



# One little Practice about Newton Method

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# Problem

求

$$\operatorname{argmin} f(x, y) = x^2 + y^2$$

从  $(x, y) = (1, 1)$  开始迭代, 函数的求导用差分近似, 间隔 0.001,

最终输出误差在 **1e-6** 范围内即可

具体要求: 写一个函数 `gradient(f)`, 然后 `main()` 调用该函数

# Newton Method

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2 + \dots + \frac{1}{n!}f^{(n)}(x_0)(x - x_0)^n \dots$$

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2$$

$$f'(x) = f'(x_0) + f''(x_0)(x - x_0) = 0$$

$$x = x_0 - \frac{f'(x_0)}{f''(x_0)}$$

# Iteration

$$x_{t+1} = x_t - \frac{f'(x_t)}{f''(x_t)}$$

# Hesse Matrix

$$\mathbf{H} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} & \cdots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

$$\mathbf{H}_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j}$$

# Hesse Matrix

$$f(x, y) = x^2 + y^2$$

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial^2 x} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial^2 y} \end{bmatrix}$$

$$\frac{\partial f}{\partial x} = 2x$$

$$\frac{\partial f}{\partial y} = 2y$$

# Hesse Matrix

$$f(x, y) = x^2 + y^2$$

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial^2 x} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial^2 y} \end{bmatrix}$$

$$\frac{\partial f}{\partial x} = 2x$$

$$\frac{\partial f}{\partial y} = 2y$$

$$= \begin{bmatrix} \frac{\partial(2x)}{\partial x} & \frac{\partial(2y)}{\partial x} \\ \frac{\partial(2x)}{\partial y} & \frac{\partial(2y)}{\partial y} \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

# Define in code

- `def f(x):`
  - `return x[0] ** 2 + x[1] ** 2`
- `def gradient(x):`
  - `return np.array([2*x[0], 2*x[1]])`
- `hessian_inv = np.array([[0.5, 0.0],`
  - `[0.0, 0.5]])`



# Main loop

```
• def newton(x):  
•     '''  
•     x -= f'(x) / f''(x)  
•     -= hessian_inv * f'(x)  
•     '''  
•     while True:  
•         new_x = x - np.dot(hessian_inv, gradient(x))  
•         if abs(f(x) - f(new_x)) < 0.000001:  
•             break  
•         x = new_x  
•     return x
```

```
if __name__ == '__main__':
```

- `def main():`
- `x = np.array([1, 1])`
- `ans = newton(x)`
- `print("{:.8f} {:.8f}".format(ans[0], ans[1]))`

# Thank you

- End