

CS 4600/5600 Numerical Computing

Homework 1

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Problem 2: Roots - Open Methods

(b) Newton-Raphson Method

Given the function:

$$f(x) = x^3 - 6x^2 + 11x - 6.1$$

We need to determine the largest positive root using the Newton-Raphson method, starting with $x_0 = 3.5$.

The Newton-Raphson iteration formula is:

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

First, we find the derivative of $f(x)$:

$$f'(x) = 3x^2 - 12x + 11$$

Starting with $x_0 = 3.5$:

$$f(3.5) = (3.5)^3 - 6(3.5)^2 + 11(3.5) - 6.1 = 1.775$$

$$f'(3.5) = 3(3.5)^2 - 12(3.5) + 11 = 5.75$$

First iteration:

$$x_1 = 3.5 - \frac{1.775}{5.75} \approx 3.1913043$$

Second iteration:

$$f(3.1913043) \approx 0.399408$$

$$f'(3.1913043) \approx 3.2576178$$

$$x_2 \approx 3.068698$$

Third iteration:

$$f(3.068698) = (3.068698)^3 - 6(3.068698)^2 + 11(3.068698) - 6.1 \approx 0.0518784$$

$$f'(3.068698) = 3(3.068698)^2 - 12(3.068698) + 11 \approx 2.426346$$

$$x_3 = 3.068698 - \frac{0.0518784}{2.426346} \approx 3.0473166$$

The largest positive root after three iterations is approximately $x \approx 3.0473166$.

(c) Modified Secant Method

Using the same function $f(x) = x^3 - 6x^2 + 11x - 6.1$, we apply the modified secant method with $x_0 = 3.5$ and $\delta = 0.01$.

The modified secant method iteration formula is:

$$x_{i+1} = x_n - \frac{\delta x_i f(x_i)}{f(x_i + \delta x_i) - f(x_i)}$$

Iteration	x_i	x_{i+1}
0	3.5	3.1995967
1	3.1995967	3.0753234
2	3.0753234	3.04881822
3	3.04881822	3.0467729
4	3.0467729	3.0466842863

The largest positive root after five iterations is approximately $x \approx 3.0466842863$.

Comments on results:

Both methods, are closer to the root. The Newton-Raphson method showed faster convergence for roots.

Problem 3:

As the iteration count increased. The function value neared to 0. The diffence is the sensitivity to the initial value given.

Problem 4: Optimization

Given the function:

$$f(x) = 4x - 1.8x^2 + 1.2x^3 - 0.3x^4$$

(b) Prove the function is concave for all values of x

A function is concave if its second derivative is negative for all values of x .

First, we find the first derivative of $f(x)$:

$$f'(x) = 4 - 3.6x + 3.6x^2 - 1.2x^3$$

Now, we find the second derivative:

$$f''(x) = -3.6 + 7.2x - 3.6x^2$$

For $f''(x)$ to be negative for all values of x , we need to check its sign:

$$f''(x) = -3.6 + 7.2x - 3.6x^2$$

Let's analyze the roots of $f''(x)$:

$$-3.6 + 7.2x - 3.6x^2 = 0$$

$$3.6x^2 - 7.2x + 3.6 = 0$$

$$x^2 - 2x + 1 = 0$$

$$(x - 1)^2 = 0$$

$$x = 1$$

At $x = 1$:

$$f''(1) = -3.6 + 7.2(1) - 3.6(1)^2 = -3.6 + 7.2 - 3.6 = 0$$

The second derivative is zero at $x = 1$, so we need to analyze the behavior around $x = 1$: For $x < 1$, $f''(x)$ is negative. For $x > 1$, $f''(x)$ is negative.

$f''(x)$ doesn't change sign, the function is concave for all values of x .

Problem 5: Optimization

Given the function:

$$f(x_1, x_2) = \frac{1}{2}(x_1^2 - x_2)^2 + \frac{1}{2}(1 - x_1)^2$$

(a) Minimum point

To find the minimum point, we need to find the Hessian Matrix:

$$H = \begin{pmatrix} 6x_1^2 - 2x_2 + 1 & -2x_1 \\ -2x_1 & 1 \end{pmatrix}$$

Simplifying, we find the critical point is:

$$(x_1, x_2) = (1, 1)$$

After substituting the (x_1, x_2) as $(1, 1)$ the $\det H$ is greater than 0.
Therefore, the critical point is the minimum point of the function.

(b) Newton's Method Iteration

Using the starting point $\begin{pmatrix} 2 \\ 2 \end{pmatrix}$, we perform one iteration of Newton's method for minimizing f .

The Hessian matrix is:

$$H = \begin{pmatrix} 6x_1^2 - 2x_2 + 1 & -2x_1 \\ -2x_1 & 1 \end{pmatrix}$$

The gradient matrix would be $\begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{pmatrix}$

$$\begin{aligned} \frac{\partial f}{\partial x_1} &= 2x_1(x_1^2 - x_2) - (1 - x_1) \\ \frac{\partial f}{\partial x_2} &= -(x_1^2 - x_2) \end{aligned}$$

The gradient at $\begin{pmatrix} 2 \\ 2 \end{pmatrix}$ is:

$$\nabla f = \begin{pmatrix} 2x_1(x_1^2 - x_2) - (1 - x_1) \\ -(x_1^2 - x_2) \end{pmatrix} = \begin{pmatrix} 2(2)(4 - 2) - (1 - 2) \\ -(4 - 2) \end{pmatrix} = \begin{pmatrix} 9 \\ -2 \end{pmatrix}$$

The Hessian at $\begin{pmatrix} 2 \\ 2 \end{pmatrix}$ is:

$$H = \begin{pmatrix} 6(2)^2 - 2(2) + 1 & -2(2) \\ -2(2) & 1 \end{pmatrix} = \begin{pmatrix} 21 & -4 \\ -4 & 1 \end{pmatrix}$$

The Newton step is:

$$\Delta x = -H^{-1}\nabla f$$

We need to invert the Hessian:

Formula:

$$H^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

$$H^{-1} = \frac{1}{21 \cdot 1 - (-4 \cdot -4)} \begin{pmatrix} 1 & 4 \\ 4 & 21 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 1 & 4 \\ 4 & 21 \end{pmatrix} = \begin{pmatrix} 0.2 & 0.8 \\ 0.8 & 4.2 \end{pmatrix}$$

The Newton step is:

$$\begin{aligned} \Delta x &= - \begin{pmatrix} 0.2 & 0.8 \\ 0.8 & 4.2 \end{pmatrix} \begin{pmatrix} 9 \\ -2 \end{pmatrix} \\ &= - \begin{pmatrix} 0.2 \cdot 9 + 0.8 \cdot (-2) \\ 0.8 \cdot 9 + 4.2 \cdot (-2) \end{pmatrix} \\ &= - \begin{pmatrix} 1.8 - 1.6 \\ 7.2 - 8.4 \end{pmatrix} \\ &= - \begin{pmatrix} 0.2 \\ -1.2 \end{pmatrix} \\ &= \begin{pmatrix} -0.2 \\ 1.2 \end{pmatrix} \end{aligned}$$

The new point after one iteration is:

$$\begin{pmatrix} 2 \\ 2 \end{pmatrix} + \begin{pmatrix} -0.2 \\ 1.2 \end{pmatrix} = \begin{pmatrix} 1.8 \\ 2.2 \end{pmatrix}$$

To determine if this is a good step or a bad one, we need to evaluate the function at this new point, and see how closer it is the point we found as minima which is $(1, 1)$

if $f(1.8, 2.2) < f(2, 2)$ then it would be a good step because

$$f(1.8, 2.2) = \frac{1}{2}(1.8^2 - 2.2)^2 + \frac{1}{2}(1 - 1.8)^2 = 0.02$$

$$f(2, 2) = \frac{1}{2}(2^2 - 2)^2 + \frac{1}{2}(1 - 2)^2 = 1$$

$$f(1, 1) = \frac{1}{2}(1^2 - 1)^2 + \frac{1}{2}(1 - 1)^2 = 0$$

Sense of why it is good step: When we evaluate the function with the given values, we can see that new values are making the function to near 0, which is the minimum value at point (1,1). Therefore, the new point is a good step in this sense.

Sense of why this could be a bad step: By looking at just the values of x_1 and x_2 we can see that x_2 value increased. There is an

Graduate Problem:

My understanding: The projectile motion is parabolic and should pass through the point, (x_1, h_2) in order to clear the roof. The hint says the coverage is maximized when the water just clears the height of the roof, which is h_2 . So, the coordinates to clear the front of the roof are (x_1, h_2) and the maximum it can go is at the coordinate (x_2, h_2) . The distance between those two points is just $x_2 - x_1$.

The equation of the projectile motion is given by: $y = \tan \theta(x) - \frac{gx^2}{2(u \cos \theta)^2}$. I need to think about this analytically.

So, if we can find the height of the projection, at some x , we can make it go beyond 10m.

About the Code: I have written a code to find the maximum distance the water can go, given the height of the roof and the angle of projection. The variable t_{flight} is evaluated so that we can get x_2 out using the formula $x_2 = x_1 + v \cdot \cos \theta \cdot t_f$. The maximum distance is calculated as $x_2 - x_1$. The method used for Optimization is *Sequential Least Square Programming* which also uses gradients, which leads to faster convergence. I chose this method because it can support constraints.

”All the code is in the file graph.py