

APPM 5600 - Homework 2

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1. Which of the following iterations will converge to the indicated fixed point x_* (provided x_0 is sufficiently close to x_*)? If it does converge, give the order of convergence; for linear convergence, give the rate of linear convergence.

(a) $x_{n+1} = -16 + 6x_n + \frac{12}{x_n}$, $x_* = 2$.

Let g be defined by $g(x) = -16 + 6x + \frac{12}{x}$. The first derivative of g is given by

$$g'(x) = 6 + \frac{12}{x^2}.$$

We can perform a first-order Taylor expansion of g centered at 2 as follows:

$$\begin{aligned} g(x) &= g(2) + g'(\xi_x)(x - 2) \\ &= 2 + \left(6 + \frac{12}{\xi_x^2}\right)(x - 2), \end{aligned}$$

where ξ_x is a number between x and 2. Evaluating the above Taylor expansion at x_n , we see that

$$x_{n+1} = 2 + \left(6 + \frac{12}{\xi_x^2}\right)(x_n - 2) \Rightarrow \left| \frac{x_{n+1} - 2}{x_n - 2} \right| = \left| 6 + \frac{12}{\xi_x^2} \right|.$$

On the interval $[1.8, 2.5]$, the quantity $\left| 6 + \frac{12}{\xi_x^2} \right|$ is minimized at $\xi_x = 2.5$, with the minimum being 7.92.

Since $\min_{x \in [1.8, 2.5]} \left| 6 + \frac{12}{\xi_x^2} \right|$ is greater than 1, the fixed point iteration cannot converge to x_* for any x in the interval $[1.8, 2.5]$, other than for the fixed point.

(b) $x_{n+1} = \frac{2}{3}x_n + \frac{1}{x_n^2}$, $x_* = 3^{1/3}$.

Our fixed point iteration is $x_{n+1} = g(x_n)$, where $g(x) = \frac{2}{3}x + \frac{1}{x^2}$. The derivative of g is given by

$$g'(x) = \frac{2}{3} - \frac{2}{x^3}.$$

Consider the interval $[a, b] = [1, 2]$. We see that $g(a) = \frac{2}{3} + 1 = \frac{5}{3}$, which is greater than a and less than b , and $g(b) = \frac{4}{3} + \frac{1}{4} = \frac{16}{12} + \frac{3}{12} = \frac{19}{12}$, which is greater than a and less than b . Furthermore, for every $x \in [a, b]$, $a \leq g(x) \leq b$, i.e., $g([a, b]) \subset [a, b]$. On the interval $[a, b]$, g' is a positive and decreasing function, and

$$\begin{aligned} \max_{x \in [a, b]} |g'(x)| &= |g'(b)| \\ &= \frac{2}{3} - \frac{1}{4} \\ &= \frac{5}{12} \\ &< 1. \end{aligned}$$

Thus, by the contraction mapping theorem, the iteration will converge to $x_* = 3^{1/3}$.

The order of convergence is quadratic, which can be seen by taking a second-order Taylor expansion of g centered at x_* . The second derivative of g is given by

$$g''(x) = -\frac{6}{x^4}.$$

Then, we have that

$$\begin{aligned} g'(x) &= g(x_*) + g'(x_*)(x - x_*) + \frac{g''(\xi_x)}{2}(x - x_*)^2 \\ &= x_* + \frac{g''(\xi_x)}{2}(x - x_*)^2, \end{aligned}$$

where ξ_x is a number between x and x_* . In addition, we have used the fact that $g'(x_*) = 0$. Evaluating this Taylor expansion at x_n , we have that

$$\begin{aligned} g(x_n) &= x_* + \frac{g''(\xi_x)}{2}(x_n - x_*)^2 \Rightarrow x_{n+1} = x_* + \frac{g''(\xi_x)}{2}(x_n - x_*)^2 \\ &\Rightarrow \frac{x_{n+1} - x_*}{(x_n - x_*)^2} = \frac{g''(\xi_x)}{2}, \end{aligned}$$

where ξ_x is a number between x_n and x_* . We can then take the limit of the absolute value of both sides, letting $n \rightarrow \infty$, and we obtain that

$$\begin{aligned} \lim_{n \rightarrow \infty} \left| \frac{x_{n+1} - x_*}{(x_n - x_*)^2} \right| &= \lim_{n \rightarrow \infty} \left| \frac{g''(\xi_x)}{2} \right| \\ &= \left| \frac{g''(x_*)}{2} \right| \\ &= \frac{3}{3^{4/3}} \\ &= 3^{-1/3}, \end{aligned}$$

where we used the fact that the x_n 's converge to x_* as $n \rightarrow \infty$.

(c) $x_{n+1} = \frac{12}{1+x_n}$, $x_* = 3$.

Our fixed point iteration is $x_{n+1} = g(x_n)$, where $g(x) = \frac{12}{1+x}$. The derivative of g is given by

$$g'(x) = -\frac{12}{(1+x)^2}.$$

Consider the interval $[a, b] = [2.5, 4]$. We see that $g(a) = \frac{24}{7}$, which is greater than a and less than b . We also see that $g(b) = 3$, which is greater than a and less than b . Furthermore, for every $x \in [a, b]$, $a \leq g(x) \leq b$, i.e., $g([a, b]) \subset [a, b]$. On the interval $[a, b]$, g' is a negative and increasing function, and we see that

$$\begin{aligned} \max_{x \in [a, b]} |g'(x)| &= |g'(a)| \\ &= \left| -\frac{12}{(1 + \frac{7}{2})^2} \right| \\ &= \frac{48}{49} \\ &< 1. \end{aligned}$$

Thus, by the contraction mapping theorem, the iteration will converge to $x_* = 3$.

The order of convergence is linear, which can be seen by taking a first-order Taylor expansion of g centered at x_* :

$$g(x) = g(x_*) + g'(\xi_x)(x - x_*),$$

where ξ_x is a number between x and x_* . Evaluating this Taylor expansion at x_n , we have that

$$\begin{aligned} g(x_n) = x_* + g'(\xi_x)(x_n - x_*) &\Rightarrow x_{n+1} = x_* + g'(\xi_x)(x_n - x_*) \\ &\Rightarrow \frac{x_{n+1} - x_*}{x_n - x_*} = g'(\xi_x), \end{aligned}$$

where ξ_x is a number between x_n and x_* . We can then take the limit of the absolute value of both sides, letting $n \rightarrow \infty$, and we obtain that

$$\begin{aligned} \lim_{n \rightarrow \infty} \left| \frac{x_{n+1} - x_*}{x_n - x_*} \right| &= \lim_{n \rightarrow \infty} |g'(\xi_x)| \\ &= |g'(x_*)| \\ &= \left| -\frac{12}{(1+3)^2} \right| \\ &= \left| \frac{12}{16} \right| \\ &= \frac{3}{4} \\ &< 1. \end{aligned}$$

Thus the rate of linear convergence for this iteration is $\frac{3}{4}$.

2. In laying water mains, utilities must be concerned with the possibility of freezing. Although soil and weather conditions are complicated, reasonable approximations can be made on the basis of the assumption that soil is uniform in all directions. In that case, the temperature in degrees Celsius $T(x, t)$, at a distance x (in meters) below the surface and t seconds after the beginning of a cold snap, approximately satisfies

$$\frac{T(x, t) - T_s}{T_i - T_s} = \operatorname{erf} \left(\frac{x}{2\sqrt{\alpha t}} \right),$$

where T_s is the constant temperature during a cold period, T_i is the initial soil temperature before the cold snap, α is the thermal conductivity (in meters² per second), and

$$\operatorname{erf}(t) = \frac{2}{\sqrt{\pi}} \int_0^t \exp(-s^2) ds.$$

Assume that $T_i = 20$ [degrees Celsius], $T_s = -15$ [degrees Celsius], and $\alpha = 0.138 \cdot 10^{-6}$ [meters² per second]. For parts (II) and (III), run your experiments with a tolerance of $\epsilon = 10^{-13}$.

- (a) We want to determine how deep a water main should be buried so that it will only freeze after 60 days exposure at this constant surface temperature. Formulate the problem as a root finding problem $f(x) = 0$. What is f and f' ? Plot the function f on $[0, \bar{x}]$, where \bar{x} is chosen so that $f(\bar{x}) > 0$.

Note: See attached code and plots.

Here, f is the function $T(x, t = 60 \text{ days}) = T(x)$, which is given by

$$f(x) = T(x) = (T_i - T_s) \left(\frac{2}{\sqrt{\pi}} \int_0^{x_f} \exp(-s^2) ds \right) + T_s,$$

where $x_f = \frac{x}{2\sqrt{\alpha t_f}}$ and $t_f = 5184000$ [seconds]. The derivative of f is given by

$$\begin{aligned} f'(x) &= T'(x) = (T_i - T_s) \left(\frac{2}{\sqrt{\pi}} \exp\left(-\frac{x^2}{4\alpha t_f}\right) \frac{1}{2\sqrt{\alpha t_f}} \right) \\ &= (T_i - T_s) \left(\frac{1}{\sqrt{\pi\alpha t_f}} \exp\left(-\frac{x^2}{4\alpha t_f}\right) \right). \end{aligned}$$

The goal is to then find a root of f via a root-finding technique.

- (b) Compute an approximate depth using the bisection method with starting values $a_0 = 0$ [meters] and $b_0 = \bar{x}$ [meters].

We take \bar{x} to be 1 because $f(\bar{x}) > 0$. Applying the bisection method with the above parameters, we find an approximate depth of 0.6769618544819309 meters.

- (c) Compute an approximate depth using Newton's method with starting value $x_0 = 0.01$ [meters]. What happens if you start with $x_0 = \bar{x}$? Which is your preferred method and why?

Applying Newton's method with the above parameters and $x_0 = 0.01$ [meters], we find an approximate depth of 0.6769618544819355 meters in 14 iterations. If Newton's method is applied with the same parameters and $x_0 = \bar{x}$ [meters], the method returns an approximate depth of 0.6769618544819372 meters in 14 iterations.

I prefer Newton's method over the bisection method because Newton's method converges to an approximate root of T in far fewer iterations, especially if the initial iterate is close enough to the root of T .

3. Consider applying Newton's method to a real cubic polynomial.

- (a) In the case that the polynomial has three distinct real roots, $x = \alpha$, $x = \beta$, and $x = \gamma$, show that the starting guess $x_0 = \frac{1}{2}(\alpha + \beta)$ will yield the root γ in one step.

Suppose f is a real cubic polynomial whose roots are distinct and are given by α , β , and γ . We can write f as

$$\begin{aligned} f(x) &= (x - \alpha)(x - \beta)(x - \gamma) \\ &= (x^2 - \beta x - \alpha x + \alpha\beta)(x - \gamma) \\ &= (x^2 - (\alpha + \beta)x + \alpha\beta)(x - \gamma) \\ &= x^3 - \gamma x^2 - (\alpha + \beta)x^2 + (\alpha + \beta)\gamma x + \alpha\beta x - \alpha\beta\gamma \\ &= x^3 - (\alpha + \beta + \gamma)x^2 + (\gamma(\alpha + \beta) + \alpha\beta)x - \alpha\beta\gamma \end{aligned}$$

and

$$f'(x) = 3x^2 - 2(\alpha + \beta + \gamma)x + (\gamma(\alpha + \beta) + \alpha\beta).$$

The first iteration of Newton's method is given by

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)},$$

which is given by

$$\begin{aligned}
x_1 &= \frac{1}{2}(\alpha + \beta) - \frac{\left(\frac{1}{2}(\alpha + \beta) - \alpha\right)\left(\frac{1}{2}(\alpha + \beta) - \beta\right)\left(\frac{1}{2}(\alpha + \beta) - \gamma\right)}{3\left(\frac{1}{2}(\alpha + \beta)\right)^2 - 2(\alpha + \beta + \gamma)\left(\frac{1}{2}(\alpha + \beta)\right) + (\gamma(\alpha + \beta) + \alpha\beta)} \\
&= \frac{1}{2}(\alpha + \beta) + \frac{\frac{1}{4}(\beta - \alpha)^2\left(\frac{1}{2}(\alpha + \beta) - \gamma\right)}{\frac{-\alpha^2 - 2\alpha\beta - \beta^2 + 4\alpha\beta}{4}} \\
&= \frac{1}{2}(\alpha + \beta) + \frac{(\beta - \alpha)^2\left(\frac{1}{2}(\alpha + \beta) - \gamma\right)}{-\alpha^2 + 2\alpha\beta - \beta^2} \\
&= \frac{1}{2}(\alpha + \beta) + \frac{(\beta - \alpha)^2\left(\frac{1}{2}(\alpha + \beta) - \gamma\right)}{-(\beta - \alpha)^2} \\
&= \frac{1}{2}(\alpha + \beta) - \frac{1}{2}(\alpha + \beta) + \gamma \\
&= \gamma,
\end{aligned}$$

as desired.

- (b) Give a heuristic, e.g., geometric, argument showing that if two roots coincide (say $\beta = \gamma$), there is precisely one starting guess x_0 (other than the double root) for which Newton will fail, and that this one separates the basins of attraction for the distinct roots.

Recall that Newton's method is a fixed point iteration, i.e., we are trying to find an α such that $\alpha = g(\alpha)$, where $g(x) = x - \frac{f(x)}{f'(x)}$ and α is a root of f . The interval around α for which g is a contraction is called the basin of convergence for α .

A cubic that has three real roots, with two roots having multiplicity two, has the property that its first derivative is zero at the double root.

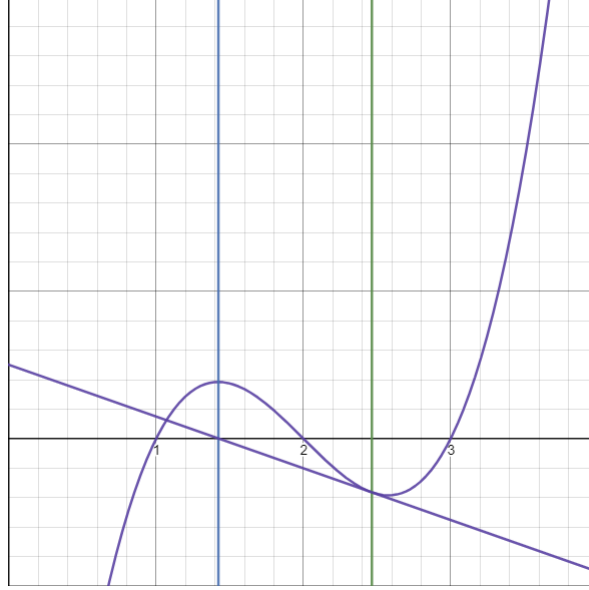
The derivative of f , assuming that $\alpha = \beta$, is given by

$$f'(x) = 3x^2 - 2(2\alpha + \gamma)x + (2\alpha\gamma + \alpha^2).$$

The first derivative f' has two roots, one at the double root $\alpha = \beta$ and another point x^0 . Newton's method will fail if the initial iterate is x^0 as $\frac{f(x^0)}{f'(x^0)}$ is undefined. Furthermore, this is the only initial starting point for which Newton's method will fail in this case. This is because it is not possible to pick another initial iterate such that the tangent line of f at this iterate crosses the x -axis at x^0 , which would cause the Newton iteration to fail.

- (c) Extend the argument in part (II) to the case when all three roots again are distinct. Explain why there are now infinitely many starting guesses x_0 for which the iteration will fail.

Suppose that f is a cubic polynomial with three real and distinct roots. As before, f' has two roots, but now, its roots do not coincide with the roots of f . Newton's method fails at the roots of f' , but it is possible to arrive at these roots even if the initial Newton iterate is not one of the roots of f' . Consider the cubic $f(x) = (x - 1)(x - 2)(x - 3)$:



One of the roots of f' is located approximately at $x = 1.42264973081$, indicated by the vertical blue line. If the initial Newton iterate is approximately 2.465, i.e., $x^0 \approx 2.465$, the tangent line of f at x^0 will pass through $x = 1.42264973081$, at which the next Newton iteration fails. However, it is possible to find another initial iterate that produces 2.465 (indicated by the vertical green line) as the next iterate, which leads to the eventual failure to Newton's method. This process can be repeated ad infinitum, which implies that there are infinitely many initial iterates that will cause Newton's method to eventually fail when trying to find the roots of the above cubic, and in fact, any cubic with real, distinct roots.

4. The sequence x_k produced by Newton's method is quadratically convergent to x_* with $f(x_*) = 0$, $f'(x_*) \neq 0$, and $f''(x)$ continuous at x_* .

Let $f(x) = (x - x_*)^p q(x)$, with p being a positive integer, q twice continuously differentiable, and $q(x_*) \neq 0$. (**Note:** $f'(x_*) = 0$.) In the following sub-problems, for x_k , let $f_k = f(x_k)$ and $e_k = |x_* - x_k|$.

- (a) Prove that Newton's method converges linearly for $f(x)$.

To show that Newton's method converges linearly for f , we have to show that there exists a constant C such that $C \in (0, 1)$ and for all $k \geq 0$,

$$|x_* - x_{k+1}| \leq C |x_* - x_k| \iff \left| \frac{x_{k+1} - x_*}{x_k - x_*} \right| \leq C.$$

The first derivative of f is given by

$$f'(x) = p(x - x_*)^{p-1} q(x) + (x - x_*)^p q'(x).$$

Then, the Newton iteration is given by

$$\begin{aligned} x_{k+1} &= x_k - \frac{f(x_k)}{f'(x_k)} \Rightarrow x_{k+1} = x_k - \frac{(x_k - x_*)^p q(x_k)}{p(x_k - x_*)^{p-1} q(x_k) + (x_k - x_*)^p q'(x_k)} \\ &\Rightarrow x_{k+1} = x_k - \frac{(x_k - x_*) q(x_k)}{pq(x_k) + (x_k - x_*) q'(x_k)}. \end{aligned}$$

Subtracting x_* from both sides and taking absolute values, we see that

$$\begin{aligned} |x_{k+1} - x_*| &= \left| x_k - x_* - \frac{(x_k - x_*) q(x_k)}{pq(x_k) + (x_k - x_*) q'(x_k)} \right| \\ &= |x_k - x_*| \left| 1 - \frac{q(x_k)}{pq(x_k) + (x_k - x_*) q'(x_k)} \right| \end{aligned}$$

Dividing both sides by $|x_k - x_*|$ and taking the limit as $k \rightarrow \infty$, we see that

$$\begin{aligned} \lim_{k \rightarrow \infty} \left| \frac{x_{k+1} - x_*}{x_k - x_*} \right| &= \lim_{k \rightarrow \infty} \left| 1 - \frac{q(x_k)}{pq(x_k) + (x_k - x_*) q'(x_k)} \right| \\ &= \lim_{k \rightarrow \infty} \left| 1 - \frac{q(x_k)}{pq(x_k)} \right| \\ &= \left| 1 - \frac{q(x_*)}{pq(x_*)} \right| \\ &= \left| 1 - \frac{1}{p} \right| \\ &< 1 \end{aligned}$$

because p is a positive integer and x_k converges to x_* as $k \rightarrow \infty$. Thus, Newton's method converges linearly for f .

(b) Consider the modified Newton iteration defined by

$$x_{k+1} = x_k - p \frac{f_k}{f'_k}.$$

Prove that if x_k converges to x_* , then the rate of convergence is quadratic, i.e., show that

$$|e_{k+1}| \leq C |e_k|^2$$

for some positive constant C as $x_k \rightarrow x_*$.

Using the Newton iteration from earlier, we see that the modified Newton iteration is given by

$$x_{k+1} = x_k - p \frac{(x_k - x_*) q(x_k)}{pq(x_k) + (x_k - x_*) q'(x_k)}.$$

Subtracting x_* from both sides and taking absolute values, we see that

$$\begin{aligned} |x_{k+1} - x_*| &= \left| x_k - x_* - p \frac{(x_k - x_*) q(x_k)}{pq(x_k) + (x_k - x_*) q'(x_k)} \right| \\ &= \left| \frac{(x_k - x_*)^2}{x_k - x_*} - p \frac{(x_k - x_*)^2 q(x_k)}{(x_k - x_*) (pq(x_k) + (x_k - x_*) q'(x_k))} \right| \\ &= \left| (x_k - x_*)^2 \left| \frac{1}{x_k - x_*} - p \frac{q(x_k)}{(x_k - x_*) (pq(x_k) + (x_k - x_*) q'(x_k))} \right| \right| \\ &= \left| (x_k - x_*)^2 \left| \frac{pq(x_k) + (x_k - x_*) q'(x_k) - pq(x_k)}{(x_k - x_*) (pq(x_k) + (x_k - x_*) q'(x_k))} \right| \right| \\ &= \left| (x_k - x_*)^2 \left| \frac{q'(x_k)}{(pq(x_k) + (x_k - x_*) q'(x_k))} \right| \right|. \end{aligned}$$

Dividing both sides by $|x_k - x_*|^2$ and taking the limit as $k \rightarrow \infty$, we see that

$$\begin{aligned} \lim_{k \rightarrow \infty} \left| \frac{x_{k+1} - x_*}{(x_k - x_*)^2} \right| &= \lim_{k \rightarrow \infty} \left| \frac{q'(x_k)}{(pq(x_k) + (x_k - x_*)q'(x_k))} \right| \\ &= \left| \frac{q'(x_*)}{pq(x_*)} \right|, \end{aligned}$$

where we have used the fact that x_k converges to x_* as $k \rightarrow \infty$. Thus, the modified Newton iteration converges quadratically for f .

- (c) Write MATLAB codes for both Newton and modified Newton methods. Apply these to the function

$$f(x) = (x - 1)^5 \exp(x)$$

and compare the results. Use $x_0 = 0$ as a starting point. I would like this in the form of two tables, one for Newton and one for modified Newton. Each of these should have two columns, the first for the iterates x_k and the second for the error $e_k = |x_k - x_*|$. Use `format long e` or equivalent formatting. Do enough iterations in each code to assure a relative error of 10^{-15} in the approximate solution x . Plot k vs e_k . Comment about the plots vs. the theory.

Note: See attached code and plots.

In the attached plots, we see that when Newton's method is applied to f , Newton's method takes 149 iterations (starting from an initial iterate of 0) to converge to a root of f . When viewing the plot of k vs. the logarithm of the relative error between the Newton iterates and the root $x_* = 1$, where k is the iteration counter, the resulting plot appears linear, which agrees with the theoretical result that was proved in (4a).

On the other hand, when the modified Newton applied is applied to f with an initial iterate of 0, the modified Newton method takes 6 iterates to converge to a root of f . Furthermore, when viewing the plot of k vs. the logarithm of the relative error between the modified Newton iterates and $x_* = 1$, the resulting plot shows quadratic behavior, which agrees with the theoretical result that was proved in (4b).

Problem 2

```
fprintf("Homework 2, Problem 2\n");
% To call our bisection implementation
addpath(genpath("C:\Users\eapge\Google Drive\1st Year\APPM
5600\Homework 1"));
close all;

T_i = 20;
T_s = -15;
alpha = 0.138 * 1e-6;
tol = 1e-13;

% x is in meters, t is in seconds
% 60 days after the cold snap, which is 60*24*60*60 seconds
t = 60*24*60*60;
% Temperature function
T = @(x) (T_i - T_s)*erf(x./(2*sqrt(alpha*t))) + T_s;
% Derivative of temperature function
T_prime = @(x) (T_i - T_s)*(1/sqrt(pi*alpha*t))*exp(-
((x.^2))./4*alpha*t));

a = 0;
b = 1;
x = linspace(a, b, 1000);

figure;
plot(x, T(x), "LineWidth", 2);
xlabel("x (meters below surface)");
ylabel("T(x, t) (temperature in degrees Celsius)");
title("Plot of T(x, t)");

% Find a root of T on the interval [x_0, x_bar] via bisection
[r, iters] = bisect(a, b, T, tol);
fprintf("Root computed via bisection: %0.16f (%d iterations)\n", r,
iters);

% Find a root of T on the interval [x_0, x_bar] via Newton's method
% Start off with an initial guess close to 0
x_0 = 0.01;
n_max = 100;
[r, iters, ~] = newton(x_0, T, T_prime, tol, n_max);
fprintf("Root computed via Newton's method: %0.16f (%d iterations, x_0
= %0.16f)\n", ...
r(iters), iters, x_0);

% Use Newton's method again, but make the initial guess the endpoint
of the
% interval [a, b], i.e., x_0 = b
x_0 = b;
[r, iters, ~] = newton(x_0, T, T_prime, tol, n_max);
```

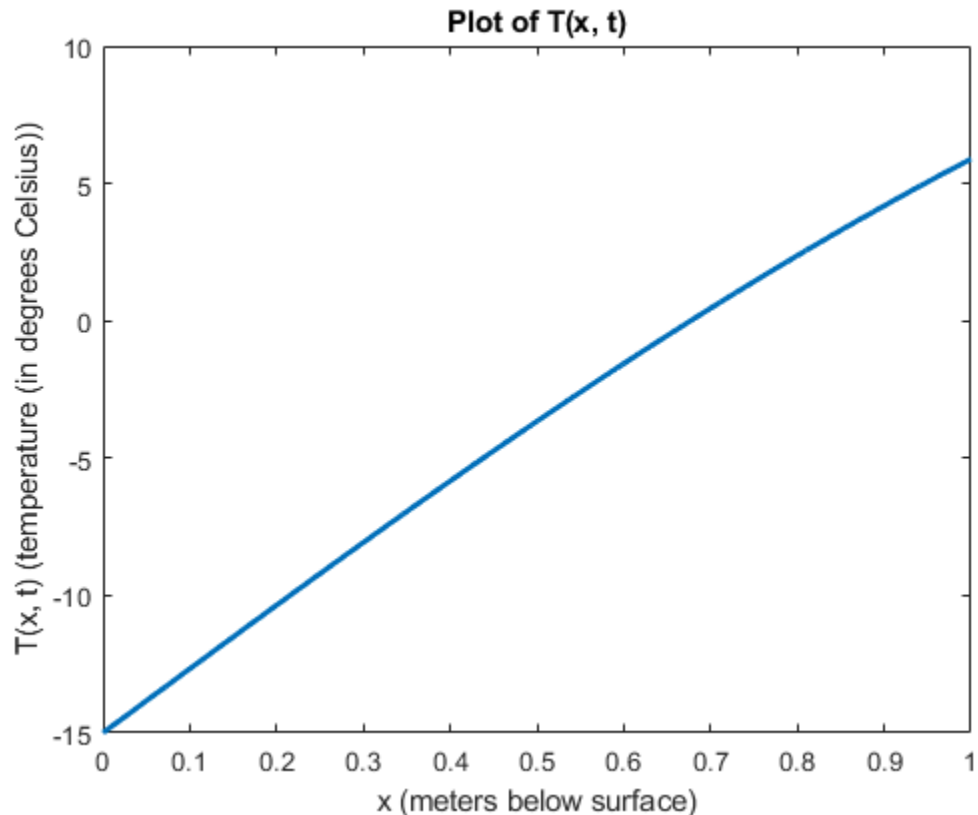
```
fprintf("Root computed via Newton's method: %0.16f (%d iterations, x_0
= %0.16f)\n", ...
r(iters), iters, x_0);
```

Homework 2, Problem 2

Root computed via bisection: 0.6769618544819309 (44 iterations)

Root computed via Newton's method: 0.6769618544819355 (14 iterations,
x_0 = 0.010000000000000000)

Root computed via Newton's method: 0.6769618544819372 (14 iterations,
x_0 = 1.0000000000000000)



Problem 4

```
fprintf("Homework 2, Problem 4\n");
close all;

p = 5;
f = @(x) ((x-1).^(p)).*exp(x);
f_prime = @(x) (p*(x-1).^(p-1)).*exp(x) + ((x-1).^(p)).*exp(x);
% f_prime = @(x) ((x-1).^(4)).*(x+4).*exp(x);

tol = 1e-15;
n_max = 150;

x_0 = 0;
[r, iters, ~] = newton(x_0, f, f_prime, tol, n_max);
```

```

fprintf("Root computed via Newton's method: %0.16f (%d iterations, x_0
    = %0.16f)\n", ...
        r(iters), iters, x_0);
fprintf("Relative error between approximate root and 1: %0.16e\n
\n", ...
        abs(r(iters) - 1));

Newton_iterates = r(1:iters);
Newton_relative_errors = abs(r(1:iters) - 1);
T1 = table(Newton_iterates, Newton_relative_errors);
disp(T1);

figure;
semilogy(1:iters, abs(r(1:iters) - 1), "LineWidth", 2);
xlabel("k");
ylabel("log(|e_{k}|)");
title("Plot of relative errors computed via Newton's method");

[r, iters, ~] = modified_newton(x_0, f, f_prime, p, tol, n_max);
fprintf("Root computed via modified Newton's method: %0.16f (%d
    iterations, x_0 = %0.16f)\n", ...
        r(iters), iters, x_0);
fprintf("Relative error between approximate root and 1: %0.16e\n
\n", ...
        abs(r(iters) - 1));

Modified_Newton_iterates = r(1:iters);
Modified_Newton_relative_errors = abs(r(1:iters) - 1);
T2 = table(Modified_Newton_iterates, Modified_Newton_relative_errors);
disp(T2);

figure;
semilogy(1:iters, abs(r(1:iters) - 1), "LineWidth", 2);
xlabel("k");
ylabel("log(|e_{k}|)");
title("Plot of relative errors computed via modified Newton's
    method");

```

Homework 2, Problem 4

Root computed via Newton's method: 0.99999999999999964 (149 iterations,
 $x_0 = 0.0000000000000000$)

Relative error between approximate root and 1: 3.5527136788005009e-15

Newton_iterates	Newton_relative_errors
0.0000000000000000e+00	1.0000000000000000e+00
2.5000000000000000e-01	7.5000000000000000e-01
4.26470588235294e-01	5.73529411764706e-01
5.56038694547586e-01	4.43961305452414e-01
6.53483280780737e-01	3.46516719219263e-01
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8.30314023003544e-01	1.69685976996456e-01

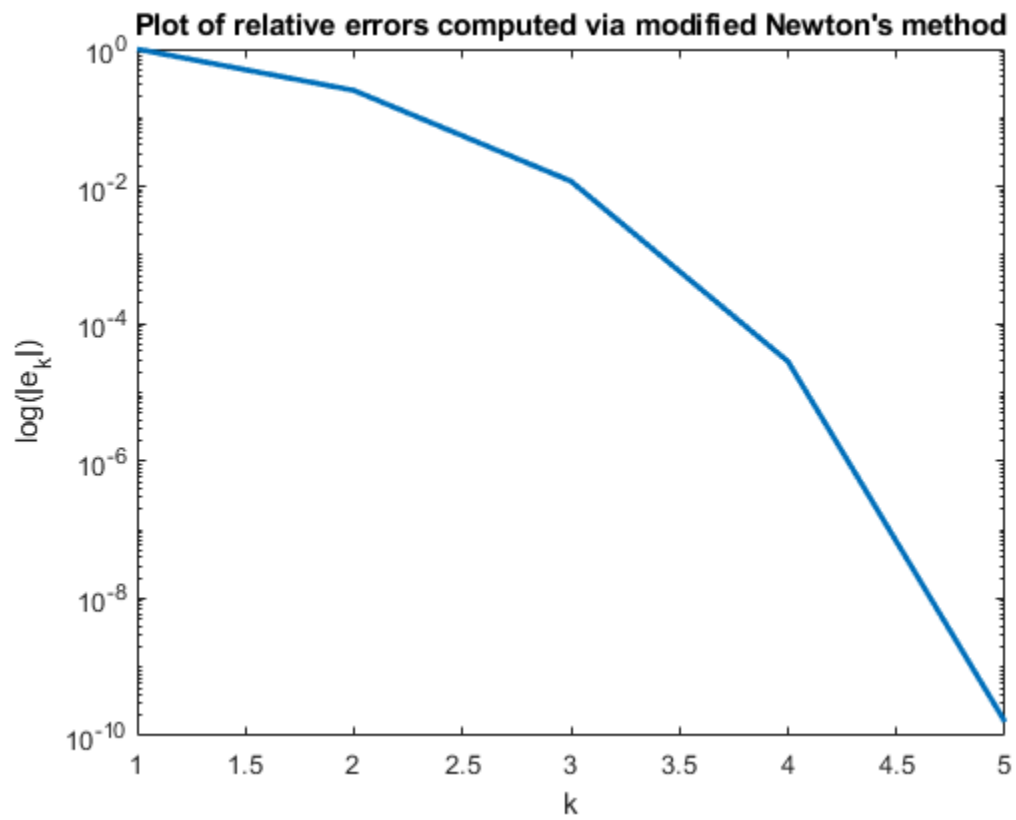
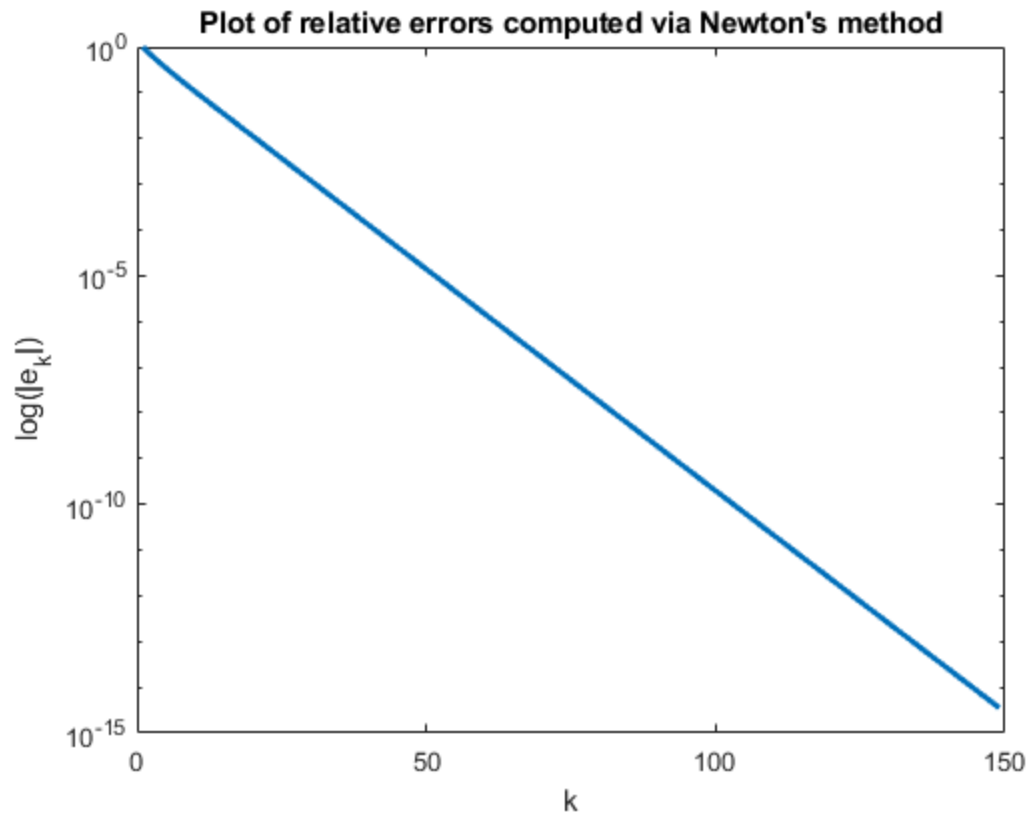
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9.99991146029771e-01	8.85397022876155e-06
9.99992916826953e-01	7.08317304731754e-06
9.99994333463569e-01	5.66653643097048e-06
9.99995466772140e-01	4.53322786042598e-06
9.99996373418534e-01	3.62658146635386e-06
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9.99998143191111e-01	1.85680888920015e-06
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9.9999999998179e-01	1.82109882729264e-12
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9.9999999999068e-01	9.32365296080206e-13
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9.9999999999875e-01	1.25122134875255e-13
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9.9999999999920e-01	8.01581023779363e-14
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9.9999999999949e-01	5.12923037376822e-14
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9.9999999999974e-01	2.63122856836162e-14
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9.9999999999983e-01	1.68753899743024e-14
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9.9999999999989e-01	1.08801856413265e-14
9.9999999999991e-01	8.65973959207622e-15
9.9999999999993e-01	6.88338275267597e-15
9.9999999999994e-01	5.55111512312578e-15
9.9999999999996e-01	4.44089209850063e-15
9.9999999999996e-01	3.55271367880050e-15

Root computed via modified Newton's method: 1.0000000000000000 (6 iterations, $x_0 = 0.0000000000000000$)
Relative error between approximate root and 1: 0.0000000000000000e+00

<u>Modified_Newton_iterates</u>	<u>Modified_Newton_relative_errors</u>
0.0000000000000000e+00	1.0000000000000000e+00
1.2500000000000000e+00	2.5000000000000000e-01
1.01190476190476e+00	1.19047619047619e-02
1.00002827734419e+00	2.82773441917517e-05
1.00000000015992e+00	1.59920743314501e-10
1.0000000000000000e+00	0.0000000000000000e+00



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```

%{
    Name: bisect
    Inputs: a - the left endpoint of the interval
           b - the right endpoint of the interval
           f - a continuous function defined on [a, b] such that
               f(a)f(b)
                   < 0
           tol - the user-specified tolerance
               When the approximate root, c, is such that |b-c| <
tol,
               bisect stops performing additional iterations
    Outputs: c - an approximate root of f on the interval [a, b]
           iters - the number of iterations it took for bisect to
               converge, which occurs either
                   - when f(c) == 0, to machine precision, or
                   - when |b-c| < tol
%}
function [c, iters] = bisect(a, b, f, tol)
    iters = 1;
    % Compute first approximation to root
    c = (a+b)/2;
    % While the interval of interest is large enough and
    % f evaluated at the current iterate is not 0,
    % perform bisection
    while ((f(c) ~= 0) && (abs(b-c) >= tol))
        iters = iters + 1;
        if (f(a)*f(c) < 0)
            b = c;
        else
            a = c;
        end
        c = (a+b)/2;
    end
end

```

Not enough input arguments.

Error in bisect (line 19)
 c = (a+b)/2;

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```

%{
Name: newton
Inputs:
    x_0 - the initial iterate
    f - the once differentiable function whose root we are trying to
    find
    f_prime - the derivative of f
    tol - the tolerance for the relative error in the iterates
    n_max - the maximum number of Newton iterations to perform
Outputs:
    r - the approximate root found by modified Newton's method
    iters
    ier - an error message
        If ier = 0, then Newton's method successfully converged to a
        root
            of f starting from x_0 within n_max iterations
            If ier /= 0, then Newton's method did not successfully
            converge
%}
function [r, iters, ier] = newton(x_0, f, f_prime, tol, n_max)
    iters = 1;
    ier = 1;

    % Set the initial iterate and compute the derivative of f at it
    r = zeros(n_max, 1);
    r(iters) = x_0;
    df = f_prime(r(iters));

    % Set the initial iterate and compute the derivative of f at it
    r = zeros(n_max, 1);
    r(iters) = x_0;
    df = f_prime(r(iters));

    % Perform Newton iterations while the max number of iterations
    % has not been reached yet
    while iters < n_max
        % If the derivative at the current iterate is 0, stop the
        iteration
        if df == 0
            ier = 2;
            break;
        end

        % Compute new iterate via Newton's method
        r(iters + 1) = r(iters) - f(r(iters))./df;
        iters = iters + 1;

        % If the relative error between the two most recent iterates
        is
        % below the provided tolerance, stop the iteration
        if abs((r(iters) - r(iters-1))/r(iters-1)) < tol
            ier = 0;

```

```
        break;
    end

    % Compute derivative at most recent iterate
    df = f_prime(r(iters));
end

% If the max number of iterations were performed, indicate this to
the
% user
if iters == n_max
    ier = 1;
end
end

Not enough input arguments.

Error in newton (line 22)
    r = zeros(n_max, 1);
```

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```

%{
Name: modified_newton
Inputs:
    x_0 - the initial iterate
    f - the once differentiable function whose root we are trying to
    find
    f_prime - the derivative of f
    p - the multiplicity of the root we are trying to find
    tol - the tolerance for the relative error in the iterates
    n_max - the maximum number of Newton iterations to perform
Outputs:
    r - the approximate root found by modified Newton's method
    iters
    ier - an error message
        If ier = 0, then Newton's method successfully converged to a
        root
            of f starting from x_0 within n_max iterations
            If ier /= 0, then Newton's method did not successfully
            converge
%}
function [r, iters, ier] = modified_newton(x_0, f, f_prime, p, tol,
n_max)
    iters = 1;
    ier = 1;

    % Set the initial iterate and compute the derivative of f at it
    r = zeros(n_max, 1);
    r(iters) = x_0;
    df = f_prime(r(iters));

    % Perform Newton iterations while the max number of iterations
    % has not been reached yet
    while iters < n_max
        % If the derivative at the current iterate is 0, stop the
        iteration
        if df == 0
            ier = 2;
            break;
        end

        % Compute new iterate via the modified Newton method
        r(iters + 1) = r(iters) - p*f(r(iters))./df;
        iters = iters + 1;

        % If the relative error between the two most recent iterates
        is
        % below the provided tolerance, stop the iteration
        if abs((r(iters) - r(iters-1))/r(iters-1)) < tol
            ier = 0;
            break;
        end
    end
end

```

```
        % Compute derivative at most recent iterate
        df = f_prime(r(iters));
    end

    % If the max number of iterations were performed, indicate this to
    the
    % user
    if iters == n_max
        ier = 1;
    end
end

Not enough input arguments.

Error in modified_newton (line 23)
    r = zeros(n_max, 1);
```

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