Queens College, CUNY, Department of Computer Science Numerical Methods CSCI 361 / 761 Fall 2018

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2 Homework set 2

- Please email your solution, as a file attachment, to Sateesh.Mane@qc.cuny.edu.
- Please submit one zip archive with all your files in it.
 - 1. The zip archive should have either of the names (CS361 or CS761):

StudentId_first_last_CS361_hw2.zip StudentId_first_last_CS761_hw2.zip

- 2. The archive should contain one "text file" named "hw2.[txt/docx/pdf]" and one cpp file per question named "Q1.cpp" and "Q2.cpp" etc.
- 3. Note that not all homework assignments may require a text file.
- 4. Note that not all questions may require a cpp file.

2.1 Functions

- We shall write code to implement the bisection and Newton–Raphson root finding algorithms.
- For bisection, we require only the function value.
- For Newton–Raphson, we also require the derivative.
- Write the following forward declarations, which we shall use below.

```
double func(double x);
void func(double x, double &f, double &fprime);
```

2.2 Bisection

• The function signature for our bisection function is

- The return type is "int" not void, because the algorithm might not converge. If the calculation succeeds, we return 0. It it fails, we return 1.
- The inputs are (i) double target, (ii) double tol_f, (iii) double tol_x, (iv) int max_iter, (v) double x_low, (vi) double x_high.
- The outputs are (vii) double & x (the root), (viii) int & num_iter (number of iterations).
- Obviously target is the target value. We wish to solve f(x) = target.
- We also need tolerance parameters. Remember that an iterative algorithm needs a cutoff parameter to stop iterating. We use tol_f for convergence along the y-axis and tol_x for convergence along the x-axis.
- As a safety check, we also input an upper limit max_iter on the number of iterations, in case the computations take too long.

The following pseudocode describes the steps. You can use it as the basis to write a working function.

- 1. Initialize x = 0 and $num_iter = 0$.
- 2. Calculate a value double y_low = func(x_low).
 Also calculate double diff_y_low = y_low target.
- 3. If |diff_y_low| ≤ tol_f, then we are done.

 The value of y_low is already within the tolerance.

 Set x = x_low and "return 0" (= success) and exit.
- 4. Next calculate a value double y_high = func(x_high).
 Also calculate double diff_y_high = y_high target.
- 5. If |diff_y_high| ≤ tol_f, then we are done.

 The value of y_high is already within the tolerance.

 Set x = x_high and "return 0" (= success) and exit.
- 6. Next we must check if we have bracketed a root.

 In order to bracket a root, y_low and y_high must lie on opposite sides of target.

 This means diff_y_low and diff_y_high must have opposite signs.
- 7. Test if (diff_y_low * diff_y_high > 0.0).
 If yes, then we have failed. The inputs x_low and x_high do not bracket a root.
 Set x = 0 and "return 1" (= fail) and exit.

- 8. If we have made it this far, then we know that we have bracketed a root. We know that there is a solution for x somewhere between x_low and x_high.
- 9. Hence we begin the bisection (iteration) loop.

```
for (num_iter = 1; num_iter < max_iter; ++num_iter) {
    // (to be filled in below)
}</pre>
```

- 10. In the loop, set $x = (x_low + x_high)/2.0$ and calculate double y = func(x).
- 11. Also calculate double diff_y = y target.
- 12. If $|diff_y| \le tol_f$, then we are done. The value of y is within the tolerance. We have found a "good enough" value for x. Hence "return 0" (= success) and exit.
- 13. Next test if (diff_y * diff_y_low > 0.0).
 If yes, it means y and y_low are on the same side of target.
 This means x and x_low are on the same side of the root.
 Update x_low = x.
- 14. Else obviously y and y_high are on the same side of target.
 Hence x and x_high are on the same side of the root.
 Update x_high = x.
- 15. Don't be in rush to iterate! There is one more test!
- 16. We have tested for convergence in y, now we must test for convergence in x.
- 17. If $|x_high x_low| \le tol_x$, then the algorithm has converged (up to the tolerance). Note that the tolerance is tol_x in this test (convergence in x). Hence "return 0" (= success) and exit.
- 18. If we have come this far, continue with the iteration loop.
- 19. If we exit the iteration loop after max_iter steps and the calculation still has not converged, then set x = 0 and num_iter = max_iter and "return 1" (= fail) and exit.
- 20. We have reached the end of the function. By now either we have a "good enough" answer ("return 0" = success) or not ("return 1" = fail).

2.3 Testing of function

- You will have to write a main program to call and test your function.
- However, you also need an actual example of a function f(x).
- One simple example is a parabola.

```
double func(double x)
{
  return x*x;
}
```

• For later use with Newton–Raphson, also write the following function.

```
void func(double x, double &f, double &fprime)
{
   f = x*x;
   fprime = 2.0*x;
}
```

- The equation is therefore f(x) = target. The solution is $x = \pm \sqrt{\text{target}}$.
- Try target = 4.0. If your function works correctly, it should output x = 2.0 or -2.0.
- Set tol_f = tol_x = 1.0e-6 and max_iter=100.
- If x_low=0.0 and x_high=5.0, the algorithm should converge to x = 2.0.
- If $x_low=-5.0$ and $x_high=5.0$, the algorithm should return 1 and fail.
- You can devise other tests. Use your imagination.
- Try x_low=-10.0 and x_high=0.0. Try target =-1.0.

2.4 Newton-Raphson

- Next let us write a function to implement the Newton–Raphson algorithm. This function is in fact shorter than the one for bisection.
- The function signature for our Newton–Raphson function is

- The return type is "int" not void, because the algorithm might not converge. If the calculation succeeds, we return 0. It it fails, we return 1.
- The inputs are (i) double target, (ii) double tol_f, (iii) double tol_x, (iv) int max_iter, (v) double x0 (initial iterate).
- The outputs are (vi) double & x (the root), (vii) int & num_iter (number of iterations).
- Obviously target is the target value. We wish to solve f(x) = target.
- We also need tolerance parameters.

 Remember that an iterative algorithm needs a cutoff parameter to stop iterating.

 We use tol_f for convergence along the y-axis and tol_x for convergence along the x-axis.
- As a safety check, we also input an upper limit max_iter on the number of iterations, in case the computations take too long.

The following pseudocode describes the steps. You can use it as the basis to write a working function.

1. Declare some useful variables at the start of the function.

```
const double tol_fprime = 1.0e-12;
double f = 0;
double fprime = 0;
```

- 2. The const parameter "tol_fprime" is to guard against division by zero. The value of 1.0e-12 is arbitrary.
- 3. We do not need any preliminary tests.

```
Set x = x0 (= initial iterate) and begin the main iteration loop.
```

```
x = x0;
for (num_iter = 1; num_iter < max_iter; ++num_iter) {
    // (to be filled in below)
}</pre>
```

4. We could begin with num_iter = 0 but this is not important.

- 5. In the loop, call func(x,f,fprime).
- 6. Also calculate double diff_f = f target.
- 7. If |diff_f| ≤ tol_f, then we are done.
 The value of f is within the tolerance.
 We have found a "good enough" value for x.
 Hence "return 0" (= success) and exit.
- 8. Next test if |f_prime| ≤ tol_fprime.
 If yes, then this will cause a division by zero and the Newton-Raphson iteration fails.
 Set x = 0 and "return 1" (= fail) and exit.
- 9. Calculate double delta_x = diff_f/fprime.
- 10. Now test for convergence in x.
- 11. If $|\text{delta_x}| \leq \text{tol_x}$, then the algorithm has converged (up to the tolerance). Note that the **tolerance is tol_x** in this test (convergence in x). Hence "return 0" (= success) and exit.
- 12. If the convergence tests have not passed, update the value of x (note the minus sign).

```
x -= delta_x;
```

- 13. Continue with the iteration loop.
- 14. If we exit the iteration loop after max_iter steps and the calculation still has not converged, then set x = 0 and num_iter = max_iter and "return 1" (= fail) and exit.
- 15. We have reached the end of the function. By now either we have a "good enough" answer ("return 0" = success) or not ("return 1" = fail).
- 16. Probably the best test is to use the same main program that you wrote to call the bisection algorithm and use it to also call the Newton-Raphson function. Give the same target and tolerance parameters to both functions. For bisection you have to input x_low and x_high whereas for Newton-Raphson you have to input a starting oterate x0. If they both converge, the output value of x should be the same in both cases (up to the tolerance). Also print out the number of iterations num_iter from both function calls. In general, Newton-Raphson should converge (much) more quickly.

2.5 Cumulative Normal Distribution

- A more interesting function is the **cumulative normal distribution**.
- The probability density function for the normal probability distribution is (with mean = 0 and variance = 1)

$$p(x) = \frac{e^{-x^2/2}}{\sqrt{2\pi}} \qquad (-\infty < x < \infty).$$
 (2.5.1)

• The **cumulative normal distribution** is defined via the integral

$$N(x) = \int_{-\infty}^{x} \frac{e^{-t^2/2}}{\sqrt{2\pi}} dt \qquad (-\infty < x < \infty).$$
 (2.5.2)

- The difficulty is, how do we compute N(x) for arbitrary values of x?

 The above formula is an integral, which by itself is complicated to compute.
- Fortunately C++ (also Excel/Visual Basic) and other languages have a function $\operatorname{erf}(x)$. It is known as the "error function" (hence "erf") as is defined as

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt.$$
 (2.5.3)

• Hence we obtain N(x) via

$$N(x) = \frac{1 + \text{erf}(x/\sqrt{2})}{2}.$$
 (2.5.4)

• The following functions will serve our needs.

```
double cum_norm(double x)
{
   const double root = sqrt(0.5);
   return 0.5*(1.0 + erf(x*root));
}
double func(double x)
{
   return cum_norm(x);
}
void func(double x, double &f, double &fprime)
{
   const double pi = 4.0*atan2(1.0,1.0);
   f = cum_norm(x);
   fprime = exp(-0.5*x*x)/sqrt(2.0*pi);
}
```

• Use this and run some tests with both bisection and Newton-Raphson. Unlike a quadratic, it is not simple to solve the equation N(x) = target. By construction, $N(-\infty) = 0$ and $N(\infty) = 1$ and for finite values of x, then 0 < N(x) < 1. Hence to obtain convergence, the target value must lie in the range 0 < target < 1.

2.6 Multiple test functions

- Write only one bisection function and one Newton-Raphson function.
- Do not write root_bisection_quadratic() and root_bisection_cum_norm(), etc.
- Also do not write a new cpp file every time you write a new test function.
- Although this will work for a homework assignment, in a real software library function we cannot write a new function or new file for each new test function. What will happen if the client wants to use $f(x) = x^3$ and solve $x^3 = \text{target}$? Or some other function?
- A real software library root-finding function would take an extra input argument for a function pointer. I decided this is too complicated and too much programming burden on you. A lot of programming formalism, which has nothing to do with the subject matter of this course.
- What to do?
- Use #ifdef preprocessor directives. Here is a sample code of what I mean. It is not pretty but it works. I added a third selection "cubic" for $f(x) = x^3$ to illustrate the basic idea. Uncomment one of the #define statements to select a test function.
- Then root_bisection() etc. calls only func() and not a different function for each test case.
- It is true that you must recompile the program every time you change the choice of test function. It is not a pretty solution, but it works.

2.6.1 Preprocessor code

uncomment one of the #define choices below to select a test function

```
//#define func_quadratic
//#define func_cubic
#define func_cum_norm
#ifdef func_quadratic
double func(double x){
    return x*x;
};
void func(double x, double &f, double &fprime){
    fprime=2.0*x;
};
#endif // func_quadratic
#ifdef func_cubic
double func(double x){
    return x*x*x;
};
void func(double x, double &f, double &fprime){
    f=x*x*x;
    fprime=3.0*x*x;
};
#endif // func_cubic
#ifdef func_cum_norm
double cum_norm(double x)
    const double root = sqrt(0.5);
    return 0.5*(1.0 + erf(x*root));
double func(double x)
    return cum_norm(x);
void func(double x, double &f, double &fprime)
{
    const double pi = 4.0*atan2(1.0,1.0);
    f = cum_norm(x);
    fprime = \exp(-0.5*x*x)/\operatorname{sqrt}(2.0*pi);
#endif // func_cum_norm
```

2.6.2 Main program pseudocode

```
// header files etc
int main()
{
    // set inputs
    // call root_bisection(...)
    // call root_NR(...)
    // check if the results from bisection and NR match (up to tolerance)

    // CALL THE TEST FUNCTION USING X = ROOT (bisec and NR roots)
    // IS THE VALUE F(X_ROOT) CLOSE TO ZERO?

    // in other words, did the algorithm converge correctly?

    return 0;
}
```