

Lecture 3 - Maps from \mathbb{R}^n to \mathbb{R}^m

- Read section 6.1
- Screen flickering issue on podcast
- Course Piazza for discussing homework (clarification) / course material
- See weekly schedule on page 2 of the course schedule file
- My OH today from 9 am to 10 am (see Zoom LTI PRO on Canvas for link)

• A function $f: A \rightarrow B$ is a map from a set A to a set B , assigning to each element x in A (denote $x \in A$) an element $f(x) \in B$.

• We call $f(A) = \{f(x) : x \in A\}$ the image of f . $f(A) \subseteq B$

ex, $f: \mathbb{R} \rightarrow \mathbb{R}$, $f(x) = e^x$. Image $\{e^x : x \in \mathbb{R}\} = (0, \infty) \subsetneq \mathbb{R}$.

In this course, A is usually a subset of \mathbb{R}^n ($n=1,2,3$) and B is usually a subset of \mathbb{R}^m ($m=1,2,3$)

ex, $f: \mathbb{R}^n \rightarrow \mathbb{R}$ real-valued/scalar function

$\vec{F}: \mathbb{R}^n \rightarrow \mathbb{R}^n$ vector field

$\vec{c}: \mathbb{R} \rightarrow \mathbb{R}^n$ curve

$\Phi: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ surfaces

Properties

Continuity: A function $f: A \rightarrow B$ is continuous at $x_0 \in A$ if: $\lim_{x \rightarrow x_0} f(x) = f(x_0)$.

\mathbb{R}^n \cup \mathbb{R}^m
 \mathbb{R}^n \cup \mathbb{R}^m

$$\lim_{x \rightarrow x_0} f(x) = f(x_0).$$

Differentiability: A function $f: A \rightarrow B$,

$f = (f_1, f_2, \dots, f_{m-1}, f_m)$, is differentiable at $\vec{x}_0 \in A$ if:

$\frac{\partial f_j}{\partial x_i}$ exists at \vec{x}_0 for $j = 1, \dots, m$
 $i = 1, \dots, n$.

We define its derivative matrix at $\vec{x}_0 \in A$

$$Df(\vec{x}_0) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} \Big|_{\vec{x}_0} & \dots & \frac{\partial f_1}{\partial x_n} \Big|_{\vec{x}_0} \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1} \Big|_{\vec{x}_0} & \dots & \frac{\partial f_m}{\partial x_n} \Big|_{\vec{x}_0} \end{pmatrix}$$

$n=2, m=1$
 $Df(\vec{x}_0)$

This is an $m \times n$ matrix,

$Df(\vec{x}_0)$ is a linear transformation $\mathbb{R}^n \rightarrow \mathbb{R}^m$.

$$[Df(\vec{x}_0)]_{ij} = \frac{\partial f_i}{\partial x_j}(\vec{x}_0), \quad \begin{matrix} i=1, \dots, m \\ j=1, \dots, n. \end{matrix}$$

ex/ $f: \mathbb{R}^2 \rightarrow \mathbb{R}$
 $f(x, y) = x$ not invertible "lose information"

Def:

Say $f: A \rightarrow B$ is injective (or one-to-one) if:

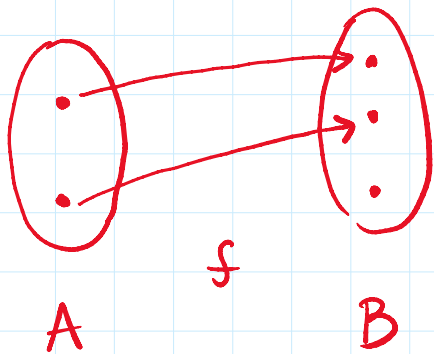
rec.

Say $f: A \rightarrow B$ is injective (or one-to-one) if:

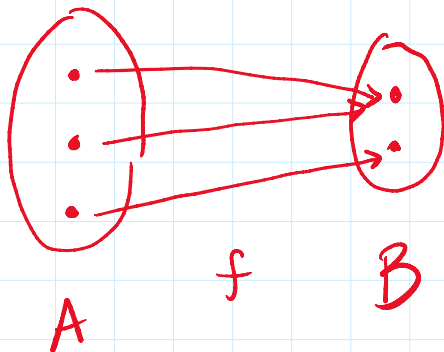
For every $u, v \in A$ such that $u \neq v$, then $f(u) \neq f(v)$

(Equivalently, $f(u) = f(v) \Rightarrow u = v$)

implies



injective



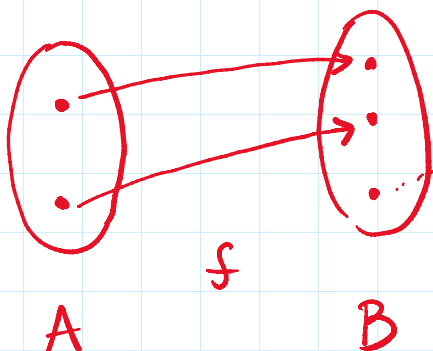
not
injective

Def:

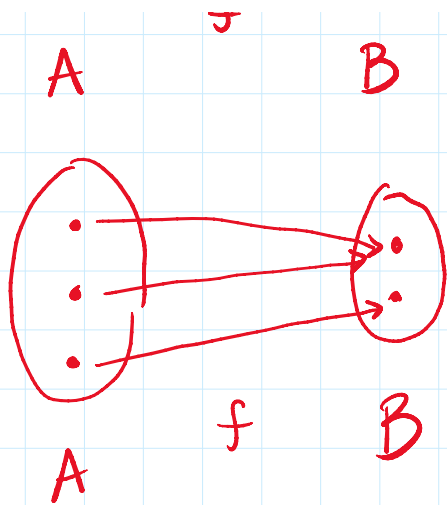
A function $f: A \rightarrow B$ is surjective (or onto)

if: the image equals codomain, $f(A) = B$.

For every $y \in B$, there exists $x \in A$ such that $f(x) = y$.



not surjective



surjective

(aside: domain/range are part of the definition of a function
 $f(x) = e^x$, $f: \mathbb{R} \rightarrow \mathbb{R}$ not surj.
 $f: \mathbb{R} \rightarrow (0, \infty)$ surj.

Say a function is bijective if it is both injective & surjective.

Theorem (Invertibility)

If $f: A \rightarrow B$ is bijective, then there exists

$f^{-1}: B \rightarrow A$ s.t. ^{such that}

$$\left. \begin{aligned} f^{-1}(f(x)) &= x \quad \text{for all } x \in A, \\ f(f^{-1}(y)) &= y \quad \text{for all } y \in B. \end{aligned} \right\} (*)$$

Proof:

Injectivity: there is at most one solution $x \in A$ to the equation $f(x) = y$, for $y \in B$.

Surjectivity: there is at least one solution $x \in A$ to the eqn $f(x) = y$, for $y \in B$.

→ u

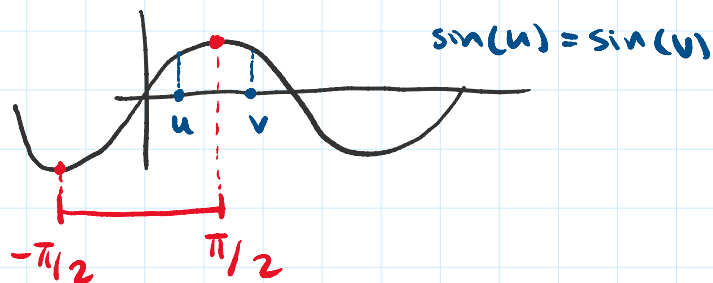
⇒ there is exactly one solution $x \in A$
to $f(x) = y$ for each $y \in B$.

We can define $f^{-1}(y) = x$.

Check (*)

□

ex/ $\sin(x)$ $\sin: \mathbb{R} \rightarrow [-1, 1]$



$\sin: [-\pi/2, \pi/2] \rightarrow [-1, 1]$ is a bijection.
⇒ can define \sin^{-1} or $\arcsin: [-1, 1] \rightarrow [-\pi/2, \pi/2]$.

ex/ Linear Transformations

• Let T be a linear transf. $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$,
viewed as an $m \times n$ matrix

$$T = \begin{pmatrix} T_{11} & \dots & T_{1n} \\ \vdots & \ddots & \vdots \\ T_{m1} & \dots & T_{mn} \end{pmatrix}$$

• Recall two of the fundamental subspaces

$$\text{Ker}(T) = \left\{ x \in \mathbb{R}^n : T(x) = 0 \right\} \leftarrow \text{injectivity}$$

$$\text{Ker}(T) = \{ x \in \mathbb{R}^n : T(x) = 0 \}$$

$$\text{col}(T) = \text{span} \left\{ \begin{pmatrix} T_{11} \\ \vdots \\ T_{m1} \end{pmatrix}, \dots, \begin{pmatrix} T_{1n} \\ \vdots \\ T_{mn} \end{pmatrix} \right\}$$

claim: T is injective \iff $\text{Ker}(T) = \{0\}$.

proof:

(\Rightarrow) Assume T is injective,
 $T(x) = T(y) \Rightarrow x = y$.
Want to show $\text{Ker}(T) = \{0\}$.

Let $x \in \text{Ker}(T)$. \mathbb{R}^m \mathbb{R}^n
 $T(x) = 0 = T(0) \Rightarrow x = 0$.
 $\Rightarrow \text{Ker}(T) = \{0\}$.

(\Leftarrow) $\text{Ker}(T) = \{0\}$.

Assume $T(x) = T(y)$

$$\Rightarrow T(x-y) = 0$$

$$\Rightarrow x-y \in \text{Ker}(T) = \{0\}$$

$$\Rightarrow x-y=0 \Rightarrow x=y \Rightarrow T \text{ injective} \quad \square$$

claim: T is surjective $\iff \text{col}(T) = \mathbb{R}^m$

proof: Observe $\text{col}(T) = T(\mathbb{R}^n)$ \square

• consider square $n=m$ $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ linear transf.

• Consider square $n \times n$, $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ linear transf.,

Theorem: A square matrix is injective \Leftrightarrow it's surjective.

Proof:

$$\begin{array}{c} \text{Rank-Nullity Theorem} \\ \text{rank} \qquad \qquad \text{nullity} \\ \dim(\text{Col}(T)) + \dim(\text{Ker}(T)) = n \\ \underbrace{\hspace{10em}} \\ = n \qquad \qquad \qquad = 0 \end{array}$$

Recall: a square matrix T is invertible when $\det(T) \neq 0$.