Numerical Optimization 2024 - Homework $4\,$

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April 16, 2024

Problem 1.

The contour plot of the Rosenbrock function, points (1,1),(0,0),(-1,1),(0,1) and the function values where f(x) attains the values 0,1,4 are shown in Figure: 1.

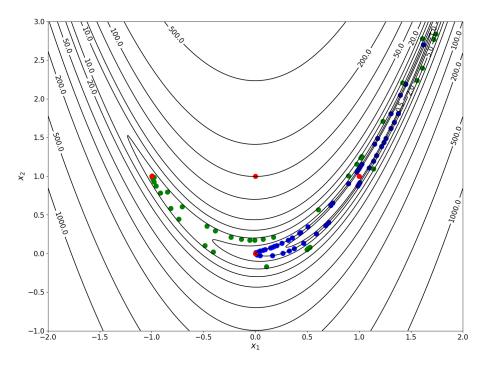


Figure 1: Contour plot of $f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2$

Problem 2.

The gradient of the Rosenbrock function is defined as:

$$\nabla f(x) = \begin{bmatrix} -400x_1(x_2 - x_1^2) - 2(1 - x_1) \\ 200(x_2 - x_1^2) \end{bmatrix}$$
 (1)

The Hessian of the Rosenbrock function is defined as:

Iteration 0: 4

$$\nabla^2 f(x) = \begin{bmatrix} -400(x_2 - 3x_1^2) + 2 & -400x_1 \\ -400x_1 & 200 \end{bmatrix}$$
 (2)

Steepest descent moves along $p_k = -\nabla f(x_k)$ at each step. For α_k as the global minimizer of $\phi(\alpha) = f(x_k + \alpha p_k)$, the following statements hold:

- i) I expect the steepest descent method to converge to the global min. of the function at (1,1) (depending on the starting point and α , the algorithm will find the local min.)
- ii) The rate of convergence will be linear. I plotted the graph for the convergence, C is less than 10^{-5} after 5000 iterations, shown in Figure: 3. (starting at (-1,1) after 10000 iterations, f(x) = 1.0256665448988478e 08, see below for the exact values)
- iii) Starting at $x_0 = (-1, 1)$, the trajection does not go directly to the limit but move up and down and follow the deep valley of the function to reach the global min. at (1, 1), shown in Figure: 7.

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Iteration 1: 101.0
Iteration 2: 101.0
Iteration 3: 0.980101239863517
Iteration 4: 0.7790048295335187
Iteration 5: 0.7487672432382151
Iteration 6: 0.7281119086406248
Iteration 7: 0.7045080495384312
Iteration 9989: 1.0309324143070513e-08
Iteration 9990: 1.0304426606525326e-08
Iteration 9991: 1.0299695480712044e-08
Iteration 9992: 1.0294813298468107e-08
Iteration 9993: 1.0290097502686015e-08
Iteration 9994: 1.0285230626158758e-08
Iteration 9995: 1.0280530111705617e-08
Iteration 9996: 1.0275678491906015e-08
Iteration 9997: 1.0270993209860194e-08
Iteration 9998: 1.0266156798725102e-08
Iteration 9999: 1.0261486700313638e-08
Iteration 10000: 1.0256665448988478e-08
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Problem 3.

Line search with backtracking is defined as:

choose
$$\bar{\alpha} > 0, \rho, c \in (0, 1)$$
; Set $\alpha \leftarrow \bar{\alpha}$;
repeat until $f(x_k + \alpha p_k) \le f(x_k) + c\alpha \nabla f_k^T p_k$
$$\alpha \leftarrow \rho \alpha$$
end at $\alpha_k = \alpha$

i) The plot for the backtracking line search method (BLSM) for $\rho = 0.5, c = 0.0001, \alpha = 1$ is shown in Figure: 2, after 5 iterations the function begins to converge. Starting from x_k at (1.2, 1.2), after 5000 iterations the function value is minimized to the order of 10^{-5} , shown in Figure: 3. The corresponding contour and line plots for $\alpha = 1$ and $\alpha = 0.3$ are shown in Figure: 4, Figure: 5 and Figure: 6.

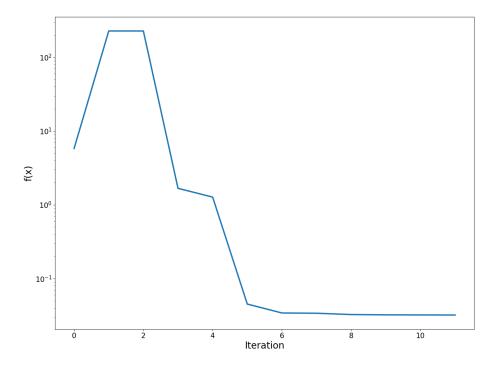


Figure 2: BLSM for $\rho=0.5, c=0.0001, \alpha=1$ for 10 iterations

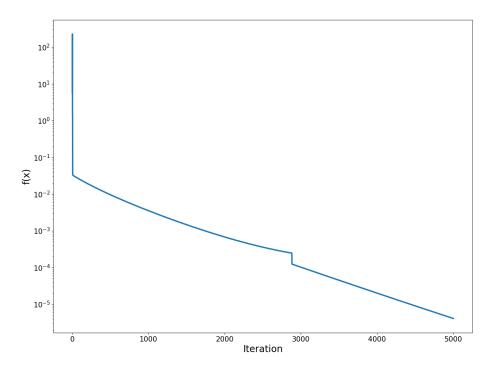


Figure 3: BLSM for hyperparameters $\rho=0.5, c=0.0001, \alpha=1$ for 5000 iterations

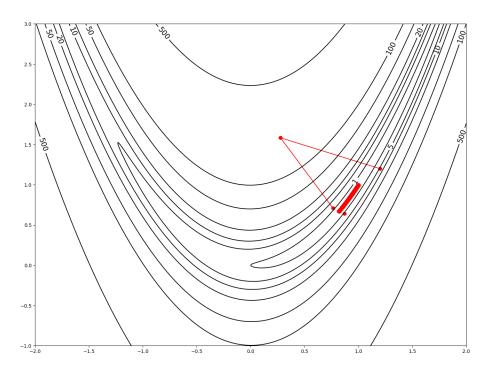


Figure 4: Steepest descent with BLSM after 5000 iterations for hyperparameters ($\rho=0.5, c=0.0001, \alpha=1$)

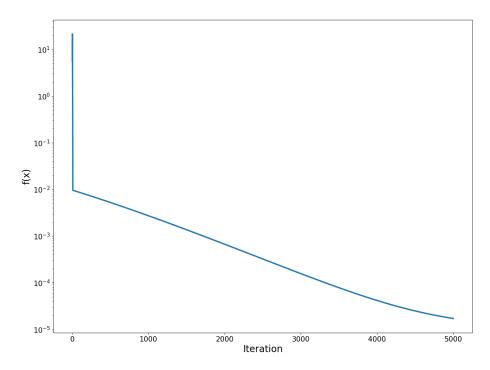


Figure 5: BLSM for hyperparameters $\rho=0.5, c=0.0001, \alpha=0.3$ for 5000 iterations

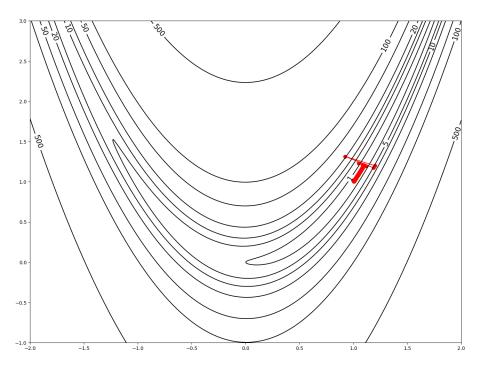


Figure 6: Steepest descent with BLSM after 5000 iterations for hyperparameters ($\rho=0.5, c=0.0001, \alpha=0.3$)

ii) For the starting point (-1,1), the plot is shown in Figure: 7 (included the starting point (-1,1) in the submission as it is easier to show the changes). For the starting point (-1.2,1), the plot is shown in Figure: 8. With increasing α values the convergence rate to reach the global min. is faster.

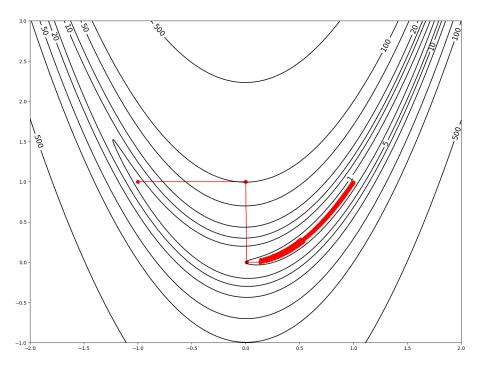


Figure 7: Steepest descent with BLSM after 5000 iterations for stating point (-1,1) with hyperparameters $(\rho=0.5,c=0.0001,\alpha=1)$

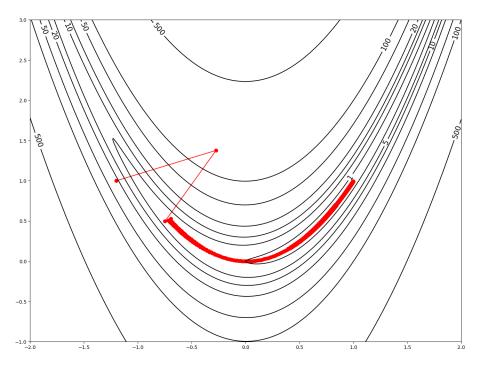


Figure 8: Steepest descent with BLSM after 5000 iterations for stating point (-1.2, 1) with hyperparameters $(\rho = 0.5, c = 0.0001, \alpha = 1)$