Machine Learning

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1   Gradients, Jacobian, Ferchet Drivative, and Sub-Gradients

Theorem 1.1. *Gradient*
Let $f : \mathbb{R}^n \to \mathbb{R}$ be a Differentiable function.
Then $\nabla f : \mathbb{R}^n \to \mathbb{R}^n$ where:

$$\nabla f(x) := \begin{bmatrix} \frac{\partial}{\partial x_1} f(x) \\ \vdots \\ \frac{\partial}{\partial x_n} f(x) \end{bmatrix} \forall x \in \mathbb{R}^n$$

is called the Gradient of f.

Theorem 1.2. *Jacobian*
Let $f : \mathbb{R}^n \to \mathbb{R}^m$ where:

$$f(x) = \begin{bmatrix} f_1(x) \\ \vdots \\ f_m(x) \end{bmatrix} \forall x \in \mathbb{R}^n \text{ and } (\forall j \in \mathbb{N}_m)(f_j : \mathbb{R}^n \to \mathbb{R})$$

Then $J_f : \mathbb{R}^n \to \mathbb{R}^{n \times m}$ where:

$$J_f(x) := \begin{bmatrix} \frac{\partial}{\partial x_1} f_1(x) & \cdots & \frac{\partial}{\partial x_n} f_1(x) \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1} f_m(x) & \cdots & \frac{\partial}{\partial x_n} f_m(x) \end{bmatrix}$$

Theorem 1.3. *When the Jacobian is the Gradient*
Let $f : \mathbb{R}^n \to \mathbb{R}$
Then $(\nabla f(x) = (J_f(x))^T)(\forall x \in \mathbb{R}^n)$

Theorem 1.4. *Chain Rule* *Let $f : \mathbb{R}^n \to \mathbb{R}^m$ and $g : \mathbb{R}^m \to \mathbb{R}^k$ be differentiable functions.*
Then: $J_{g \circ f}(x) = J_g(f(x)) J_f(x)$

2   Finite Composition Operator

Definition 2.1. *Finite Composition Operator*
Let the collection $X = \{X_j\}_{j=0}^n$ be a finite sequence of sets.

Further let $\{T_j\}_{j=0}^{n-1}$ be a finite sequence of operators such that $(\forall j \in \mathbb{N}_{n-1})(T_j : X_j \to X_{j+1})$
Then $T : X_0 \to X_n$ defined by:

$$T := \bigcirc_{j=0}^{n-1} T_j := \ldots$$

*is called the **Finite Composition Operator defined on X**.*

Theorem 2.1. ***Finite Composition Jacobian***
Let T be defined as above.
Then:

$$J_T(x) = \ldots$$

And so to calculate J_T we need to calculate each J_{T_j} for each j only once.
Proof on the next page:

Proof:

Case: $n = 1$

$$X = \{X_0, X_1\}$$
$$T_0 : X_0 \to X_1$$
$$T^1(x) = T_0(x)$$
$$J_{T^1}(x) = J_{T_0}(x)$$

Case: $n = 2$

$$X = \{X_0, X_1, X_2\}$$
$$T_0 : X_0 \to X_1$$
$$T_1 : X_1 \to X_2$$
$$T^2(x) = (T_1 \circ T_0)(x)$$
$$J_{T^2}(x) = J_{T_1 \circ T_0}(x) = J_{T_1}(T_0(x))J_{T_0}(x) = (J_{T_1} \circ T_0)(x) * J_{T_0}(x)$$

Case: $n = 3$

$$X = \{X_0, X_1, X_2, X_3\}$$
$$T_0 : X_0 \to X_1$$
$$T_1 : X_1 \to X_2$$
$$T_2 : X_2 \to X_3$$
$$T^3(x) = (T_2 \circ T_1 \circ T_0)(x) = (T_2 \circ T^2)(x)$$
$$J_{T^3}(x) = J_{T_2 \circ T^2}(x) = J_{T_2}(T^2(x))J_{T^2}(x) = (J_{T_2} \circ T_1 \circ T_0)(x) * J_{T^2}(x) = (J_{T_2} \circ T_1 \circ T_0)(x) * ((J_{T_1} \circ T_0)(x) * J_{T_0}(x))$$

Case: $n = k$

$$X = \{X_0, ..., X_k\}$$
$$T_0 : X_0 \to X_1$$
$$\vdots$$
$$T_{k-1} : X_{k-1} \to X_k$$
$$T^k(x) = (\bigcirc_{j=0}^{k-1} T_j)(x) = (T_{k-1} \circ T^{k-1})(x)$$
$$J_{T^k}(x) = J_{T_{k-1} \circ T^{k-1}}(x) = J_{T_{k-1}}(T^{k-1}(x))J_{T^{k-1}}(x)$$

Theorem 2.2. ***Examples***:
Let $T : \mathbb{R}^n \to \mathbb{R}^n$ such that:

$$T(x) := \begin{bmatrix} \arctan(\pi_1(x)) \\ \vdots \\ \arctan(\pi_n(x)) \end{bmatrix} = \begin{bmatrix} \arctan(x_1) \\ \vdots \\ \arctan(x_n) \end{bmatrix}$$

Then

$$J_T(x) = \begin{bmatrix} \frac{\partial}{\partial x_1} T_1(x) & \cdots & \frac{\partial}{\partial x_n} T_1(x) \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1} T_n(x) & \cdots & \frac{\partial}{\partial x_n} T_n(x) \end{bmatrix} = \begin{bmatrix} \frac{\partial}{\partial x_1} \arctan(x_1) & \cdots & \frac{\partial}{\partial x_n} \arctan(x_1) \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1} \arctan(x_n) & \cdots & \frac{\partial}{\partial x_n} \arctan(x_n) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{\partial}{\partial x_1} \arctan(x_1) & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \frac{\partial}{\partial x_n} \arctan(x_n) \end{bmatrix} = \begin{bmatrix} \frac{1}{x_1^2+1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \frac{1}{x_n^2+1} \end{bmatrix} = I \begin{bmatrix} \frac{1}{x_1^2+1} \\ \vdots \\ \frac{1}{x_n^2+1} \end{bmatrix}$$

Theorem 2.3. ***Examples***:
Let $T_b : \mathbb{R}^n \to \mathbb{R}^n$ such that:

$$T_b(x) := \begin{bmatrix} \pi_1(x) + b_1 \\ \vdots \\ \pi_n(x) + b_n \end{bmatrix} = \begin{bmatrix} x_1 + b_1 \\ \vdots \\ x_n + b_n \end{bmatrix}$$

Then

$$J_{T_b}(x) = \begin{bmatrix} \frac{\partial}{\partial x_1}[x_1 + b_1] & \cdots & \frac{\partial}{\partial x_n}[x_1 + b_1] \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1}[x_1 + b_1] & \cdots & \frac{\partial}{\partial x_n}[x_n + b_n] \end{bmatrix} = \begin{bmatrix} x_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & x_n \end{bmatrix} = Ix$$

Theorem 2.4. ***Examples***:
Let $T_A : \mathbb{R}^n \to \mathbb{R}^m$ such that:

$$T_A(x) := \begin{bmatrix} \langle a_1, x \rangle \\ \vdots \\ \langle a_m, x \rangle \end{bmatrix} = \begin{bmatrix} \sum_{j=1}^n a_{1,j} x_j \\ \vdots \\ \sum_{j=1}^n a_{m,j} x_j \end{bmatrix} = \begin{bmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \ddots & \vdots \\ a_{m,1} & \cdots & a_{m,n} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = Ax$$

Then:

$$J_{T_A}(x) = \begin{bmatrix} \frac{\partial}{\partial x_1}\left[\sum_{j=1}^n a_{1,j} x_j\right] & \cdots & \frac{\partial}{\partial x_n}\left[\sum_{j=1}^n a_{1,j} x_j\right] \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1}\left[\sum_{j=1}^n a_{m,j} x_j\right] & \cdots & \frac{\partial}{\partial x_n}\left[\sum_{j=1}^n a_{m,j} x_j\right] \end{bmatrix} = \begin{bmatrix} a_{1,1} x_1 & \cdots & a_{1,n} x_n \\ \vdots & \ddots & \vdots \\ a_{m,1} x_1 & \cdots & a_{m,n} x_n \end{bmatrix}$$

3   Surjective Continuous Non-decreasing Bounded Functionals

Let $a, b \in \mathbb{R}$ where $a < 0 < b$

Let $B = \{f : [a, b] \to [0, 1] \mid f$ is surjective, continuous, and non-decreasing.$\}$

Theorem 3.1. *B is convex.*

Let $f, g \in B$

$h(x) := \lambda f(x) + (1 - \lambda)g(x)$ where $\lambda \in [0, 1]$

Then h is still continuous since the linear combination of continuous function is continuous.

Since both f and g are surjective and non-decreasing, then $f(a) = g(a) = 0$ and $f(b) = g(b) = 1$

Theorem 3.2. *B is translation invariant.*

Theorem 3.3. *B is not complete.*

Theorem 3.4. *Every element in B can be decomposed as a finite non-trivial convex combination from B*