

Stochastic gradient descent.

loop $\{$

for $i = 1$ to $i = m \{$

$$\theta_j := \theta_j - \alpha (h_0(x^{(i)}) - y^{(i)}) x_j^{(i)} \beta$$

(for $j = 0 \rightarrow n$)

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Useful for large datasets
(i.e. large value of m)

Matrix derivatives

$f: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$
gradient of $f(A)$ w.r.t A

$$\nabla_A f(A) = \begin{bmatrix} \frac{\partial f}{\partial A_{11}} & \frac{\partial f}{\partial A_{12}} & \dots & \frac{\partial f}{\partial A_{1n}} \\ \frac{\partial f}{\partial A_{21}} & \frac{\partial f}{\partial A_{22}} & \dots & \frac{\partial f}{\partial A_{2n}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f}{\partial A_{m1}} & \frac{\partial f}{\partial A_{m2}} & \dots & \frac{\partial f}{\partial A_{mn}} \end{bmatrix}_{m \times n} \Rightarrow (i, j)^{\text{th}} \text{ element} \downarrow \frac{\partial f}{\partial A_{ij}}$$

derivative of f w.r.t. A

$$A = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & & & A_{2n} \\ \vdots & & & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{bmatrix}_{m \times n}$$

If A is a 2×2 matrix $\Rightarrow A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$

and funcⁿ $f: \mathbb{R}^{2 \times 2} \rightarrow \mathbb{R}$

$$f(A) = \frac{3}{2} A_{11} + 5 A_{12}^2 + A_{21} A_{22}$$

$$\nabla_A f(A) = \begin{bmatrix} \frac{3}{2} & 10 A_{12} \\ A_{22} & A_{21} \end{bmatrix}$$

Trace of a matrix

$A \rightarrow n \times n$ matrix

$$\text{tr } A = \sum_{i=1}^n A_{ii} = A_{11} + A_{22} + \dots + A_{nn} \quad (\text{only for square matrix})$$