

$$\min_x f(x)$$

Gradient descent

update rule: $x_{t+1} = x_t + \eta_t \underbrace{(-f'(x_t))}_{\text{what is so special about } -f'(x)?}$

Taylor's series: $f(\underbrace{x + \eta d}_{\bar{x} = x + \eta d}) = \underbrace{f(x)}_{\text{evaluations are all at } x} + \eta d \underbrace{f'(x)}_{\text{evaluations are all at } x} + \frac{\eta^2 d^2}{2} \underbrace{f''(x)}_{\text{evaluations are all at } x} + \dots$

Local information gives global information

$$f(x + \eta d) = f(x) + \eta d f'(x) + \underbrace{f''(x) \frac{\eta^2 d^2}{2} + \dots}_{\text{higher order terms. } \times}$$

\downarrow
 \uparrow Small positive step-size \uparrow "direction"
 (circled around η)

For small enough η

$$f(x + \eta d) \approx f(x) + \eta d f'(x)$$

$$\Rightarrow \boxed{f(x + \eta d) - f(x)} \approx \eta d f'(x)$$

\uparrow Function evaluation at updated point along "direction" d \uparrow Function evaluation at the current point

Want to choose a direction d s.t.

$$\boxed{f(x + \eta d) - f(x) < 0}$$

$$\Rightarrow \text{Want 'd' s.t. } \eta d f'(x) < 0$$

Want $\eta d f'(x) < 0$

↑
Small
+ve
Constant

\Rightarrow

Want
d s.t

$$\boxed{-d f'(x) < 0}$$

For the choice of $\boxed{d = -f'(x)}$

$$d f'(x) = - (f'(x))^2 < 0$$

Higher dimensions

$$\underline{f(x_1, x_2)} = \underline{x_1^2 + 4x_2 + 8x_2^2}$$

Derivative \Leftrightarrow Gradient
vector of
partial derivatives.

$$\nabla f\left(\begin{bmatrix} a \\ b \end{bmatrix}\right) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \Big|_{x_1=a} \\ \frac{\partial f}{\partial x_2} \Big|_{x_2=b} \end{bmatrix}$$