})]^{-1} \mathbf{F}(\mathbf{x}^{(0)})\| \leq \alpha$$

greška

$$\|\bar{\mathbf{x}} - \mathbf{x}^{(k)}\| \leq \alpha \frac{h^{2^k - 1}}{1 - h^{2^k}}$$

$$\mathbf{x}, \mathbf{y} \text{ u konveksnom skupu } C \in \mathcal{X}, \quad h = \frac{\alpha\beta\gamma}{2} < 1$$

Metoda iteracije $F(\mathbf{x}) = 0 \iff \mathbf{x} = \mathbf{G}(\mathbf{x}) \equiv \{g_i(\mathbf{x})\}_{i=1}^n$

$$\mathbf{x}^{(0)}, \quad \mathbf{x}^{(k+1)} = \mathbf{G}(\mathbf{x}^{(k)}), \quad k = 0, 1, \dots,$$

$$q = \max_x \|J(\mathbf{G}(\mathbf{x}))\| < 1 \quad \longrightarrow \quad \lim_{k \rightarrow \infty} \mathbf{x}^{(k)} = \bar{\mathbf{x}}, \quad \bar{\mathbf{x}} = \mathbf{G}(\bar{\mathbf{x}})$$

greška $\|\bar{\mathbf{x}} - \mathbf{x}^{(k)}\| \leq \varepsilon$ za $\|\mathbf{x}^{(k)} - \mathbf{x}^{(k-1)}\| \leq \frac{1-q}{q} \varepsilon$ iii $k \geq \frac{\ln \frac{\varepsilon(1-q)}{\|\mathbf{G}(\mathbf{x}^{(0)}) - \mathbf{x}^{(0)}\|}}{\ln q}$

Gradijentne metode $F(\bar{\mathbf{x}}) = 0 \iff F(\bar{\mathbf{x}}) = \min F(\mathbf{x})$

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \lambda_k \mathbf{d}^{(k)}, \quad \lambda_k = -\frac{\nabla F(\mathbf{x}^{(k)}) \cdot \mathbf{d}^{(k)}}{(\mathbf{d}^{(k)})^\top A(\mathbf{x}^{(k)}) \mathbf{d}^{(k)}}, \quad k = 0, 1, \dots$$

$$F(\mathbf{x}^{(k)} + \lambda_k \mathbf{d}^{(k)}) \approx F(\mathbf{x}^{(k)}) + \lambda_k \nabla F(\mathbf{x}^{(k)}) \cdot \mathbf{d}^{(k)} + \frac{1}{2} \lambda_k^2 (\mathbf{d}^{(k)})^\top A(\mathbf{x}^{(k)}) \mathbf{d}^{(k)}$$

Sopstvene vrednosti i vektori

$$A \mathbf{x}_k = \lambda_k \mathbf{x}_k$$

$$D(\lambda_k) = \det(A - \lambda_k I) = (-1)^n (\lambda_k^n + p_1 \lambda_k^{n-1} + \dots + p_n) = 0$$

Interpolacija

$$D(\lambda) \equiv \sum_{i=0}^n \left(\prod_{\substack{j=0 \\ j \neq i}}^n \frac{x - \mu_j}{\mu_i - \mu_j} \right) D(\mu_i), \quad |\mu_i| \leq \|A\|$$

Le Verrier

$$p_k = -\frac{1}{k} (S_k + p_1 S_{k-1} + p_2 S_{k-2} + \dots + p_{k-1} S_1), \quad k = 2, \dots, n.$$

$$p_1 = -S_1, \quad S_k = \text{tr}(A^k) = \sum_{i=1}^n a_{i,i}^{(k)} = \lambda_1^k + \dots + \lambda_n^k$$

Krilov

$$b_i^{(n-1)} p_1 + b_i^{(n-2)} p_2 + \dots + b_i^{(0)} p_n = -b_i^{(n)}, \quad i = 1, \dots, n,$$

$$\mathbf{b}^{(k)} = (b_1^{(k)}, b_2^{(k)}, \dots, b_n^{(k)})^T = A \mathbf{b}^{(k-1)} = A^k \mathbf{b}^{(0)}, \quad \mathbf{b}^{(0)} \text{ proizvoljno}$$

Danilevski

$$T^{-1} A T = P = \begin{pmatrix} p_1 & \dots & p_{n-1} & p_n \\ 1 & & 0 & 0 \\ & \dots & & \\ 0 & & 1 & 0 \end{pmatrix}, \quad P \mathbf{y}_k = \lambda_k \mathbf{y}_k, \quad k = 1, \dots, n,$$
$$\mathbf{x}_k = T \mathbf{y}_k,$$

Givens-ova metoda rotacije

(U matrica rotacije, $U^{-1} = U^*$)

$$A \sim A' = U_{n-1,n}^* \cdots U_{24}^* U_{23}^* A U_{23} U_{24} \cdots U_{n-1,n} = \begin{pmatrix} * & \cdots & \cdots & * \\ * & \ddots & & \vdots \\ 0 & & * & * \end{pmatrix}$$

$$B = U_{k,l}^* A U_{k,l}, \quad U_{kl} = \begin{pmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & \cos \phi & & -e^{-i\psi} \sin \phi & \\ & & e^{i\psi} \sin \phi & \cdots & \cos \phi & \\ & & & & & \ddots & \\ & & & & & & 1 \end{pmatrix}$$

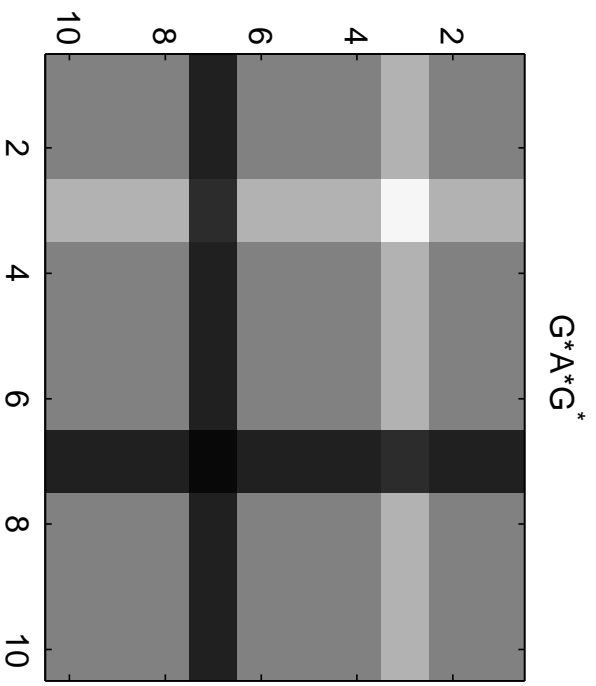
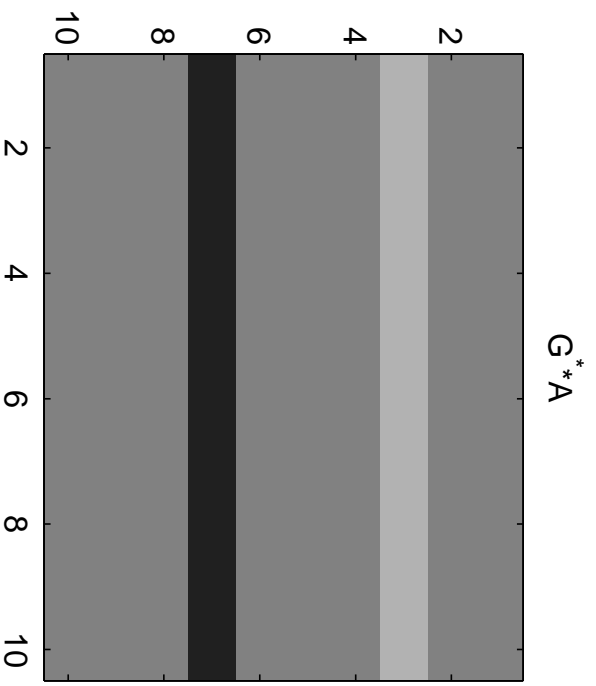
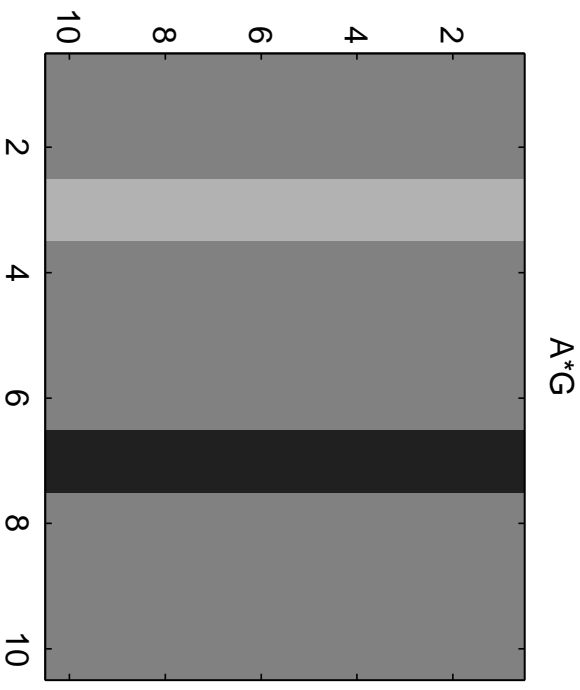
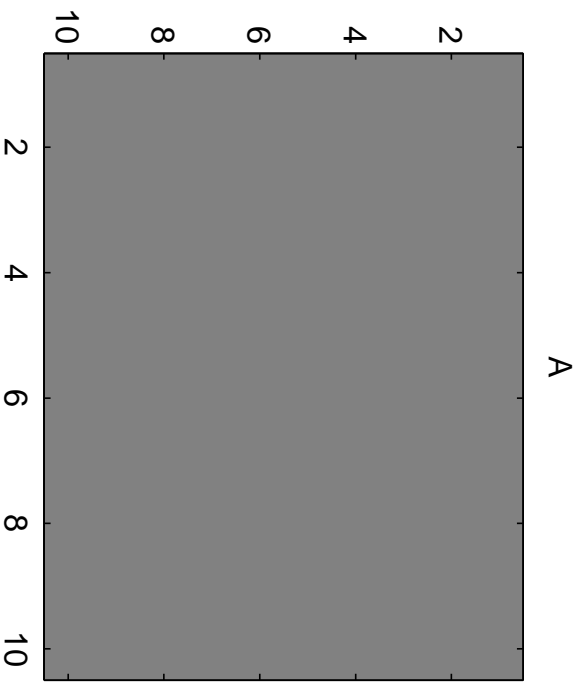
$\leftarrow k$
 $\leftarrow l$

$$\cos \phi = \frac{|a_{k,k-1}|}{\sqrt{|a_{k,k-1}|^2 + |a_{l,k-1}|^2}},$$

$$e^{-i\psi} \sin \phi = \frac{a_{l,k-1}}{a_{k,k-1}} \alpha$$

\longrightarrow

$$b_{l,k-1} = 0$$



Jacobijeva metoda

$$A^* = A$$

$$\lim_{m \rightarrow \infty} A_m = D = \text{diag}(\lambda_1, \dots, \lambda_n)$$

$$A_m = U_m^T \cdots U_1^T A U_1 \cdots U_m$$

$$\lim_{m \rightarrow \infty} U_1 \cdots U_m = U = (\mathbf{x}_1 \ \cdots \ \mathbf{x}_n)$$

$$U_m \equiv U_{kl} = \begin{pmatrix} 1 & & & & & & & & & \\ & \ddots & & & & & & & & \\ & & \cos \phi & & & & & & & \\ & & \sin \phi & & & & & & & \\ & & & \ddots & & & & & & \\ & & & & \cos \phi & & & & & \\ & & & & \sin \phi & & & & & \\ & & & & & \ddots & & & & \\ & & & & & & & & & 1 \end{pmatrix} \begin{matrix} \leftarrow k \\ \leftarrow l \\ \leftarrow l \\ \leftarrow k \\ \leftarrow l \\ \leftarrow k \\ \leftarrow l \\ \leftarrow k \end{matrix}$$

$$\tan 2\phi = \frac{2a_{kl}}{a_{kk} - a_{ll}}$$

$$\longrightarrow a_{kl}^{(m)} = 0$$

$$A_m = \{a_{i,j}^{(m)}\}, \quad v_m^2 = \sum_{\substack{i,j=1 \\ i \neq j}}^n a_{i,j}^{(m)2}, \quad v^2 = \sum_{\substack{i,j=1 \\ i \neq j}}^n a_{i,j}^2, \quad \sigma \geq n, \quad v_m^2 \leq v^2 \left(1 - \frac{2}{\sigma^2}\right)^m \xrightarrow{m \rightarrow \infty} 0$$

tačnost $v_m \leq \epsilon v$ za $a_{i,j}^{(m)} \leq \frac{\epsilon}{n} v$ ili $m \geq \frac{2 \ln \epsilon}{\ln(1 - \frac{2}{\sigma^2})}$

Householder-ova metoda

$$A \sim A' = Q^* A Q = T_{n-2} \dots T_1 A T_1 \dots T_{n-2} = \begin{pmatrix} * & \dots & \dots & * \\ * & \dots & \dots & \vdots \\ 0 & \dots & * & * \end{pmatrix}$$

$$T_m = \begin{pmatrix} 1 & \dots & 0 & 0 & \dots & 0 \\ 0 & \dots & \vdots & 0 & \dots & \vdots \\ 0 & \dots & 1 & 0 & \dots & 0 \\ \vdots & \dots & \vdots & \vdots & \dots & \vdots \\ 0 & \dots & 0 & 0 & \dots & 0 \end{pmatrix}, \quad Q = T_1 \dots T_{n-2}$$

$$M = \dim(H_m) = n - m, \quad \mathbf{x}_m = \begin{pmatrix} a_{m+1,m}^{(m-1)} & \dots & a_{n,m}^{(m-1)} \end{pmatrix}^T \quad \left(A_m = \left\{ a_{i,j}^{(m)} \right\} = T_m A_{m-1}, A_0 \equiv A \right)$$

$$\mathbf{x} = (x_1, \dots, x_M)^T, \quad \sigma = \sqrt{\sum_{i=1}^M |x_i|^2}, \quad x_1 = |x_1| e^{i\alpha}, \quad k = -\sigma e^{i\alpha}, \quad \beta = (\sigma(\sigma + |x_1|))^{-1},$$

$$\mathbf{u} = \mathbf{x} - k \mathbf{e}_1, \quad H = I - \beta \mathbf{u} \mathbf{u}^* \quad (H = H^* = H^{-1})$$

L R metoda

$$A_i = L_i R_i, \quad L_i = \begin{pmatrix} 1 & & & 0 \\ & \ddots & & \\ & & \ddots & \\ * & \dots & & 1 \end{pmatrix}, \quad R_i = \begin{pmatrix} * & \dots & * \\ & \ddots & \\ 0 & & \dots & * \end{pmatrix}$$

$$A_{i+1} = R_i L_i = L_{i+1} R_{i+1}, \quad i = 1, 2, \dots$$

$$A_{i+1} = L_i^{-1} A_i L_i = (L_1 \dots L_i)^{-1} A_1 (L_1 \dots L_i), \quad A^i \equiv A_1^i = T_i U_i = (L_1 \dots L_i) (R_i \dots R_1)$$

Teorema:

$$\lim_{i \rightarrow \infty} A_i = \lim_{i \rightarrow \infty} R_i = \begin{pmatrix} \lambda_1 & \dots & * \\ & \ddots & \vdots \\ 0 & & \lambda_n \end{pmatrix}, \quad \lim_{i \rightarrow \infty} L_i = I \quad \text{ako}$$

- $L R$ dekompozicije $A_i = L_i R_i$ postoje za svako $i = 1, 2, \dots$,
- $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n|$,
- postoje dekompozicije $X = L_x R_x$, $Y = L_y R_y$, gde je

$$A = X D Y, \quad D = \text{diag}(\lambda_1, \dots, \lambda_n), \quad X = (\mathbf{x}_1, \dots, \mathbf{x}_n)^T, \quad Y = X^{-1}$$

$$\mathbf{QR} \text{ metoda} \quad A_i = Q_i R_i, \quad Q_i^{-1} = Q_i^*, \quad R_i = \begin{pmatrix} * & \cdots & * \\ & \ddots & \vdots \\ 0 & & * \end{pmatrix}$$

$$A_{i+1} = R_i Q_i = Q_{i+1} R_{i+1}, \quad i = 1, 2, \dots$$

$$A_{i+1} = Q_i^* A_i Q_i = (Q_1 \cdots Q_i)^* A_1 (Q_1 \cdots Q_i), \quad A^i \equiv A_1^i = P_i U_i = (Q_1 \cdots Q_i) (R_i \cdots R_1)$$

Teorema:

$$\begin{aligned} \lim_{k \rightarrow \infty} S_{k-1}^* Q_k S_k &= I, & \lim_{k \rightarrow \infty} S_{k-1}^* A_k S_{k-1} &= \lim_{k \rightarrow \infty} S_k^* R_k S_{k-1} = \begin{pmatrix} \lambda_1 & \cdots & * \\ & \ddots & \vdots \\ 0 & & \lambda_n \end{pmatrix} \\ S_k &= \text{diag}(e^{i\phi_1^{(k)}}, \dots, e^{i\phi_n^{(k)}}), & \lim_{k \rightarrow \infty} a_{jj}^{(k)} &= \lambda_j, \quad j = 1, \dots, n, & (A_k = \{a_{jj}^{(k)}\}) \end{aligned}$$

ako je $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n|$ i postoji dekompozicija $Y = L_y R_y$,

$$A = X D Y, \quad D = \text{diag}(\lambda_1, \dots, \lambda_n), \quad Y = X^{-1}, \quad L_y = \begin{pmatrix} 1 & & 0 \\ \vdots & \ddots & \vdots \\ * & \dots & 1 \end{pmatrix}, \quad R_y = \begin{pmatrix} * & \cdots & * \\ & \ddots & \vdots \\ 0 & & * \end{pmatrix}$$

Delimičan problem sopstvenih vrednosti

$$\lambda_1, \mathbf{x}_1$$

$$|\lambda_1| > |\lambda_2| \geq \dots \geq |\lambda_n|$$

Metoda proizvoljnog vektora

$$\lim_{k \rightarrow \infty} \frac{v_j^{(k+1)}}{v_j^{(k)}} = \lambda_1, \quad \lim_{k \rightarrow \infty} \mathbf{v}^{(k)} = \mathbf{x}_1$$

$$\mathbf{v}^{(k)} = \begin{pmatrix} v_1^{(k)} \\ \dots \\ v_n^{(k)} \end{pmatrix}^T, \quad \mathbf{v}^{(k)} = A \mathbf{v}^{(k-1)} = A^k \mathbf{v}^{(0)}, \quad \mathbf{v}^{(0)} \text{ proizvoljno}$$

Metoda skalarnog proizvoda

$$\lim_{k \rightarrow \infty} \frac{(\mathbf{v}^{(k)}, \mathbf{w}^{(k)})}{(\mathbf{v}^{(k-1)}, \mathbf{w}^{(k)})} = \lambda_1, \quad \lim_{k \rightarrow \infty} \mathbf{v}^{(k)} = \mathbf{x}_1$$

$$\mathbf{v}^{(k)} = A \mathbf{v}^{(k-1)} = A^k \mathbf{v}^{(0)}, \quad \mathbf{w}^{(k)} = A^* \mathbf{w}^{(k-1)} = (A^*)^k \mathbf{w}^{(0)}, \quad \mathbf{v}^{(0)}, \mathbf{w}^{(0)} \text{ proizvoljno}$$

Metoda tragova

$$|\lambda_1| = \lim_{k \rightarrow \infty} \sqrt[k]{|\operatorname{tr}(A^k)|}, \quad \lambda_1 = \lim_{k \rightarrow \infty} \frac{\operatorname{tr}(A^{k+1})}{\operatorname{tr}(A^k)}, \quad \operatorname{tr}(A^k) = \sum_{i=1}^n a_{i,i}^{(k)}$$

$$\mathbf{x}_1 \approx A^k \mathbf{v}^{(0)}, \quad \mathbf{v}^{(0)} \text{ proizvoljno}$$

Metoda iscrpljivanja

$$A_1 = A - \lambda_1 \mathbf{x}_1 \mathbf{y}_1^*$$

$$A_1 \mathbf{x}_1 = \mathbf{0}, \quad A_1 \mathbf{x}_k = \lambda_k \mathbf{x}_k, \quad k = 2, \dots, n.$$

$$A^* \mathbf{y}_k = \bar{\lambda}_k \mathbf{y}_k, \quad (\mathbf{x}_1, \mathbf{y}_1) = \mathbf{y}_1^* \mathbf{x}_1 = 1$$