

## Home Assignment – 3

Aarav Aryaman, 220012

### Problem:

Write a computer program (preferably in python) to minimize the following function using steepest descent method

$$f(x_1, x_2) = (x_1^2 + x_2 - 11)^2 + (x_2^2 + x_1 - 7)^2$$

Use  $x_1^{(0)} = x_2^{(0)} = 0$

Solution converges when norm of descent direction is less than 0.001

Do not use a constant step size, rather use an appropriate line search technique to find the optimum step size. Plot the function as a surface and as contours. Show the path of convergence connecting the points  $(x_1^{(k)}, x_2^{(k)})$  for  $k = 0, 1, 2, 3, \dots$

### Solution Algorithm:

#### 1. Define Objective Function and Gradient:

- Objective Function is defined as given in the problem statement.

$$f(x_1, x_2) = (x_1^2 + x_2 - 11)^2 + (x_2^2 + x_1 - 7)^2$$

- Gradient Function is defined to determine the direction of steepest descent.

$$\nabla f(x_1, x_2) = [4x_1(x_1^2 + x_2 - 11) + 2(x_2^2 + x_1 - 7), 2(x_1^2 + x_2 - 11) + 4x_2(x_2^2 + x_1 - 7)]$$

#### 2. Line Search Function:

- This function determines an appropriate step size (or learning rate) for the steepest descent method.
- It performs a backtracking line search to ensure that the step size is suitable for decreasing the objective function.
- Start with an initial *stepsize* (1 in this case) and decrease it by a factor of  $\tau = 0.5$  until the condition for sufficient decrease is satisfied.
- The condition checked is:

$$f(x - \text{stepsize} * \text{gradient}) \leq f(x) - \beta * \text{stepsize} * ||\text{gradient}||^2$$

where  $\beta$  is a small constant (0.1 in this case), ensuring the step size provides a sufficient decrease in the function value.

#### 3. Steepest Descent Method:

- Set up the parameters: maximum iterations ( $\text{maxit} = 10000$ ), convergence tolerance ( $\text{epsilon} = 0.001$ ) and initial value ( $x = [0, 0]$ ).
- For each iteration:

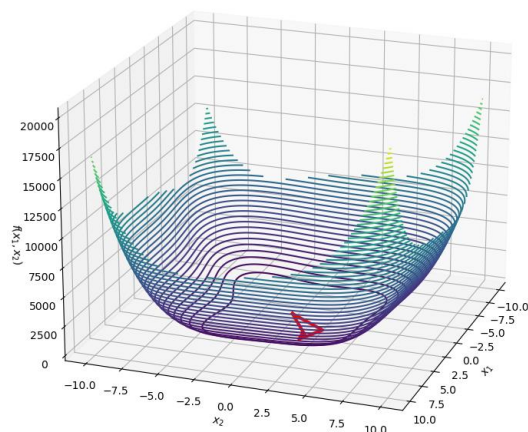
- Store the current values of  $x_1$  and  $x_2$  for later plotting.
  - Compute the gradient at the current point.
  - Check the norm of the gradient ( $b$ ). If it's less than the tolerance (*epsilon*), stop the iteration as the solution is sufficiently close to the minimum.
  - Use the line search function to determine the optimal step size.
  - Update the current point  $x$  by moving in the direction of the negative gradient scaled by the step size.
- If the gradient's norm ( $b$ ) is below *epsilon*, the algorithm stops early, assuming convergence. Otherwise, it continues until reaching the maximum number of iterations.

#### 4. Output Results:

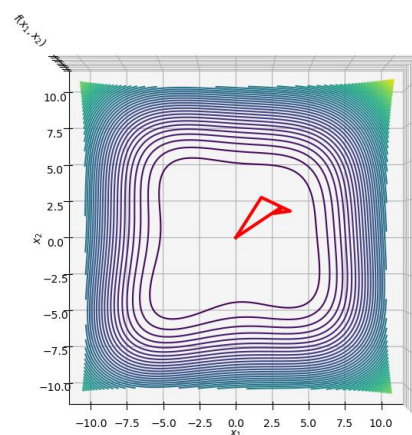
- Optimum stepsize: 0.015625
- Minimum value: 7.368947568261478e-09
- Minimum location: [2.99999944 1.99997953]
- Iteration: 16

#### 5. Plotting:

- Create a mesh grid for  $x_1$  and  $x_2$  over a specified range.
- Calculate the objective function values over this grid.
- Plot the 3D surface of the objective function and overlay the path taken by the steepest descent method in red.
- Contour is similar to the 3D plot but viewed from above to show the trajectory of the optimization in the  $x_1$ - $x_2$  plane.



Surface Plot



Contour Plot