



Istanbul Technical University
Department of Computer Engineering

BLG 202E – Numerical Methods Assignment 1 Solutions

Solution 1.a

$\sin \alpha - \sin \beta = 2 \sin(\alpha - \beta) \cos(\alpha + \beta)$ is given as trigonometric equation and

$f_1(x_0, h) = \sin(x_0 + h) - \sin(x_0)$ is given as a function.

In the same time,

$$f_2(x_0, h) = \sin(x_0 + h) - \sin(x_0) = 2 \cos\left(\frac{x_0+h+x_0}{2}\right) \sin\left(\frac{x_0+h-x_0}{2}\right) \text{ then}$$

$$f_2(x_0, h) = 2 \cos\left(\frac{2x_0+h}{2}\right) \sin\left(\frac{h}{2}\right) \text{ can be transformed like this.}$$

Solution 1.b

Derivative of $f(x) = \sin(x)$ at $x = x_0$:

$$f'(x) = \frac{f(x_0+h)-f(x_0)}{h} = \frac{\sin(x_0+h)-\sin(x_0)}{h}$$

However, in this case there can be cancellation errors for computing the approximation $\frac{f(x_0+h)-f(x_0)}{h}$ to the derivative of $f(x) = \sin(x)$ at $x = x_0$. In some value of h , $\frac{\sin(x_0+h)-\sin(x_0)}{h}$ will be $\frac{\sin(x_0)-\sin(x_0)}{h}$ so that $\frac{0}{h}$. For avoiding from this situation, $\frac{\sin(x_0+h)-\sin(x_0)}{h}$ can be written as $\frac{2 \cos\left(\frac{x_0+h+x_0}{2}\right) \sin\left(\frac{x_0+h-x_0}{2}\right)}{h} = \frac{2 \cos\left(\frac{2x_0+h}{2}\right) \sin\left(\frac{h}{2}\right)}{h}$ using trigonometric equations.

In Matlab:

```

x0 = 1.2; %%Initial Value
i = -20:1:0;
h = 10.^i; %%Desired value of h form 1e-20 to 1

syms x
derivative = inline(diff(sin(x), 'x')); %%derivative of sin(x) without
approximation

f0 = cos((2*x0 + h) / 2);
f1 = sin(h/2);
approximationDerivativeWithoutCancellationError = (2.*f0.*f1)./h;
err1 = abs(approximationDerivativeWithoutCancellationError -
derivative(x0));

approximationDerivativeWithCancellationError = (sin(x0 + h) - sin(x0)) ./
h;
err2 = abs(approximationDerivativeWithCancellationError - derivative(x0));

fprintf('h : %d\n\n', h')
fprintf('Without cancellation error: %f\n\n',
approximationDerivativeWithoutCancellationError')
fprintf('With cancellation error: %f\n\n',
approximationDerivativeWithCancellationError')

```

Note: Some required explanation about variables are given following:

derivative = derivative of $\sin(x)$ without approximation

approximationDerivativeWithoutCancellationError = derivative of $\sin(x)$ with respect to $h = 1e-20; 1e-19; \dots 1$ and $x_0 = 1.2$. In order to get rid of cancellation error, definition of derivative transformed to another function using trigonometric equations.

approximationDerivativeWithCancellationError = derivative of $\sin(x)$ with respect to $h = 1e-20; 1e-19; \dots 1$ and $x_0 = 1.2$. Definition of derivative is used.

With this Matlab script, we can see that $err1$ is zero for the first six values of h ; on the other hand, $err2$ is never zero because of the cancellation error.

h : 1.000000e-20	Without cancellation error: 0.362358
h : 1.000000e-19	Without cancellation error: 0.362358
h : 1.000000e-18	Without cancellation error: 0.362358
h : 1.000000e-17	Without cancellation error: 0.362358
h : 1.000000e-16	Without cancellation error: 0.362358
h : 1.000000e-15	Without cancellation error: 0.362358
h : 1.000000e-14	Without cancellation error: 0.362358
h : 1.000000e-13	Without cancellation error: 0.362358
h : 1.000000e-12	Without cancellation error: 0.362358
h : 1.000000e-11	Without cancellation error: 0.362358
h : 1.000000e-10	Without cancellation error: 0.362358
h : 1.000000e-09	Without cancellation error: 0.362358
h : 1.000000e-08	Without cancellation error: 0.362358
h : 1.000000e-07	Without cancellation error: 0.362358
h : 1.000000e-06	Without cancellation error: 0.362357
h : 1.000000e-05	Without cancellation error: 0.362353
h : 1.000000e-04	Without cancellation error: 0.362311
h : 1.000000e-03	Without cancellation error: 0.361892
h : 1.000000e-02	Without cancellation error: 0.357692
h : 1.000000e-01	Without cancellation error: 0.315191
h : 1	Without cancellation error: -0.123543

Solution 2

Horner's Method says that polynomial of degree n requires n multiplications and n additions which be totally $2n$ elementary operations. For example;

$f(x) = 4x^2 + 5x + 9$ can be written as following

$f(x) = x(4x + 5) + 9$ as nested. In this case function can be calculated 4 elementary operations (2 multiplications and 2 additions). The operation will iterate n times to reach polynomial of degree n . Hence complexity will be $O(n)$.

Solution 3

The finite floating numbers can be represented using floating point system (i.e., some finite integer base β and precision t). However, some values continue indefinitely like $\frac{1}{3}$, π , e . These values would not represent precisely. Briefly, no.

Solution 4.a

In the fixed point iteration there is $f(x) = 0$ can be rewritten as $x = g(x)$.

$g(x) = x^2 + \frac{3}{16}$ has two fixed points:

$$g(x) = x = x^2 + \frac{3}{16} \quad \rightarrow \quad x_1 = 0.25, \quad x_2 = 0.75$$

Solution 4.b

Iterations can converge to fixed point if only if $|g'(x^*)| < \rho$, $\rho < 1$. In this case,

$|g'(0.25)| = 0.5$ and $|g'(0.75)| = 1.5$. Converging fixed point will be $x_1 = 0.25$.

For example, 0.55 can be chosen as initial guess as x_0 .

$$x_0 = 0.55, \quad x_{n+1} = g(x_n)$$

$$x_1 = 0.49,$$

$$x_2 = 0.4276$$

.

.

.

$$x_{17} = 0.25001250092667$$

$$x_{18} = 0.25000625061961$$

0.55 is converge to the **0.25** which be fixed point of $|g'(x^*)| < \rho$, $\rho < 1$, not **0.75**

Solution 4.c

$|x_{k+1} - x^*| = |g(x_k) - g(x^*)| \leq \rho |x_k - x^*|$ This is a **contraction** by the factor of ρ . So

$|x_{k+1} - x^*| \leq \rho |x_k - x^*| \leq \rho^2 |x_k - x^*| \leq \dots \leq \rho^{k+1} |x_0 - x^*|$. Since $\rho < 1$, we have $\rho^{k+1} \rightarrow 0$ as $k \rightarrow \infty$.

$$\rho |x_k - x^*| \leq \rho^{k+1} |x_0 - x^*| \quad \rightarrow \quad |x_k - x^*| \leq \rho^k |x_0 - x^*|$$

By a factor of 10, we get $|x_k - x^*| \approx 0.1 |x_0 - x^*|$. So,

$\rho^k \approx 0.1$ Taking \log_{10} of both sides yields $k \log_{10} \rho \approx -1$. Rate of converges can be defined as $rate = -\log_{10} \rho$. k was the iteration numbers so $k = \lceil 1/rate \rceil$.

$$k = \lceil 1 / -\log_{10} \rho \rceil$$

$$\lim_{n \rightarrow \infty} \left(\frac{x_{n+1} - x^*}{x_n - x^*} \right) = |g'(x^*)| = \rho \quad \rightarrow \quad |g'(x^*)| = g'(0.25) = 0.5$$

$$\rho = 0.5 \quad \rightarrow \quad k = \lceil 1 / -\log_{10} 0.5 \rceil$$

$rate = -\log_{10} 0.5 \approx 0.301$. This yields $k = 4$.

Briefly, 4 iterations will be required to reduce the convergence error by a factor of 10.

Solution 5.a

In the Newton-Raphson Method, the theorem of the equation can be written following:

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$$

$$x_{k+1} = x_k - \frac{40x_k^{1.5} - 875x_k + 35000}{60x_k^{0.5} - 875}$$

Let take the initial guess $x_0 = 50$

Iteration I

$$x_1 = 50 - \frac{40 * 50^{1.5} - 875 * 50 + 35000}{60 * 50^{0.5} - 875}$$

$$x_1 = 61.962959350326933$$

Absolute relative approximate error:

$$\frac{61.962959350326933 - 50}{61.962959350326933} * 100 = 19.306630083128550 \%$$

Iteration II

$$x_2 = 61.962959350326933 - \frac{40*61.962959350326933^{1.5} - 875*61.962959350326933 + 35000}{60*61.962959350326933^{0.5} - 875}$$

$$x_2 = 62.689189096270965$$

Absolute relative approximate error:

$$\frac{62.689189096270965 - 61.962959350326933}{62.689189096270965} * 100 = 1.158460902770285 \%$$

Still absolute relative approximate error greater than 1%.

Iteration III

$$x_3 = 62.689189096270965 - \frac{40*62.689189096270965^{1.5} - 875*62.689189096270965 + 35000}{60*62.689189096270965^{0.5} - 875}$$

$$x_3 = 62.691697121424490$$

Absolute relative approximate error:

$$\frac{62.691697121424490 - 62.689189096270965}{62.691697121424490} * 100 = 0.004000569881952 \%$$

Absolute relative approximate error is less than 1%.

Solution 5.b

```
syms x
x0 = 50; %%Initial x
xn = 0; %%Next value of x
i = 0; %%Iteration step
err = 1; %% Absolute relative approximate error
fprintf('Initial x = %.20f\n', x0)
while err > 10^-2
    f = inline((40*x^1.5)-875*x+35000, 'x'); %%f(x)
    derivative = inline(diff((40*x^1.5)-875*x+35000, 'x')); %%f'(x)
    xn = x0 - f(x0)/derivative(x0); %%Newton-Raphson Method Iteration
    err = (xn-x0)/xn*100;
    x0 = xn;
    i = i + 1;
    fprintf('Iteration %d: x = %.20f, err = %.20f\n', i, xn, err)
end
```

$f_x \gg$