

Tehtävä 1 on kotitehtävä. Kotitehtävä palautetaan laskuharjoituksiin mennessä huoneen Y323b edessä sijaitsevaan lokeroon tai laskuharjoitusten alussa assistentille.

- (Kotitehtävä, palautus 26.1.) (Problem 1.1 p. 53) Show that $f_x(x^0) \in \mathbb{R}^{n \times m}$ is onto if and only if it has a right inverse $f_x(x^0)^I \in \mathbb{R}^{m \times n}$
- (Remark 1 p. 56) Show that, if \mathbf{A} has a right inverse \mathbf{A}^I and $\mathbf{I} + \mathbf{V}^T \mathbf{A}^I \mathbf{U}$ is invertible, then $\mathbf{A}^I - \mathbf{A}^I \mathbf{U} (\mathbf{I} + \mathbf{V}^T \mathbf{A}^I \mathbf{U})^{-1} \mathbf{V}^T \mathbf{A}^I$ is a right inverse of $\mathbf{A} + \mathbf{U} \mathbf{V}^T$.
- (Problem 2.1 p.57) Suppose we have a QR-factorization of the matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$, and want to find a QR-factorization of $\mathbf{A} + \mathbf{u} \mathbf{v}^T$. Show how to achieve this with Givens rotations at a cost of $O(mn)$ flops. Conclude from this that one can implement Broyden's method by updating a sequence of QR-factorizations of matrices differing always by a rank one correction.
- (Problem 5.4 p.70) Let $\mathbf{A} \in \mathbb{R}^{n \times n}$ be invertible and $\mathbf{b}, \mathbf{c} \in \mathbb{R}^n$. Show the following handy identity

$$\det(\mathbf{A} + \mathbf{b} \mathbf{c}^T) = (1 + \mathbf{c}^T \mathbf{A}^{-1} \mathbf{b}) \det \mathbf{A}.$$

Then, deduce that if \mathbf{b} is an eigenvector of \mathbf{A} such that $\mathbf{b}^T \mathbf{c} = 0$, then \mathbf{A} and $\mathbf{A} + \mathbf{b} \mathbf{c}^T$ have the same eigenvalues.

- (Problem 5.7 p.71) The inverse of $\mathbf{f}_{\mathbf{u}}(\mathbf{u}_0)|_{W^\perp}$ in the proof of Theorem 4.1 is called the *Moore–Penrose pseudoinverse* of $\mathbf{f}_{\mathbf{u}}(\mathbf{u}_0)$ (restricted to the subspace W^\perp). Generally, for a matrix \mathbf{A} it is defined as follows. If $\mathbf{A} \in \mathbb{C}^{n \times m}$ then $N(\mathbf{A})^\perp$ and $R(\mathbf{A})$ have the same dimension (i.e., $\text{rank}(\mathbf{A})$) and the pseudoinverse of \mathbf{A} , denoted by \mathbf{A}^\dagger , is defined to be the inverse of

$$\mathbf{A}|_{N(\mathbf{A})^\perp} : N(\mathbf{A})^\perp \rightarrow R(\mathbf{A}).$$

- Let $\mathbf{A} \in \mathbb{C}^{m \times p}$, $m \geq p$. Show that $\mathbf{x} = \mathbf{A}^\dagger \mathbf{b}$ is the solution of the least-squares problem $\min_{\mathbf{x}} \|\mathbf{A} \mathbf{x} - \mathbf{b}\|$, that has the smallest norm. As usual, the norms are Euclidean. In particular, if \mathbf{A} has full rank, then $\mathbf{A}^\dagger = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T$. Also, show that $(\mathbf{A}^T)^\dagger = (\mathbf{A}^\dagger)^T$.

- Let $\mathbf{A} \in \mathbb{C}^{m \times p}$, $m \geq p$, have rank r . Show that

$$\mathbf{A}^\dagger = \lim_{\varepsilon \rightarrow 0^+} (\mathbf{A}^* \mathbf{A} + \varepsilon \mathbf{I})^{-1} \mathbf{A}^* = \lim_{\varepsilon \rightarrow 0^+} \mathbf{A}^* (\mathbf{A} \mathbf{A}^* + \varepsilon \mathbf{I})^{-1} = \mathbf{V} [\mathbf{S}^\dagger \quad \mathbf{0}] \mathbf{U}^*,$$

where $\mathbf{A} = \mathbf{U} \begin{bmatrix} \mathbf{S} \\ \mathbf{0} \end{bmatrix} \mathbf{V}^*$ is the singular value decomposition of \mathbf{A} : $\mathbf{S} \in \mathbb{R}^{p \times p}$, $\mathbf{S} = \text{diag}(\sigma_1, \dots, \sigma_r, 0, \dots, 0)$, $\sigma_i \neq 0$, $i = 1, \dots, r$, and $\mathbf{S}^\dagger = \text{diag}(1/\sigma_1, \dots, 1/\sigma_r, 0, \dots, 0)$.