TANA09 Datatekniska beräkningar

Laboration 1. Error analysis and equations

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1 Introduction

If we want to use a computer to perform mathematical calculations we have to approximate the set of real numbers a finite floating point number system. This means that the basic arithmetic operations can only be performed at a fixed level of precision. In this exercise we will investigate how these errors influence the results of numerical calculations performed using a computer.

Most often the errors introduced by the floating point system are small compared to other errors that appear in scientific computations. This is not always the case for instance if cancellation of significant digits occur.

One of the points of the exercise is to demonstrate that mathematically equivalent formulations of a problem may lead to different results when implemented on a computer. An analysis of the errors in numerical calculations is important since it allows the researcher to pick the specific formulation of a problem that leads to the smallest error in the result.

Note that many of the questions are of theoretical nature and can be answered after the scheduled time for the exercise.

2 Representation of Real Numbers

A floating point system is characterized by (β, t, L, U) , where β is the base, t is number of digits in the fractional part, L and U are the smallest and largest allowed exponents.

An upper limit for the relative error when storing a real number x in a floating point system (β, t, L, U) is given by

$$\left|\frac{\Delta x}{x}\right| \le \frac{1}{2} \cdot \beta^{-t} = \mu,$$

where μ is the unit roundoff, or the machine precision. We can interpret the number μ as follows:

$$f[1+y] = 1 \text{ if } y \in [0,\mu] \text{ and } f[1+y] > 1 \text{ if } y > \mu.$$

where f[1+y] is the result when storing the number 1+y in the floating point system. The unit roundoff is an upper limit for the error done when storing a real number in the floating point system (β, t, L, U) .

Exercise 2.1 The base is $\beta = 2$. Experiment with different values of n in the expression $1 + 2^{-n}$. For instance you can write

$$>> n=50$$
 , $1+2^-n$

What is μ and t?

$$\mu = t = t$$

Hint Matlab displays the result as 1.0000 unless the result is exactly 1.

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3 Numerical Computation of Limits

In this section we wills study the limit of a function. It is known that

$$\lim_{x \to 0} f_1(x) = 1$$
, where $f_1(x) = \frac{e^x - 1}{x}$.

Investigate how accurate it is possible to calculate the limit in Matlab by plotting the error (in log-scale) for a wide range of x-values.

The following Matlab code will create a vector x, that contains values between 1 and 10^{-16} , and compute the a vector f1 that contains the corresponding function values. The error $|f_1(x) - 1|$ is also displayed

```
>> x=10.^-(0:0.1:16);
>> f1=(exp(x)-1)./x;
>> loglog(x,abs(f1-1)), xlabel('x'), ylabel('|f_1(x)-1|')
```

Note the dot in front of the division on the second line. What is its purpose?

Exercise 3.1 Which value x gives the smallest error? What happens for very small values of x?

Answer:

Exercise 3.2 Investigate the accuracy when using the mathematically equivalent formula,

$$f_2(x) = \frac{e^x - 1}{\ln(e^x)}.$$

Compute a vector **f2** conatining the function values obtained using the new formula using the Matlab command

>>
$$f2=(exp(x)-1)./log(exp(x));$$

What happens now?

Answer:

Obviously the two equivalent mathematical experessions behave differently. In the following few exercises we will calculate the dominating terms of the errors $|f_1(x) - 1|$ and $|f_2(x) - 1|$. There are two contributing sources of error. The truncation error and the computational error.

Exercise 3.3 Use Taylor series expansion to estimate the truncation error,

$$R_T = |f_1(x) - 1| \le \underline{\hspace{1cm}}$$

The truncation error is the same for both expressions $f_1(x)$ and $f_2(x)$. Why is that?

Answer:

Exercise 3.4 Assume that all the numerical operations $(-,/,\exp, \text{ etc})$ are performed with a *relative* error less than, or equal to, the unit roundoff μ . Perform an analysis of the computational errors for both expressions $f_1(x)$ and $f_2(x)$. Find an estimates

$$R_X \le |\mathrm{fl}[f_1(x)] - f_1(x)|.$$

Answer:	
Exercise 3.5 Illustrate the results by plotting $ f_1(x)-1 $ and R_T+R_X , for x-values between 10^{-16} , in a log-log scale. Do the same for the expression $f_2(x)$. How does the error analysis co with the actual errors?	
Answer:	
Print and hand-in the graph together with your report.	

4 Solution of Non-Linear Equations

4.1 Fixed Point Iteration

Fixed point iteration is based on rewriting the equation f(x) = 0 in the form $x = \varphi(x)$, and computing a sequence,

$$x_{k+1} = \varphi(x_k), \qquad k = 0, 1, 2, \dots$$

The sequence $\{x_k\}_{k=0}^{\infty}$ will converge if $|\varphi'(x)| \leq m < 1$ close to the root x^* . A smaller value for $|\varphi'(x)|$ leads to faster convergence.

In this exercise we will use fixed point iteration to find the roots of the equation

$$f(x) = \cos^2(x) - x.$$

The following fixed point iterations can be used:

- (1) $x_{k+1} = \cos^2(x_k) = \varphi_1(x_k).$
- (2) $x_{k+1} = \arccos\sqrt{x_k} = \varphi_2(x_k).$

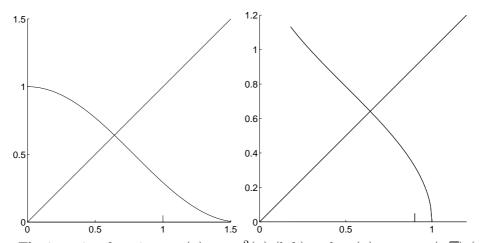
Exercise 4.1 Do the above fixed point iterations (1) and (2) converge? Present your calculations!

Answer:

Exercise 4.2 Compute the sequences $\{x_k\}$ for the fixed point formulas (1) and (2) above. First use $x_0 = 1$ and iterate $x_{k+1} = \varphi_1(x_k)$. Then use and $x_0 = 0.9$ and the iteration $x_{k+1} = \varphi_2(x_k)$. What happens?

Answer:

Illustrate the results by drawing the patterns for $x_{k+1} = \cos^2(x_k)$, $x_0 = 1$ and $x_{k+1} = \arccos(\sqrt{x_k})$, $x_0 = 0.9$ in the corresponding graphs.



The iteration functions $\varphi_1(x) = \cos^2(x)$ (left) and $\varphi_2(x) = \arccos(\sqrt{x})$ (right).

4.2 Newton Iteration

In this section we will find a root x^* for the equation

$$f(x) = x^2 - 4\sin(x).$$

In Matlab you can use a function handle to define the function f(x). Write

$$>> f=0(x) x.^2-4*sin(x);$$

Note that the function is written to accept vector arguments x. This makes it easier to plot the function. By plotting the function on the interval 0 < x < 4 we see that there is a positive root $x^* \approx 2$.

Newton's method for solving non-linear equations f(x) = 0 is defined by the following iteraton,

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x)}, \qquad k = 0, 1, 2, \dots$$

In the following exercises we will investigate the properties of the Newton iteration in detail.

Exercise 4.3 Calculate the derivative of the above function f(x) and create a corresponding inline function df.

Answer:		

Exercise 4.4 The Matlab function NewtonSolv computes the vector of iterates x_k given a starting approximation x_0 , inline functions defining f(x) and f'(x), and a tolerance tol. The program also computes the vector of quotients $e_k = f(x_k)/f'(x_k)$.

The iterations stop as soon as $|f(x_k)/f'(x_k)| < \text{tol.}$ Why is this a good stopping criteria?

Answer:

Use tol = 10^{-15} and $x_0 = 3$. Solve the equation f(x) = 0 using Newtons method. How many iterations are needed? What is the approximation \bar{x} of the root x^* ?

Hint Use format long to see the approximation x_k to full accuracy.

Exercise 4.5 Quadratic convergence means that

$$e_{k+1} \approx Ce_k^2, \qquad e_k = |x_k - x^*|.$$

Look at the error estimates $\{e_k\}$. Did we achive quadratic convergence? Also compare the error estimates with the actual errors $|x_k - x^*|$. Is the error estimate good?

Hint Assume that the final Newton iterate is more or less exact. Then a graph over the actual errors and the error estimates is obtained by typing

>> semilogy(1:length(x),abs(x-x(end)), 1:length(e), abs(e));

4.3 The fzero solver

In this exercise we will investigare the convergence speed for the built-in Matlab function fzero. Again we will attempt to solve the equation

$$f(x) = \sqrt{1+x}e^{x/2} - 2\sin(2x)(x+x^2).$$

We are intressted in the root $x^* \approx 1.5$.

Exercise 4.6 Plot the function f(x) on the interval [0,3]. Also use fzero to find the root x^* . Type the command you use and the result.

Answer:			
A newer.			

Exercise 4.7 In order to see more clearly what fzero does we write a Matlab function

```
function[y]=fun(x)
  y=sqrt(1+x).*exp(x/2)-2*sin(2*x).*(x+x.^2);
fprintf('%18.15f %6.2e\n',x,y);
end
```

This allows us to see exactly which x-values are used during the iterations. We also change the stopping criteria and solve the equation again by typing

```
>> options = optimset('TolFun',1e-15);
>> x = fzero( @fun , 1.5 , options );
```

How many function evaluations are needed to find the root?

Answer:	
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Exercise 4.8 Copy the x values used during the iterations into a vector x. If we assume that the final x-value is exact we can obtain a vector e containing the error in each step by typing e=abs(x-x(end)). Plot the errors using semilogy. Also write down the errors during the last 5 iterations. Does this look like linear or quadratic convergence?

Answer:			
TIID WCI.			

Hint Since the function just prints the x-value and f(x) on each row you can type x=[in the Matlab terminal. Then paste the numbers and end with];. To get the first column type x=x(:,1). To see the errors more clearly use format short e.

Exercise 4.9 If the order of convergence is p we have a relation $e_k \approx Ce_{k-1}^p$. Since the errors are known obtain a relation

 $\frac{e_k}{e_{k-1}} = (\frac{e_{k-1}}{e_{k-2}})^p.$

Taking logarithms we find an expression for p. Make use of the vector \mathbf{e} containing the errors and try to find a value for p. If you type

$$>> e,k=15,e(k),p=(log(e(k)/e(k-1))/log(e(k-1)/e(k-2)))$$

you can easily see which values are used to estimate p and its value. What is your experimentally determined value of p? What iteration index k was used?

Answer:			

Hint Typically for equation solvers the approximation x_k has to be sufficiently close to x^* for the rapid convergence to occur. This means that normally no significant improvements take place during the first iterations. Also once the approximation x_k is close enough the convergence is so fast that the unit round-off is reached within just two or three steps. This means that in practice it is difficult to estimate the order of convergence from experimental data.

4.4 Implementation of Division

When designing a computer chip you have to decide what operations should be supported by hardware and which ones that should be implemented as software. For instance the square root operation is sometimes included as a basic instruction in the hardware. On the other hand if you want to keep the instruction set small and the chip as simple as possible one might look into possible instructions that might as well be left out and implemented through software.

A basic arithmetic operation sometimes left out from hardware is floating point division. In this exercise we will look at how to do an efficient implementation of the division. The problem is really simple but great care has to be taken to reduce the work needed, and also to make sure the result is accurate.

Exercise 4.10 First compute z = 1/y. This can be formulated as a non-linear equaton,

$$f(z) = y - \frac{1}{z} = 0.$$

Apply Newtons method to the equation f(z) = 0 and derive the corresponding iteration formula. Take care not to use any divisions in your expression.

Answer:		

Remark One might try to solve for z = x/y directly but then its not possible to avoid divisions in the Newton formulas.

Exercise 4.11 On the computer the floating point numbers are of the form

$$y = \pm (1.f)_2 2^e$$
.

Since $(2^e)^{-1} = 2^{-e}$ we only need to compute 1/y for $1 \le y < 2$. This means that the root of the equation f(z) = 0 satisfies $\frac{1}{2} < z \le 1$. Suppose we use the starting approximation $z_0 = \frac{3}{4}$. What is the maximum initial error $e_0 = |z - z_0|$?

The Newton iteration can be written $z_{k+1} = \phi(z_k)$, where $\phi(z)$ is the iteration function. What is $\phi(z)$ here?

Exercise 4.12 The error analysis for the Newton iteration leads to a formula,

$$e_{k+1} = \frac{\phi''(\eta)}{2}e_k^2,$$

where η belongs to the interval (z_k, z) . Use the available information to find the maximum value of $|\phi''(\eta)|$. Also find the maximum number of iterations required until $|e_k| \leq \mu$, where μ is the unit roundoff. Present the calculations!

Exercise 4.13 Use the above results and implement a function

$$>> z = Division(x, y);$$

The function should solve the problem in two steps. First compute 1/y and then multiply $z = x \cdot (1/y)$. You can assume that $1 \le y < 2$.

Use your function to compute z = 1.32/y, for y=1:0.01:2. Plot the relative error in the result. In order to compute the errors you can assume that the Matlab division x/y is exact. Is the result good? How many multiplications are needed to compute one division?

Remark It is possible that the Newton method converges faster for certain values of y. However by implementing the worst case scenario we avoid conditional stataments in the loop. This has the advantage of ensuring that a division is equally fast regardless of the arguments x and y.

It is possible to improve the speed of the algorithm by picking a better starting value x_0 . For instance divide the interval [1,2) in 8 subintervals of equal length $1+[k/8,(k+1)/8), k=0,1,2,\ldots,7$. Then pick a starting approximation $z_{0,k}$ depending on which interval y belongs to. This was the initial error is at most $e_0 \leq 1/16$ and only N=4 iterations is needed. Thus we can trade memory (i.e. a table of starting approximations) for computational speed (i.e. number of multiplications). Though this is not convinient to write in Matlab.