

Introduction to Numerical Methods Tutorial 12

Exercise 1:

The task to find a fixed point $x^* \in \mathbb{R}^2$ with $g(x^*) = x^*$ for

$$g : \mathbb{R}^2 \rightarrow \mathbb{R}^2, \quad \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mapsto \begin{pmatrix} \frac{1}{2}x_1 + \frac{1}{4}x_2 + 1 \\ \frac{2}{3}x_1 + \frac{1}{6}x_2 - 3 \end{pmatrix}$$

should be considered step-by-step in this exercise.

- (i) Show that g is contractive on \mathbb{R}^2 with respect to either the "column-sum-norm", i.e., $\|A\|_1 := \max_{j=1, \dots, n} \sum_{i=1}^n |a_{ij}|$, or the "row-sum-norm", i.e., $\|A\|_\infty := \max_{j=1, \dots, n} \sum_{i=1}^n |a_{ji}|$. This means, you have to show that it is contractive for one of these two norms and not contractive for the other one.
- (ii) Compute how many fixed point iteration steps you need to guarantee an accuracy of your solution of 10^{-2} when starting with $x^{(0)} = (0, 0)^T$, i.e., for which $k \in \mathbb{N}$ will $\|x^{(k)} - x^*\|_\infty \leq 10^{-2}$ if x^* denotes the unknown solution.
- (iii) Compute 5 steps of the fixed point iteration with initial guess $x^{(0)} = (0, 0)^T$. How good is your result? I.e., compare $x^{(5)}$ to the analytic solution.

Exercise 2:

Let $g : \mathbb{R}_+ \rightarrow \mathbb{R}, x \mapsto x + \ln(x)$ be given, with $\mathbb{R}_+ := \{x \in \mathbb{R} : x > 0\}$.

- (i) Show that $g(x) = 0$ admits a solution. You should show this analytically, i.e., a graphical solution is not sufficient.
- (ii) Determine an approximate solution using a graphic.
- (iii) Which of the following three methods would solve the equation and which one would you expect to solve it with least steps?
 - (1) $x^{(k+1)} = \ln(x^{(k)})$

$$(2) \quad x^{(k+1)} = e^{-x^{(k)}}$$

$$(3) \quad x^{(k+1)} = \frac{x^{(k)} + e^{-x^{(k)}}}{2}$$

- (iv) Make 5 fixed point iteration steps for each method in (ii) which can solve the equation. Use $x^{(0)} = 0.5$ and compute $|x^{(j)} - x^{(j-1)}|$ in every step. What do you discover? Use 4 digits of accuracy for your computation.

Exercise 3:

Let the nonlinear equation system

$$F(x) = F(x_1, x_2) = \begin{pmatrix} x_1^3 + x_2^3 - 4 \\ x_1^3 - x_2^3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

be given.

- (i) Show that the Newton iteration to solve $F(x) = 0$ converges for every initial guess $x^{(0)} \in [1, 2] \times [1, 2]$.

Hint: Use the $\|\cdot\|_1$ -norm and a theorem from the lecture.

- (ii) Compute $x^{(1)}$ and $x^{(2)}$ using Newton iteration with the initial guess $x^{(0)} = (1, 1)^T$.

(*)-Exercise 4: (6 points)

Show that linear equation systems, i.e., problems of the form find $x \in \mathbb{R}^n$ such that $Ax = b$ with $A \in \mathbb{R}^{n \times n}$ independent of x and $b \in \mathbb{R}^n$ independent of x , are solved within one step by the Newton iteration, i.e., $x^{(1)}$ is the solution of the system for every initial guess $x^{(0)}$.

Programming Exercise 4: (delivery date: 27.01.2011, 8 points)

Write a program which computes the solution of the system given in Exercise 3 up to an accuracy of 10^{-7} . Remember that you should not use the inverse in your programs.

Make a monotonicity test in every step with $L = 0.5$, i.e., test if the difference between the iterates converges monotone to zero by checking if the quotient between the current difference and the difference of the last step is smaller than L . You have two options what to do when the program is aborted because of missing monotonicity

1. the program may ask the user for a new initial guess and restart
2. the program determines a new initial guess near to the old one and restarts.

Try different initial guesses for your program to test it and the monotonicity check. Give a documentation of the computing for at least three different initial guesses where one should be $(x_1^{(0)}, x_2^{(0)}) = (-0.3, 1.3)$.

The Newton method can algorithmically be described as follows.

```
 $\mathbf{x}^{(0)} = (\mathbf{x}_1^{(0)}, \mathbf{x}_2^{(0)})$            % initial guess - input parameter  
 $k = 1;$                                % iteration count  
 $\Delta \mathbf{x}^{(k)} = (\mathbf{J}_f(\mathbf{x}^{(k)}))^{-1} \cdot \mathbf{f}(\mathbf{x}^{(k)});$   
  
while(norm( $\Delta \mathbf{x}^{(k)}$ ) >  $10^{-7}$ )  
  
     $\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - (\mathbf{J}_f(\mathbf{x}^{(k)}))^{-1} \cdot \mathbf{f}(\mathbf{x}^{(k)});$   
  
     $\Delta \mathbf{x}^{(k+1)} = (\mathbf{J}_f(\mathbf{x}^{(k+1)}))^{-1} \cdot \mathbf{f}(\mathbf{x}^{(k+1)});$   
     $\Delta \mathbf{x}^{(k)} = (\mathbf{J}_f(\mathbf{x}^{(k)}))^{-1} \cdot \mathbf{f}(\mathbf{x}^{(k)});$   
  
     $\Delta = \frac{\text{norm}(\Delta \mathbf{x}^{(k+1)})}{\text{norm}(\Delta \mathbf{x}^{(k)})};$   
  
    if( $\Delta > L$ )  
        disp('no monotonicity ');  
        break;      % the program is aborted by leaving the while-loop  
    else  
         $k = k + 1;$   
    end  
end  
  
if(norm( $\Delta \mathbf{x}^{(k)}$ ) <=  $10^{-7}$ )  
    disp(['Solution at ' num2str( $\mathbf{x}^{(k)}$ )' .']);  
end
```

Delivery: 20. January 2011