Mathematics reminders for deep learning

Part 2: Differential Calculus

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Differential of a function scalar input and scalar output

- $f: \mathbb{R} \to \mathbb{R}: x \to f(x)$ f is differentiable
- $\bullet \ y = f(x)$
- f(x+h) = f(x) + f'(x)h + o(h) $(\lim_{h\to 0} \frac{o(h)}{h} = 0)$
- dy = f'(x)dx i.e: f is "locally linear"
- $\frac{dy}{dx} \equiv f'(x)$ (notation)
- $dy = \frac{dy}{dx} dx$ ("local scale factor")
- All values are scalar

Differential of a composed function scalar input and scalar output

•
$$f: \mathbb{R} \rightarrow \mathbb{R}: x \rightarrow f(x)$$

f is differentiable

•
$$y = f(x)$$

•
$$g: \mathbb{R} \rightarrow \mathbb{R}: y \rightarrow g(y)$$

g is differentiable

•
$$z = g(y)$$

•
$$(g \circ f)'(x) = (g' \circ f)(x).f'(x) = g'(y).f'(x)$$

•
$$dy = \frac{dy}{dx} dx$$

$$dz = \frac{dz}{dy}dy$$

•
$$dz = \frac{dz}{dv} \cdot \frac{dy}{dx} dx$$

$$\frac{dz}{dx} = \frac{dz}{dy} \cdot \frac{dy}{dx}$$

Differential of a function of a vector vector input and scalar output

•
$$f: \mathbb{R}^N \to \mathbb{R}: x \to f(x)$$

f is differentiable

•
$$y = f(x)$$

$$x = (x_i)_{(1 \le i \le N)}$$

•
$$f(x + h) = f(x) + \text{grad } f(x) \cdot h + o(||h||)$$

•
$$dy = \operatorname{grad} f(x) \cdot dx = \sum_{i=1}^{i=n} \frac{\partial f}{\partial x_i}(x) \cdot dx_i = \sum_{i=1}^{i=n} \frac{\partial y}{\partial x_i} \cdot dx_i = \frac{\partial y}{\partial x} \cdot dx$$

•
$$\frac{\partial y}{\partial x} \equiv \frac{\partial f}{\partial x}(x) = \operatorname{grad} f(x)$$
 $\frac{\partial y}{\partial x_i} \equiv \frac{\partial f}{\partial x_i}(x)$ (notations)

- y, dy and f(x) are scalars;
- x, dx and h are "regular" (column) vectors;
- $\frac{\partial y}{\partial x}$ is a transpose (row) vector.

Differential of a vector function of a vector vector input and vector output

•
$$f: \mathbb{R}^N \to \mathbb{R}^P: x \to f(x)$$

f is differentiable

•
$$y = f(x)$$
 $x = (x_i)_{(1 \le i \le N)}$ $y = (y_j)_{(1 \le j \le P)}$ $f = (f_j)_{(1 \le j \le P)}$

- $f(x) f(x+h) = \operatorname{grad} f(x) \cdot h + o(\|h\|)$
- $dy = \operatorname{grad} f(x) \cdot dx = \frac{\partial f}{\partial x}(x) \cdot dx = \frac{\partial y}{\partial x} \cdot dx$ (locally linear)
- $dy_j = \sum_{i=1}^{i=n} \frac{\partial f_j}{\partial x_i}(x) \cdot dx_i = \sum_{i=1}^{i=n} \frac{\partial y_j}{\partial x_i} \cdot dx_i$
- x, dx, y, dy, f(x) and h are all "regular" vectors;
- $\frac{\partial y}{\partial x}$ is a matrix (Jacobian of $f: J_{ij} = \left(\frac{\partial y}{\partial x}\right)_{ij} = \frac{\partial y_j}{\partial x_i} = \frac{\partial f_j}{\partial x_i}(x)$).

Differential of a composed function vector inputs and vector outputs

•
$$f: \mathbb{R}^N \to \mathbb{R}^P : x \to y = f(x)$$

f is differentiable

•
$$g: \mathbb{R}^P \to \mathbb{R}^Q: y \to z = g(y)$$

g is differentiable

•
$$x = (x_i)_{(1 \le i \le N)}$$
 $y = (y_j)_{(1 \le j \le P)}$

$$y = (y_j)_{(1 \le j \le P)}$$

$$z = (z_k)_{(1 \le k \le Q)}$$

•
$$dz = \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial x} \cdot dx$$

$$\bullet \ \frac{\partial z}{\partial x} = \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial x}$$

(matrix multiplication: non commutative!)

- x, dx, y, dy, z, dz, f(x) and g(y) are all regular vectors;
- $\frac{\partial y}{\partial x}$, $\frac{\partial z}{\partial y}$ and $\frac{\partial z}{\partial x}$ are all matrices (f, g and gof Jacobians).

Differential of a composed function vector inputs and scalar output

- $f: \mathbb{R}^N \to \mathbb{R}^P: x \to y = f(x)$ f is differentiable
- $g: \mathbb{R}^P \to \mathbb{R}$ $: y \to z = g(y)$ g is differentiable
- $x = (x_i)_{(1 \le i \le N)}$ $y = (y_i)_{(1 \le i \le P)}$ $z \in \mathbb{R}$
- $\frac{\partial z}{\partial x} = \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial x}$ (left row vector × matrix mult. \rightarrow row vector)
- z, dz and g(y) are scalars;
- x, dx, y, dy, and f(x) are regular vectors;
- $\frac{\partial z}{\partial y}$ and $\frac{\partial z}{\partial x}$ are transpose (row) vectors (f and gof gradients);
- $\frac{\partial y}{\partial x}$ is a matrix (f Jacobian).