

P.S #4 Solutions

Problem 15.1

$$\begin{aligned} a) \tilde{f}(x) &= fl(x) \oplus fl(x) = [x(1+\varepsilon_1) + x(1+\varepsilon_1)](1+\varepsilon_2) \\ &= 2x(1+\varepsilon_1)(1+\varepsilon_2) = 2x(1+\varepsilon_3) = 2\tilde{x} \\ &= f(\tilde{x}) \end{aligned}$$

Now since $\frac{|x - \tilde{x}|}{|x|} = O(\varepsilon_{mach})$ we have
backward stability

$$\begin{aligned} b) \tilde{f}(x) &= fl(x) \otimes fl(x) = [x(1+\varepsilon_1) \times (1+\varepsilon_1)](1+\varepsilon_2) \\ &= x^2(1+\varepsilon_1)^2(1+\varepsilon_2) = [x(1+\varepsilon_3)]^2 = \tilde{x}^2 \\ &= f(\tilde{x}) \end{aligned}$$

Since $\frac{|\tilde{x} - x|}{|x|} = O(\varepsilon_{mach})$ we have

backward stability.

$$\begin{aligned} c) \tilde{f}(x) &= fl(x) \oslash fl(x) = \frac{x(1+\varepsilon_1)}{x(1+\varepsilon_1)}(1+\varepsilon_2) = 1+\varepsilon_2 \\ f(\tilde{x}) &= 1 \text{ for any } \tilde{x}. \end{aligned}$$

$$\frac{|\tilde{x} - x|}{|x|} = O(\varepsilon_{mach}) \text{ and } \frac{|\tilde{f}(x) - f(\tilde{x})|}{|f(\tilde{x})|} =$$

$= |1+\varepsilon_2 - 1| = |\varepsilon_2| = O(\varepsilon_{mach})$. So this is
stable but not backward stable.

$$\begin{aligned} d) \tilde{f}(x) &= fl(x) \ominus fl(x) = [x(1+\varepsilon_1) - x(1+\varepsilon_1)](1+\varepsilon_2) \\ &= 0 = f(\tilde{x}) \end{aligned}$$

$$\frac{|\tilde{x} - x|}{|x|} = O(\varepsilon_{mach}) \text{ so this is}$$

backward stable.

Problem 15.2 a) Backward stability means that the SVD matrices $\tilde{U}, \tilde{\Sigma}, \tilde{V}$ of A are the SVD matrices for the slightly perturbed matrix $A + \delta A$ where $\frac{\|\delta A\|}{\|A\|} = O(\varepsilon_{mach})$.

b) In general \tilde{U}, \tilde{V} are not unitary, they are just approximates for unitary.

c) A stable SVD factorization means

that the relative error between $\tilde{U}, \tilde{V}, \tilde{\Sigma}$ of a matrix A and the exact SVD of a perturbation $A + \delta A$ is of the order of $O(\varepsilon_{mach})$. This is:

$$\frac{\|\tilde{U}\tilde{\Sigma}\tilde{V}^* - (A + \delta A)\|}{\|A + \delta A\|} = O(\varepsilon_{mach}),$$

$$\frac{\|\delta A\|}{\|A\|} = O(\varepsilon_{mach}).$$

Problem 16.1

a) Take first $B = QA$, Q unitary.

$$\begin{aligned} \tilde{f}(A) &= (Q + \delta Q)(A + \delta A) = Q \underbrace{[Q^*(Q + \delta Q)(A + \delta A)]}_{\tilde{A}} \\ &= Q \tilde{A} \end{aligned}$$

We show that $\frac{\|\tilde{A} - A\|}{\|A\|} = O(\varepsilon_{mach})$.

$$\text{Now } \frac{\|\tilde{A} - A\|}{\|A\|} \leq \frac{\|\delta A\|}{\|A\|} + \frac{\|Q^*\delta QA\|}{\|A\|} + \frac{\|Q^*\delta Q\delta A\|}{\|A\|}$$

$$\text{Now } \frac{\|\delta A\|}{\|A\|} \leq O(\varepsilon_{mach})$$

$$(*) \frac{\|Q^*\delta QA\|}{\|A\|} \leq \|Q^*\| \|Q\| \cdot \frac{\|\delta QA\|}{\|Q\|} \cdot \frac{\|A\|}{\|A\|} = O(\varepsilon_{mach})$$

$$\frac{\|Q^*\delta Q\delta A\|}{\|A\|} \leq O(\varepsilon_{mach}).$$

So $\frac{\|\tilde{A} - A\|}{\|A\|} \leq O(\varepsilon_{mach})$ hence backward stability

Now for a general K we just have to apply an inductive procedure.

b) Q unitary allowed us to estimate $(*)$, namely we used $\|Q^*\| \|Q\| = 1$. If Q is replaced by X then $\|X^{-1}\| \|X\|$ might be huge.

Problem 16.2

$$a) [U, X] = \text{gr}(\text{randn}(50));$$

$$[V, X] = \text{gr}(\text{randn}(50));$$

$$S = \text{diag}(\text{fliplr}(\text{sort}(\text{rand}(50, 1))));$$

$$= U * S * V';$$

$$[U_2, S_2, V_2] = \text{svd}(A);$$

$$\text{norm}(U - U_2)$$

$$\text{norm}(V - V_2)$$

$$\text{norm}(S - S_2)$$

$$\text{norm}(A - U_2 * S_2 * V_2')$$

The errors in U and V are of magnitude 2.0 while the errors in singular values and A are ~~small~~ $\neq O(\epsilon_{\text{mach}})$. The problem with $U \neq V$ is that SVD factorization is not unique, since the columns of $U \neq U$ can be multiplied by -1 without changing the product.

1) To be able to compare the matrices U and V , the signs can be changed if necessary.

We can do this by:

$$\text{change} = \text{diag}(U_2' * U)$$

$$U_2 = U_2 * \text{diag}(\text{change})$$

$$V_2 = V_2 * \text{diag}(\text{change})$$