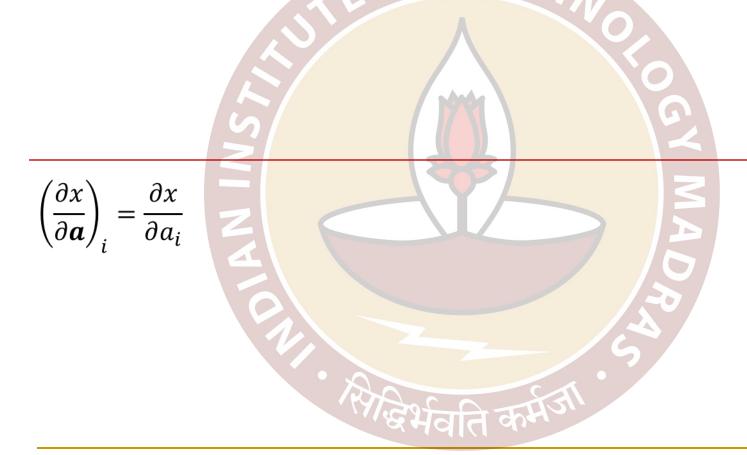


Motivation

- Machine Learning training requires one to evaluate how one vector changes with respect to another
 - For example how output changes with respect to parameters
- This requires "matrix" calculus
- We will see some initial relations in this video
 - It is useful to understand these, but most of the course can be understood without this portion too.
- More advanced relations exist
 - Suggested resource : https://explained.ai/matrix-calculus/index.html

Scalars and vectors

$$\left(\frac{\partial \mathbf{a}}{\partial x}\right)_i = \frac{\partial a_i}{\partial x}$$



Vectors and vectors

$$\left(\frac{\partial \boldsymbol{a}}{\partial \boldsymbol{b}}\right)_{ij} = \frac{\partial a_i}{\partial b_j}$$

$$\frac{\partial}{\partial \mathbf{x}}(\mathbf{x} \cdot \boldsymbol{a}) = \frac{\partial}{\partial \mathbf{x}}(\mathbf{x}^T \boldsymbol{a}) = \frac{\partial}{\partial \mathbf{x}}(\boldsymbol{a}^T \mathbf{x}) = \boldsymbol{a}$$

Matrices and vectors

$$\frac{\partial}{\partial \mathbf{x}}(AB) = \frac{\partial A}{\partial \mathbf{x}}B + A\frac{\partial B}{\partial \mathbf{x}}$$

$$\frac{\partial}{\partial \mathbf{x}}(\mathbf{x}^T A \mathbf{x}) = (A + A^T) \mathbf{x}$$

Proof slightly involved (we will see a quick verification on the next slide)

NOTE: For symmetric matrices, $\frac{\partial}{\partial \mathbf{x}}(\mathbf{x}^T A \mathbf{x}) = 2A\mathbf{x}$

Derivative of the quadratic form

