

Introduction to Engineering Optimization (ME6806)



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Multi-variable Optimization Algorithms



The iterative procedure of the Davidon-Fletcher-Powell (DFP) method can be described as follows:

- **1.** Start with an initial point \mathbf{X}_1 and a $n \times n$ positive definite symmetric matrix $[B_1]$ to approximate the inverse of the Hessian matrix of f. Usually, $[B_1]$ is taken as the identity matrix [I]. Set the iteration number as i = 1.
- **2.** Compute the gradient of the function, ∇f_i , at point \mathbf{X}_i , and set

$$\mathbf{S}_i = -[B_i] \nabla f_i$$

3. Find the optimal step length λ_i^* in the direction \mathbf{S}_i and set

$$\mathbf{X}_{i+1} = \mathbf{X}_i + \lambda_i^* \mathbf{S}_i$$

4. Test the new point \mathbf{X}_{i+1} for optimality. If \mathbf{X}_{i+1} is optimal, terminate the iterative process. Otherwise, go to step 5.





5. Update the matrix $[B_i]$, as

 $[B_{i+1}] = [B_i] + [M_i] + [N_i]$

where

$$[M_i] = \lambda_i^* \frac{\mathbf{S}_i \mathbf{S}_i^{\mathrm{T}}}{\mathbf{S}_i^T \mathbf{g}_i}$$
$$[N_i] = -\frac{([B_i]\mathbf{g}_i)([B_i]\mathbf{g}_i)^{\mathrm{T}}}{\mathbf{g}_i^T [B_i]\mathbf{g}_i}$$
$$\mathbf{g}_i = \nabla f(\mathbf{X}_{i+1}) - \nabla f(\mathbf{X}_i) = \nabla f_{i+1} - \nabla f_i$$

6. Set the new iteration number as i = i + 1, and go to step 2.



Minimize $f(x_1, x_2) = x_1 - x_2 + 2x_1^2 + 2x_1x_2 + x_2^2$ from the starting point $\mathbf{X}_1 = \{ \begin{smallmatrix} 0 \\ 0 \end{smallmatrix} \}$ using the DFP method with

$$\begin{bmatrix} B_1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \varepsilon = 0.01$$

<u>Solution</u>

Iteration 1 (i = 1)

Here

$$\nabla f_1 = \nabla f(\mathbf{X}_1) = \begin{cases} 1 + 4x_1 + 2x_2 \\ -1 + 2x_1 + 2x_2 \end{cases} \Big|_{(0,0)} = \begin{cases} 1 \\ -1 \end{cases}$$



Davidon–Fletcher–Powell Method - Example

hence

$$\mathbf{S}_1 = -\begin{bmatrix} B_1 \end{bmatrix} \nabla f_1 = -\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

To find the minimizing step length λ_1^* along S_1 , we minimize

$$f(\mathbf{X}_1 + \lambda_1 \mathbf{S}_1) = f\left(\left\{ \begin{matrix} 0\\0 \end{matrix} \right\} + \lambda_1 \left\{ \begin{matrix} -1\\1 \end{matrix} \right\} \right) = f(-\lambda_1, \lambda_1) = \lambda_1^2 - 2\lambda_1$$

with respect to λ_1 . Since $df/d\lambda_1 = 0$ at $\lambda_1^* = 1$, we obtain

$$\mathbf{X}_2 = \mathbf{X}_1 + \lambda_1^* \mathbf{S}_1 = \begin{cases} 0\\ 0 \end{cases} + 1 \begin{cases} -1\\ 1 \end{cases} = \begin{cases} -1\\ 1 \end{cases}$$

Davidon–Fletcher–Powell Method - Example



Since $\nabla f_2 = \nabla f(\mathbf{X}_2) = \{ \stackrel{-1}{-1} \}$ and $\|\nabla f_2\| = 1.4142 > \varepsilon$, we proceed to update the matrix $[B_i]$ by computing

$$\mathbf{g}_{1} = \nabla f_{2} - \nabla f_{1} = \begin{cases} -1 \\ -1 \\ -1 \end{cases} - \begin{cases} 1 \\ -1 \\ -1 \end{cases} = \begin{cases} -2 \\ 0 \\ 0 \end{cases}$$
$$\mathbf{S}_{1}^{T} \mathbf{g}_{1} = \{ -1 \\ 1 \\ 1 \\ \end{bmatrix} \begin{cases} -2 \\ 0 \\ -1 \\ 1 \\ \end{bmatrix} = 2$$
$$\mathbf{S}_{1} \mathbf{S}_{1}^{T} = \begin{cases} -1 \\ 1 \\ 1 \\ \end{bmatrix} \{ -1 \quad 1 \} = \begin{bmatrix} 1 & -1 \\ -1 & 1 \\ \end{bmatrix}$$
$$[B_{1}] \mathbf{g}_{1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \end{bmatrix} \begin{cases} -2 \\ 0 \\ \end{bmatrix} = \begin{cases} -2 \\ 0 \\ \end{bmatrix}$$
$$([B_{1}] \mathbf{g}_{1})^{T} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \end{bmatrix} \begin{cases} -2 \\ 0 \\ \end{bmatrix}^{T} = \{ -2 \quad 0 \}$$

Davidon–Fletcher–Powell Method - Example



$$\mathbf{g}_{1}^{\mathrm{T}}[B_{1}]\mathbf{g}_{1} = \{-2 \ 0\} \begin{bmatrix} 1 & 0\\ 0 & 1 \end{bmatrix} \{ -2\\ 0 \} = \{-2 \ 0\} \{ -2\\ 0 \} = 4$$
$$[M_{1}] = \lambda_{1}^{*} \frac{\mathbf{S}_{1} \mathbf{S}_{1}^{\mathrm{T}}}{\mathbf{S}_{1}^{\mathrm{T}} \mathbf{g}_{1}} = 1 \left(\frac{1}{2} \right) \begin{bmatrix} 1 & -1\\ -1 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2}\\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$$
$$[N_{1}] = -\frac{([B_{1}]\mathbf{g}_{1})([B_{1}]\mathbf{g}_{1})^{\mathrm{T}}}{\mathbf{g}_{1}^{\mathrm{T}}[B_{1}]\mathbf{g}_{1}} = -\frac{\{ -2\\ 0 \}}{4} \{ -2 \ 0 \} = -\frac{1}{4} \begin{bmatrix} 4 & 0\\ 0 & 0 \end{bmatrix} = -\begin{bmatrix} 1 & 0\\ 0 & 0 \end{bmatrix}$$
$$B_{2}] = [B_{1}] + [M_{1}] + [N_{1}] = \begin{bmatrix} 1 & 0\\ 0 & 1 \end{bmatrix} + \begin{bmatrix} \frac{1}{2} & -\frac{1}{2}\\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix} + \begin{bmatrix} -1 & 0\\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0.5 & -0.5\\ -0.5 & 1.5 \end{bmatrix}$$



Iteration 2 (i = 2)

The next search direction is determined as

$$\mathbf{S}_2 = -[B_2]\nabla f_2 = -\begin{bmatrix} 0.5 & -0.5 \\ -0.5 & 1.5 \end{bmatrix} \begin{bmatrix} -1 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

To find the minimizing step length λ_2^* along S_2 , we minimize

$$f(\mathbf{X}_{2} + \lambda_{2}\mathbf{S}_{2}) = f\left(\left\{ \begin{array}{c} -1\\1 \end{array}\right\} + \lambda_{2} \left\{ \begin{array}{c} 0\\1 \end{array}\right\} \right) = f\left(\left\{ \begin{array}{c} -1\\1 + \lambda_{2} \end{array}\right\} \right)$$
$$= -1 - (1 + \lambda_{2}) + 2(-1)^{2} + 2(-1)(1 + \lambda_{2}) + (1 + \lambda_{2})^{2}$$
$$= \lambda_{2}^{2} - \lambda_{2} - 1$$

with respect to λ_2 . Since $df/d\lambda_2 = 0$ at $\lambda_2^* = \frac{1}{2}$,



$$f(\mathbf{X}_2 + \lambda_2 \mathbf{S}_2) = \lambda_2^2 - \lambda_2 - 1$$

with respect to λ_2 . Since $df/d\lambda_2 = 0$ at $\lambda_2^* = \frac{1}{2}$, we obtain $\mathbf{X}_3 = \mathbf{X}_2 + \lambda_2 \mathbf{S}_2 = \begin{cases} -1\\1 \end{cases} + \frac{1}{2} \begin{cases} 0\\1 \end{cases} = \begin{cases} -1\\1.5 \end{cases}$

This point can be identified to be optimum since

$$\nabla f_3 = \begin{cases} 0\\ 0 \end{cases}$$
 and $||\nabla f_3|| = 0 < \varepsilon$



Thanks