

Homework No 8 (Quasi-Newton DFP method)

Use the DFP quasi-Newton update of the inverse Hessian approximation:

$$H_{k+1} = H_k - \frac{H_k y_k y_k^T H_k}{y_k^T H_k y_k} + \frac{p_k p_k^T}{y_k^T p_k} \quad (1)$$

Starting with

$$H_1 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ being the unit matrix}$$

and minimize the function:

$$f(x_1, x_2) = 100(x_1^2 - x_2)^2 + (1 - x_1)^2$$

with initial guess $X_1 = (-2, -2)^T$

$$\text{Use accuracy condition } \|\nabla f_k\| < \varepsilon \quad (2)$$

with

$$\varepsilon = 10^{-3}$$

Compute the search direction:

$$p_k = -H_k \nabla f_k$$

set:

$$x_{k+1} = x_k + \alpha_k p_k$$

where α_k is computed from a line search to satisfy Wolfe condition.

$$\text{Define } \begin{aligned} p_k &= x_{k+1} - x_k \\ y_k &= \nabla f_{k+1} - \nabla f_k \end{aligned}$$

Compute H_{k+1} by means of (1).

Check if accuracy condition (2) is satisfied if not set $k \leftarrow k + 1$ and repeat the loop.

Please write your own code or use the code `dfpmintest.f` based on Numerical Recipes:

<http://www.library.cornell.edu/nr/bookpdf/f10-7.pdf>

Use `ITMAX` maximal number of iterations as `ITMAX = 200`

If you use above code please flowchart it. Explain why the problem of minimizing f is more difficult by considering its condition number.