

CS / MATH 4334 : Numerical Analysis

Homework Assignment 2

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MatLab Problems

Problem 1 : ffalpos.m

```
1 function c = ffalpos(a, b)
2 %FFALPOS Summary of this function goes here
3 % Detailed explanation goes here
4 fa = (a-2)^2 + 1 - (1 - (a-2)^2 / 4)^0.5 - 1;
5 fb = (b-2)^2 + 1 - (1 - (b-2)^2 / 4)^0.5 - 1;
6 c = b - fb*((b-a)/(fb-fa));
7 end
```

Problem 1 : falpos.m

```
1 %find root of f(x) = 0
2 %using Bisection Method
3
4 format long e
5
6 % chosen error tolerance (TOL)
7 TOL = .000001;
8
9 % choose max number of iterations
10 MAXIT = 50;
11
12 % initial bracket
13 % Calculated from simplifying (f-g)(x)
14 % There exists a root since f(a)f(b) < 0, and
15 % f(a) > 0, f(b) < 0;
16 a = 1;
17 b = 2;
18
19 %keep track of number of iterations
20 count = 0;
21
22 %record iterates - a col vector of MAXIT length
23 cits = zeros(MAXIT,1);
24
25 %evaluate func. at a and b
26 fa = 4*a^4 - 32*a^3 + 97*a^2 - 132*a + 64;
27 fb = 4*b^4 - 32*b^3 + 97*b^2 - 132*b + 64;
28
29 %stop if not appropriate interval
30 if sign(fa)*sign(fb) >= 0
31     fprintf("Not appropriate interval\n")
32     return
```

```

33
34 end
35
36 %stop loop when error less than TOL or MAXIT reached
37 while abs(b-a)/2 >= TOL && count < MAXIT
38
39     %get midpoint(root estimate)
40     c = ffalpos(a, b);
41
42     %eval. func at midpoint
43     fc = 4*c^4 - 32*c^3 + 97*c^2 - 132*c + 64;
44
45     %stop if f(c)=0
46     if fc == 0
47         break
48     end
49
50     %update count
51     count = count + 1;
52
53     %add to list of iterates
54     cits(count) = c;
55
56     %if sign change between a and c make c the new right endpt
57     if sign(fa)*sign(fc)<0
58
59         b = c;
60
61     %if sign chg betw c and b make c the new left endpt
62     else
63
64         a = c;
65
66     end
67
68 end
69
70 %update count
71     count = count + 1;
72
73 %get final midpoint(root estimate)
74     c = ffalpos(a,b);
75
76 %add to vector of iterates
77     cits(count) = c;

```

```

78
79 %display error estimate
80 error = abs(b-a)/2
81
82 %display vector of iterates
83 cits
84
85 %display number of iterates
86 count

>> falpos.m

```

```

cits =
1.118146029604788e+00
1.062353732855704e+00
1.060497138511485e+00
1.060437095603502e+00
1.060435155605902e+00
1.060435092926107e+00
1.060435090900974e+00
1.060435090835543e+00
1.060435090833429e+00
1.060435090833361e+00
1.060435090833359e+00

```

```

count =
11

```

Problem 2: newt1.m

```
1 format long e
2
3 % Tolerance Level
4 TOL = 10−6;
5
6 % Initial Guess
7 xp = 2;
8
9 iterates = [];
10
11 % While the relative error is less than the give tolerance, we
12 will
13 % continue to use newton's method.
14 while 1
15     xf = xp − fnewt(xp) / fpnewt(xp);
16     iterates = [iterates , xf];
17
18     if(abs((xf−xp)/(xf)) < TOL)
19         break
20     end
21
22     xp = xf;
23
24 end
25
26 % Ouprinting the resulting iterates
27 clc
28 fprintf("Iterates::")
29 iterates '
```

Problem 2: newt2.m

```
1 format long e
2
3 clc
4
5 % Tolerance Level
6 TOL = 10−6;
7
8 % Initial Guess
9 xp = 2;
```

```

10
11 % Arrays to store the iterates and error values required
12 iterates = [];
13 relerr = [];
14 relerr2 = [];
15
16 % ep is 0 for the first iteration
17 ep = 1;
18
19 % While the relative error is less than the give t olerance ,
   we will
20 % continue to use newton's method.
21 while 1
22
23     % Newton's Algorithm
24     xf = xp - fnewt(xp) / fpnewt(xp);
25     iterates = [iterates , xf];
26
27     % Error and Tolerance
28     ef = abs((xf-xp)/(xf));
29     fprintf(" %f\n", ef)
30     if(ef < TOL)
31         break
32     end
33
34     relerr = [relerr , ef/ep];
35     relerr2 = [relerr2 , ef/ep^2];
36
37     xp = xf;
38     ep = ef;
39
40 end
41
42 % Ouprinting the resulting iterates
43 clc
44 fprintf(" Iterates ::\n")
45 iterates '
46
47 fprintf(" Error 1::\n")
48 relerr '
49
50 fprintf(" Error 2::\n")
51 relerr2 '
52
53 fprintf(" Since the derivative of the function is non-zero , we

```

```

    can determine that there are multiple roots.\n")
54
55 v = round(iterates(length(iterates)), 3);
56 mult = (147500 * (1 + (v/12))^58)/2*fpnewt(v);
57
58 rofc = (mult-1)/mult;
59
60 fprintf("Iterates:: %d\n", v)
61 fprintf("Multiplicity:: %d\n", mult)
62 fprintf("Rate of Convergence:: %d\n", rofc)
63
64 fprintf("We have determined that we have a rate of convergence
    of approx: 1, therefore we have linear convergence for
    this newton's method.\n")

>> newt2.m

```

- a)
- 1.a) Actual. value of pi from MatLab = 3.141593e+00
 - 1.b) Approx. value of pi using Maclaurin series: $4\arctan(1) = 3.141597\text{e}+00$
 - 2) Absolute Error: 4.140539e-06
 - 3) Relative Error: 1.317974e-06
 - 4) Number of 'k' terms needed to approx. in single percision: 16777216

b)

Eventually the next value in the series becomes extremely small (since k is constantly increasing in the denominator, the next value to be added in the series will be small). This is the result of the phenomenon SWAMPING, since we are trying to add two numbers whose sizes are very different (one large and one extremely small). Therefore, the percision will eventually lose track of the very small values computed due to the rounding and computational limitations.

```

1  format long e
2
3  clc
4
5  x1 = 5;
6  x2 = 5 + 10(-10);
7
8  fun = @(x)exp(x-1) - 1;
9  actual = 1;
10
11 fprintf("fzero(function1, x) results::\n")
12 [rootest, fval] = fzero(fun, x1);
13 backerr = abs(fun(rootest));
14 forerr = abs(actual - rootest);
15 fprintf("(x, rootest, fval): %d, %d, %d\n", x1, rootest, fval)
16 fprintf("(Backward error, Forward error): %d, %d\n", backerr,
    forerr)
17
18 [rootest, fval] = fzero(fun, x2);
19 backerr = abs(fun(rootest));
20 forerr = abs(actual - rootest);
21 fprintf("(x, rootest, fval): %d, %d, %d\n", x2, rootest, fval)
22 fprintf("(Backward error, Forward error): %d, %d\n\n", backerr
    , forerr)
23
24 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
25
26 fun = @(x)exp(4*x-4) - 2*exp(3*x-3) + 2*exp(x-1) - 1;
27
28 fprintf("fzero(function2, x) results::\n")
29 [rootest, fval] = fzero(fun, x1);
30 backerr = abs(fun(rootest));
31 forerr = abs(actual - rootest);
32 fprintf("(x, rootest, fval): %d, %d, %d\n", x1, rootest, fval)
33 fprintf("(Backward error, Forward error): %d, %d\n", backerr,
    forerr)
34
35 [rootest, fval] = fzero(fun, x2);
36 backerr = abs(fun(rootest));
37 forerr = abs(actual - rootest);
38 fprintf("(x, rootest, fval): %d, %d, %d\n", x2, rootest, fval)
39 fprintf("(Backward error, Forward error): %d, %d\n\n", backerr

```



```

    , forerr)
40
41 syms f(x)
42 f(x) = exp(x-1) - 1;
43
44 mult1 = 0;
45 while f(actual) == 0
46     f = diff(f, x);
47     mult1 = mult1 + 1;
48 end
49
50 f(x) = exp(4*x-4) - 2*exp(3*x-3) + 2*exp(x-1) - 1;
51
52 mult2 = 0;
53 while f(actual) == 0
54     f = diff(f, x);
55     mult2 = mult2 + 1;
56 end
57 fprintf("Multiplicity f1: %d\n", mult1)
58 fprintf("Multiplicity f2: %d\n", mult2)
59 fprintf("As the multiplicity increases, stability decreases.\n
    Thus, with multiplicity 1 for the first function, we\n
    have a pretty stable algorithm. However, when we\n
    introduce higher multiplicity in function 2, we start to\n
    lose some of our stability and it begins to slightly\n
    affect\n our approximations. The initial guess doesn't\n
    affect the first\n function very much, but you can see the\n
    increased variability in\n the second function from the\n
    root difference.\n")

```

```
>> hw3.m
```

```
fzero(function1, x) results:: (x, rootest, fval): 5, 1, 0  
(Backward error, Forward error): 0, 0  
(x, rootest, fval): 5.000000e+00, 1, 0  
(Backward error, Forward error): 0, 0
```

```
fzero(function2, x) results::  
(x, rootest, fval): 5, 9.999996e-01, 0  
(Backward error, Forward error): 0, 4.312810e-07  
(x, rootest, fval): 5.000000e+00, 1.000003e+00, 0  
(Backward error, Forward error): 0, 3.095297e-06
```

Multiplicity f1: 1

Multiplicity f2: 3

As the multiplicity increases, stability decreases. Thus, with multiplicity 1 for the first function, we have a pretty stable algorithm. However, when we introduce higher multiplicity in function 2, we start to lose some of our stability and it begins to slightly affect our approximations. The initial guess doesn't affect the first function very much, but you can see the increased variability in the second function from the root difference.