

Numerical Linear Algebra

Homework 7

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14.1. True or False?

(a) $\sin x = O(1)$ as $x \rightarrow \infty$.

(b) $\sin x = O(1)$ as $x \rightarrow 0$.

According to

The Meaning of $O(\epsilon_{\text{machine}})$

We now explain the precise meaning of " $O(\epsilon_{\text{machine}})$ " in (14.2)–(14.5).

The notation

$$\varphi(t) = O(\psi(t)) \quad (14.6)$$

is a standard one in mathematics, with a precise definition. This equation asserts that there exists some positive constant C such that, for all t sufficiently close to an understood limit (e.g., $t \rightarrow 0$ or $t \rightarrow \infty$),

$$|\varphi(t)| \leq C\psi(t). \quad (14.7)$$

it is very obvious that both (a) and (b) are true, since:

$$|\sin(x)| \leq 1 \quad \forall x$$

15.1. Each of the following problems describes an algorithm implemented on a computer satisfying the axioms (13.5) and (13.7). For each one, state whether the algorithm is *backward stable*, *stable but not backward stable*, or *unstable*, and prove it or at least give a reasonably convincing argument. Be sure to follow the definitions as given in the text.

(b) Data: $x \in \mathbb{C}$. Solution: x^2 , computed as $x \otimes x$.

Stability means:

is appropriate to aim for in general is *stability*. We say that an algorithm \tilde{f} for a problem f is *stable* if for each $x \in X$,

$$\frac{\|\tilde{f}(x) - f(\tilde{x})\|}{\|f(\tilde{x})\|} = O(\epsilon_{\text{machine}}) \quad (14.3)$$

for some \tilde{x} with

$$\frac{\|\tilde{x} - x\|}{\|x\|} = O(\epsilon_{\text{machine}}). \quad (14.4)$$

In words,

A stable algorithm gives nearly the right answer to nearly the right question.

Backward stability means:

stronger and simpler than stability. We say that an algorithm \tilde{f} for a problem f is *backward stable* if for each $x \in X$,

$$\tilde{f}(x) = f(\tilde{x}) \text{ for some } \tilde{x} \text{ with } \frac{\|\tilde{x} - x\|}{\|x\|} = O(\epsilon_{\text{machine}}). \quad (14.5)$$

This is a tightening of the definition of stability in that the $O(\epsilon_{\text{machine}})$ in (14.3) has been replaced by zero. In words,

A backward stable algorithm gives exactly the right answer to nearly the right question.

Examples are given in the next lecture.

The algorithm here is $\tilde{f}(x) := \text{float}(x) \otimes \text{float}(x)$
 For the problem $f(x) = x^2$ $\quad \quad \quad = \text{float}(\text{float}(x)^2)$

So for some $x \in X$ we have

$$\begin{aligned} \tilde{f}(x) &= (x(1+\epsilon_1))^2 (1+\epsilon_2) = x^2 (1+\epsilon_1)^2 (1+\epsilon_2) \\ &= x^2 (1+\epsilon_3) \end{aligned}$$

For some ϵ_3 with $|\epsilon_3| \leq 3\epsilon_{\text{machine}} + O(\epsilon_{\text{machine}}^3)$

In other words: The computed result $\tilde{f}(x)$ is exactly equal to $f(\tilde{x})$ where \tilde{x} satisfies

$$\frac{|\tilde{x} - x|}{|x|} = O(\epsilon_{\text{machine}})$$

And any $C > 0$ will suffice for the constant implicit in the "O" symbol.

Which means that the algorithm is backward stable. \square

17.3. Let $L \in \mathbb{C}^{m \times m}$ be a unit lower-triangular matrix (i.e., with diagonal entries equal to 1). For convenience, write L in the form

$$L = \begin{bmatrix} 1 & & & & \\ -l_{2,1} & 1 & & & \\ -l_{3,1} & -l_{3,2} & 1 & & \\ \vdots & \vdots & & \ddots & \\ -l_{m,1} & -l_{m,2} & -l_{m,3} & \cdots & 1 \end{bmatrix},$$

and define $M = L^{-1}$.

(a) Derive a formula for m_{ij} (which may involve other entries of M). Which entries of L does m_{ij} depend on?

We can solve this by forward substitution.

We can also already assume that M is also lower triangular, as proven already in Homework 1 Problem 1 (Textbook Problem 1.3).

So we have:

$$I = LM, \quad \underbrace{I_k = LM_k}_{\substack{\text{we can solve} \\ \text{each of these by forward sub}}}, \quad \begin{matrix} m_{ij} = 0 \\ \text{for } j > i, \\ m_{ii} = 1 \end{matrix}$$

$$\begin{matrix} \text{unit vector} \\ \downarrow \end{matrix} \quad I_k = \vec{e}_m = \sum_{i=1}^m (L^T)_i M_k I_i = \sum_{i=1}^m \left[\sum_{j=1}^i -l_{ij} m_{jk} \right] I_i$$

So we know only the $i=k$ term of the first sum will be non-trivial.

$$1 = \sum_{j=1}^k -l_{kj} m_{jk}, \quad \forall k: 1 \leq k \leq m$$

$$I = \sum_{j=1}^k -L_{kj} m_{jk}, \quad \forall k: 1 \leq k \leq m$$

We know already that $m_{jk} = 0$, $k > j$

that leaves: $I = - \underbrace{L_{kk}}_{=1} m_{kk} \Rightarrow m_{kk} = 1$, as expected

This was not actually necessary, but well. Using forward sub, we get:

$$m_{jk} = \delta_{jk} + \sum_{i=1}^{j-1} m_{ik} l_{ji}$$

The diagonal entries of L are 1, so no division needed!
 l_{ij} defined negative.
 Kronecker delta
 $\delta_{ij} = \begin{cases} 1, & i=j \\ 0, & i \neq j \end{cases}$

(a) Derive a formula for m_{ij} (which may involve other entries of M). Which entries of L does m_{ij} depend on?

As we've seen, the entries m_{ic} don't depend on L , they are always equal to one. So do the entries m_{ij} for $j > i$, they allways vanish.

The entries m_{ij} with $j < i$ depend on the $(i-1) \times (i-1)$ upper left sub-matrix of L , which is itself lower-triangular.