

Linear Algebra Review

$$T: \mathbb{R}^n \longrightarrow \mathbb{R}^m$$

$$\vec{x} \in \mathbb{R}^n$$

$$\vec{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

$$T \text{ linear means } T(\alpha \vec{x} + \beta \vec{y}) = \alpha T\vec{x} + \beta T\vec{y} \quad \forall \alpha, \beta \in \mathbb{R}, \forall \vec{x}, \vec{y}$$

For T linear transformation associate

$$\text{matrix } \{T_{ij}\}_{i=1}^m \{j=1}^n$$

$$\text{ie) } \begin{pmatrix} T_{1,1} & T_{1,2} & \dots & T_{1,n} \\ T_{2,1} & & & \\ \vdots & & & \\ T_{m,1} & & & T_{m,n} \end{pmatrix}$$

Formally,

T linear Transformation

$[T]$ its matrix

However, Linear Transformation and its matrix used interchangeably, $\boxed{T\vec{x}}$

Remark: $T_{i,j}$ in terms of linear Transformation

$$\text{Standard basis } \vec{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \end{pmatrix} \quad \vec{e}_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \end{pmatrix} \quad \vec{e}_n = \begin{pmatrix} 0 \\ \vdots \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = x_1 \vec{e}_1 + \dots + x_n \vec{e}_n$$

and $T_{i,j} = \text{coordinate \#} i \text{ of } T\vec{e}_j$

Also say $T\vec{e}_j = \text{column \#} j \text{ of matrix of } T$

Norm

$$|\vec{x}| = \|\vec{x}\| = (x_1^2 + x_2^2 + \dots + x_n^2)^{1/2}$$

$$1) \|\vec{x}\| \geq 0$$

"non-negative"

$$2) \|\vec{x}\| = 0 \text{ iff } \vec{x} = \vec{0}$$

"non-degenerate"

$$3) \|\alpha \vec{x}\| = |\alpha| \cdot \|\vec{x}\|$$

$$4) \|\vec{x} + \vec{y}\| \leq \|\vec{x}\| + \|\vec{y}\| \quad \text{"triangle inequality"}$$

Any function $\vec{x} \rightarrow \|\vec{x}\| \in \mathbb{R}$ sat. 1-4 is called a norm.

Example

$$\|\vec{x}\|_p = \left(\sum_{k=1}^n |x_k|^p \right)^{1/p}$$

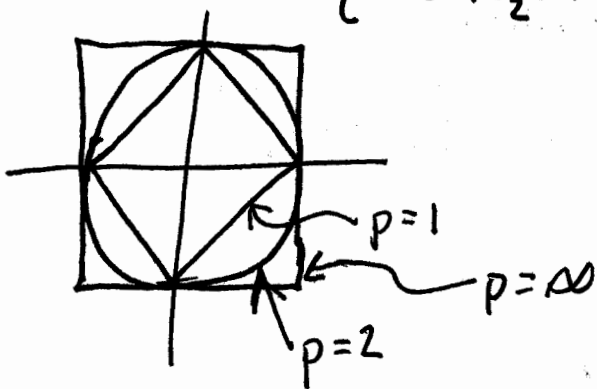
should be $|x_k|^p$?

$$1 \leq p < \infty$$

$$\|\vec{x}\|_\infty = \max_k |x_k|$$

Picture $\mathbb{R}^2$

$$\{\vec{x} \in \mathbb{R}_2 : \|\vec{x}\| \leq 1\}$$



Ball under different norms
 $p=2$ is "euclidean norm"

All these norms $\|\cdot\|_p$ are equivalent.

2 Norms are called equivalent if $\exists C < \infty$ st

$$\frac{1}{C} \|\vec{x}\| \leq \|\vec{x}\|_* \leq C \|\vec{x}\|$$

Note: $C=n$ for p -norms in $\mathbb{R}^n$

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if we have a norm we can
define $\rho(\vec{x}, \vec{y}) = \|\vec{x} - \vec{y}\|$ which gives us a topology.

Observation: Equivalent Norms give identical Topology. (Trivial)

If we change norm $\|\cdot\|$ to an equivalent $\|\cdot\|_*$ the differentiability & derivative does not change.

PF $f(\vec{x} + \vec{h}) - f(\vec{x}) - L\vec{h} =: r(\vec{h})$

$\lim_{h \rightarrow 0} \frac{r(\vec{h})}{\|\vec{h}\|} = 0$ but $h \rightarrow 0$ is equivalent to $\|\vec{h}\| \rightarrow 0$
so,

now look at $\lim_{\|\vec{h}\|_* \rightarrow 0} \frac{r(\vec{h})}{\|\vec{h}\|_*}$

$\|\vec{h}\|$
not $\|\vec{h}\|_*$

$\Rightarrow \lim_{\|\vec{h}\| \rightarrow 0} \left(\frac{r(\vec{h})}{\|\vec{h}\|} \right) \cdot \left(\frac{\|\vec{h}\|}{\|\vec{h}\|_*} \right)$
 $\downarrow$ $\downarrow$
 0 $\leq C$

$= 0$

bounded $\cdot (\lim=0) = (\lim=0)$

$O(1) \cdot o(1) = o(1)$

Thus, to reiterate, differentiability & derivative does not change

Thm: Any 2 norms on $\mathbb{R}^n$ are equivalent ("Very Advanced")

Compute $df(x)$

$$f: \underbrace{\Omega}_{\substack{\cap \\ \mathbb{R}^n}} \rightarrow \mathbb{R}^m \quad \text{open}$$

$$x \in \Omega \quad f = \begin{pmatrix} f_1 \\ f_2 \\ \vdots \\ f_m \end{pmatrix} \quad f_k: \Omega \rightarrow \mathbb{R}$$

$$f(\vec{x} + \vec{h}) = f(\vec{x}) + L \cdot \vec{h} + o(h) \quad \leftarrow \text{By Definition}$$

$$\boxed{\lim_{\vec{h} \rightarrow 0} r(\vec{h}) = o(\vec{h})}$$

$$\vec{h} = h \cdot \vec{e}_j$$

$$f_i(\vec{x} + h\vec{e}_j) = f_i(\vec{x}) + (L h \vec{e}_j)_i + r_i(h\vec{e}_j)$$

$$\frac{f_i(\vec{x} + h\vec{e}_j) - f_i(\vec{x})}{h} = \frac{(L h \vec{e}_j)_i}{h} + \frac{r_i(h\vec{e}_j)}{h}$$

$$\frac{\partial f_i}{\partial x_j} = (L \vec{e}_j)_i + 0$$

$$\boxed{\frac{\partial f_i}{\partial x_j} = L_{ij}}$$

$$0 \leq \frac{|r_i(h\vec{e}_j)|}{h} \leq \frac{\|r(h\vec{e}_j)\|}{\|h\vec{e}_j\|} \xrightarrow{h \rightarrow 0} 0$$

As a result, Matrix of $df(x) = \left\{ \frac{\partial f_i}{\partial x_j} \right\}_{i=1}^m \left\{ j=1}^n$
 called Jacobi matrix