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Numerical Methods for CSE

Problem Sheet 4

Problem 1. Order of convergence from error recursion (core problem)

In [1, Exp. 2.3.26] we have observed *fractional* orders of convergence (\rightarrow [1, Def. 2.1.17]) for both the secant method, see [1, Code 2.3.25], and the quadratic inverse interpolation method. This is fairly typical for 2-point methods in 1D and arises from the underlying recursions for error bounds. The analysis is elaborated for the secant method in [1, Rem. 2.3.27], where a linearized error recursion is given in [1, Eq. (2.3.31)].

Now we suppose the recursive bound for the norms of the iteration errors

$$||e^{(n+1)}|| \le ||e^{(n)}|| \sqrt{||e^{(n-1)}||},$$
 (9)

where $e^{(n)} = x^{(n)} - x^*$ is the error of *n*-th iterate.

(1a) • Guess the maximal order of convergence of the method from a numerical experiment conducted in MATLAB.

HINT:[1, Rem. 2.1.19]

HINT: First of all note that we may assume equality in both the error recursion (9) and the bound $||e^{(n+1)}|| \le C||e^{(n)}||^p$ that defines convergence of order p > 1, because in both cases equality corresponds to a worst case scenario. Then plug the two equations into each other and obtain an equation of the type ... = 1, where the left hand side involves an error norm that can become arbitrarily small. This implies a condition on p and allows to determine C > 0. A formal proof by induction (not required) can finally establish that these values provide a correct choice.

Problem 2. Convergent Newton iteration (core problem)

As explained in [1, Section 2.3.2.1], the convergence of Newton's method in 1D may only be local. This problem investigates a particular setting, in which global convergence can be expected.

We recall the notion of a *convex function* and its geometric definition. A differentiable function $f:[a,b]\mapsto \mathbb{R}$ is convex, if and only if its graph lies on or above its tangent at any point. Equivalently, differentiable function $f:[a,b]\mapsto \mathbb{R}$ is convex, if and only if its derivative is non-decreasing.

Give a "graphical proof" of the following statement:

If F(x) belongs to $C^2(\mathbb{R})$, is strictly increasing, is convex, and has a unique zero, then the Newton iteration [1, (2.3.4)] for F(x) = 0 is well defined and will converge to the zero of F(x) for any initial guess $x^{(0)} \in \mathbb{R}$.

Problem 3. The order of convergence of an iterative scheme (core problem)

[1, Rem. 2.1.19] shows how to detect the order of convergence of an iterative method from a numerical experiment. In this problem we study the so-called Steffensen's method, which is a derivative-free iterative method for finding zeros of functions in 1D.

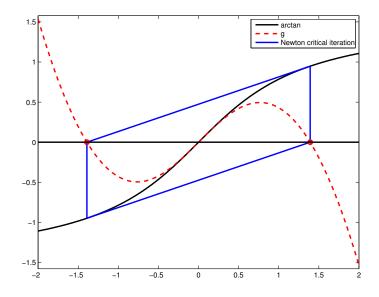
Let $f : [a, b] \to \mathbb{R}$ be twice continuously differentiable with $f(x^*) = 0$ and $f'(x^*) \neq 0$. Consider the iteration defined by

$$x^{(n+1)} = x^{(n)} - \frac{f(x^{(n)})}{g(x^{(n)})}, \text{ where } g(x) = \frac{f(x+f(x)) - f(x)}{f(x)}.$$

- (3a) \Box Write a MATLAB script that computes the order of convergence to the point x^* of this iteration for the function $f(x) = xe^x 1$ (see [1, Exp. 2.2.3]). Use $x^{(0)} = 1$.
- (3b) \Box The function g(x) contains a term like e^{xe^x} , thus it grows very fast in x and the method can not be started for a large $x^{(0)}$. How can you modify the function f (keeping the same zero) in order to allow the choice of a larger initial guess?

HINT: If f is a function and $h : [a, b] \to \mathbb{R}$ with $h(x) \neq 0, \forall x \in [a, b]$, then $(fh)(x) = 0 \Leftrightarrow f(x) = 0$.

Figure 3: Newton iterations with $F(x) = \arctan(x)$ for the critical initial value $x^{(0)}$



Problem 4. Newton's method for $F(x) := \arctan x$

The merely local convergence of Newton's method is notorious, see[1, Section 2.4.2] and [1, Ex. 2.4.46]. The failure of the convergence is often caused by the overshooting of Newton correction. In this problem we try to understand the observations made in [1, Ex. 2.4.46].

(4a) \square Find an equation satisfied by the smallest positive initial guess $x^{(0)}$ for which Newton's method does not converge when it is applied to $F(x) = \arctan x$.

HINT: Find out when the Newton method oscillates between two values.

HINT: Graphical considerations may help you to find the solutions. See Figure 3: you should find an expression for the function g.

(4b) Use Newton's method to find an approximation of such $x^{(0)}$, and implement it with Matlab.

Problem 5. Order-*p* convergent iterations

In [1, Section 2.1.1] we investigated the speed of convergence of iterative methods for the solution of a general non-linear problem $F(\mathbf{x}) = 0$ and introduced the notion of conver-

gence of order $p \ge 1$, see [1, Def. 2.1.17]. This problem highlights the fact that for p > 1 convergence may not be guaranteed, even if the error norm estimate of [1, Def. 2.1.17] may hold for some $\mathbf{x}^* \in \mathbb{R}^n$ and all iterates $\mathbf{x}^{(k)} \in \mathbb{R}^n$.

Suppose that, given $\mathbf{x}^* \in \mathbb{R}^n$, a sequence $\mathbf{x}^{(k)}$ satisfies

$$\|\mathbf{x}^{(k+1)} - \mathbf{x}^*\| \le C \|\mathbf{x}^{(k)} - \mathbf{x}^*\|^p \quad \forall k \text{ and some } p > 1.$$

(5a) \odot Determine $\epsilon_0 > 0$ such that

$$\|\mathbf{x}^{(0)} - \mathbf{x}^*\| \le \epsilon_0 \qquad \Longrightarrow \qquad \lim_{k \to \infty} \mathbf{x}^{(k)} = \mathbf{x}^*.$$

In other words, ϵ_0 tells us which distance of the initial guess from \mathbf{x}^* still guarantees local convergence.

(5b) Provided that $\|\mathbf{x}^{(0)} - \mathbf{x}^*\| < \epsilon_0$ is satisfied, determine the minimal $k_{\min} = k_{\min}(\epsilon_0, C, p, \tau)$ such that

$$\|\mathbf{x}^{(k)} - \mathbf{x}^*\| < \tau.$$

(5c) • Write a MATLAB function

$$k_{min} = @ (epsilon, C, p, tau) ...$$

and plot $k_{\min} = k_{\min}(\epsilon_0, \tau)$ for the values p = 1.5, C = 2. Test you implementation for every $(\epsilon_0, \tau) \in \text{linspace}(0, C^{\frac{1}{1-p}})^2 \cap (0, 1)^2 \cap \{(i, j) \mid i \geq j\}$

HINT: Use a MATLAB poolor plot and the commands linspace and meshgrid.

Problem 6. Code quiz

A frequently encountered drudgery in scientific computing is the use and modification of poorly documented code. This makes it necessary to understand the ideas behind the code first. Now we practice this in the case of a simple iterative method.

(6a) What is the purpose of the following MATLAB code?

```
function y = myfn(x)
log2 = 0.693147180559945;
```

```
while (x > sqrt(2)), x = x/2; y = y + log2; end
while (x < 1/sqrt(2)), x = x*2; y = y - log2; end
z = x-1;
dz = x*exp(-z)-1;
while (abs(dz/z) > eps)
z = z+dz;
dz = x*exp(-z)-1;
end
y = y+z+dz;
```

- **(6b)** Explain the rationale behind the two while loops in lines #5, 6.
- (6c) Explain the loop body of lines #10, 11.
- **(6d)** Explain the conditional expression in line #9.
- **(6e)** Replace the while-loop of lines #9 through #12 with a fixed number of iterations that, nevertheless, guarantee that the result has a relative accuracy eps.

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References

[1] R. Hiptmair. *Lecture slides for course "Numerical Methods for CSE"*. http://www.sam.math.ethz.ch/~hiptmair/tmp/NumCSE/NumCSE15.pdf. 2015.