

(14)  
9/14/2016 Identities & inverse matrices

DEFIN:  $I_n := \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & & 0 \\ 0 & 0 & 1 & & 0 \\ \vdots & & & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 1 \end{bmatrix}$  is the  $(n \times n)$  identity matrix,  
representing  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$   
 $v \mapsto v$   
the identity map

and  $AI_n = A \quad \forall n \times n A$   
 $I_n B = B \quad \forall n \times p B$

DEFIN: If  $A$  is  $m \times n$  and  $AB = I_m$  then  $A$  is called a left-inverse for  $B$   
 $B$  is  $n \times m$  and  $BA$  called a right-inverse for  $A$

EXAMPLE:  $\underbrace{\begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}}_A \underbrace{\begin{bmatrix} \frac{1}{2} & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}}_B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2$

So  $A$  has  $B$  as a right-inverse, but  $A$  has no left-inverse  
 $B$  has  $A$  as a left-inverse, i.e. a  $3 \times 2 C$  with  $CA = I_3$  (Why?)

and  $B$  has no right inverse

(They're also not unique! e.g.  $\begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & 1 \\ x & y \end{bmatrix} = I_2$  for any  $x, y \in \mathbb{R}$ )

DEFIN: An  $n \times n$  matrix  $A$  is invertible if  $\exists$  some  $n \times n$  matrix  $B$

which is both a left and right inverse for  $A$ , i.e.  $AB = I_n$  and  $BA = I_n$ .  
Then we say  $B = A^{-1}$ .

(Wow, awesome!)

Q: Is it possible that  $A$  has a left-inverse  $C$  with  $C \neq B$ ?  
and a right-inverse  $B$  but  $AB = I_m$  and  $CA = I_n$ ?

No, associativity prevents this: (PROP 1.2.4)

In this situation, calculate  $CAB$  two ways:

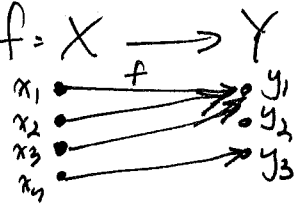
$$B = I_n B = (CA)B = C(AB) = CI_m = C$$

i.e. it forces  $B=C$

(15)

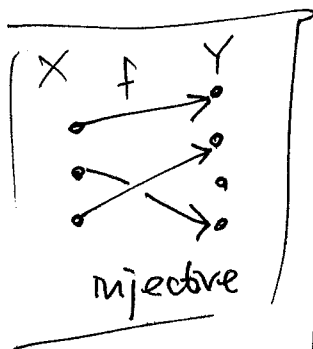
It's not clear yet ~~whether~~ whether  $A$  invertible forces  $A$  to be square (it does, we'll see later).

This is reminiscent of what happens for functions  $f: X \rightarrow Y$  between any sets  $X, Y$

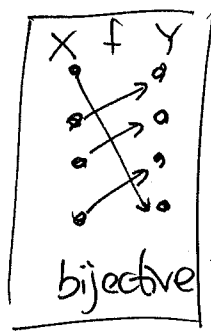
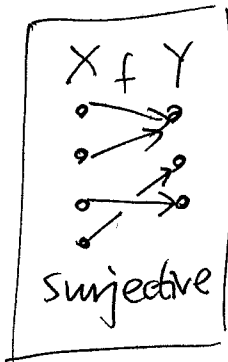


DEFIN:  $f: X \rightarrow Y$  is one-to-one if  $f(x_1) = f(x_2)$  implies  $x_1 = x_2$  (injective)

$f: X \rightarrow Y$  is onto if  $\forall y \in Y \exists x \in X$  with  $f(x) = y$  (surjective)



$f: X \rightarrow Y$  is bijective if it's both injective & surjective



It's easy to see that  $f: X \rightarrow Y$  is injective  $\iff$   $f$  has a left-inverse

i.e.  $g: Y \rightarrow X$  such that  $g \circ f = 1_X$

i.e.  $g(f(x)) = x$   $\forall x \in X$    
 the identity map on  $X$

~~But~~ (But  $g$  is far from unique!)

$f: X \rightarrow Y$  is surjective  $\iff$   $f$  has a right-inverse

i.e.  $g: Y \rightarrow X$  such that  $f \circ g = 1_Y$

i.e.  $f(g(y)) = y$   $\forall y \in Y$

(But  $g$  is far from unique!)

and  $f: X \rightarrow Y$  is bijective  $\iff$   $f$  has a left and right inverse

$g: Y \rightarrow X$  with  $g \circ f = 1_X$

(and then  $g = f^{-1}$  is unique!)  $f \circ g = 1_Y$

(16) Prop 1.3.14: A linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$

that is bijjective will have  $T^{-1}: \mathbb{R}^m \rightarrow \mathbb{R}^n$  also linear

and  $[T^{-1}], [T]$  are inverse matrices

proof: For the 1st part we need to check  $\forall \vec{v}, \vec{w} \in \mathbb{R}^m$  and  $c \in \mathbb{R}$

that  $T^{-1}(\vec{v} + \vec{w}) = T^{-1}(\vec{v}) + T^{-1}(\vec{w})$

$T^{-1}(c\vec{w}) = c T^{-1}(\vec{w})$

But note

- $T(T^{-1}(\vec{v}) + T^{-1}(\vec{w})) \stackrel{\text{linearity of } T}{=} T(T^{-1}(\vec{v})) + T(T^{-1}(\vec{w})) = \vec{v} + \vec{w} = T(T^{-1}(\vec{v} + \vec{w}))$
- $T(c T^{-1}(\vec{w})) \stackrel{\text{linearity of } T}{=} c T(T^{-1}(\vec{w})) = c \vec{w} = T(T^{-1}(c\vec{w}))$

so the fact that  $T$  is injective forces  $T^{-1}(\vec{v}) + T^{-1}(\vec{w}) = T^{-1}(\vec{v} + \vec{w})$   
 $c T^{-1}(\vec{w}) = T^{-1}(c\vec{w})$ .

Hence  $T^{-1}$  is linear. But then  $T^{-1} \circ T = 1_{\mathbb{R}^n}$ ,  $T \circ T^{-1} = 1_{\mathbb{R}^m}$

$\Rightarrow [T^{-1}][T] = [1_{\mathbb{R}^n}] = I_n$ ,  $[T][T^{-1}] = [1_{\mathbb{R}^m}] = I_m$

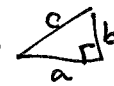
that is,  $[T], [T^{-1}]$  are inverse matrices  $\square$

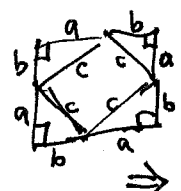
### §1.4 Geometry of $\mathbb{R}^n$

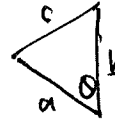
Dot products, cross products, determinants, lengths, etc...

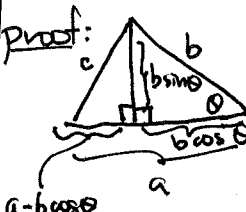
- they help us to understand distances, angles, orthogonality  
 in easy ways.

Recall 2 basic facts:

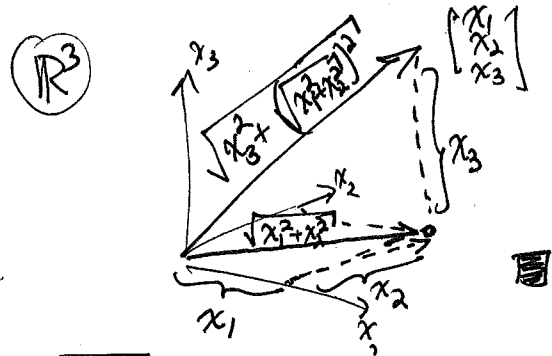
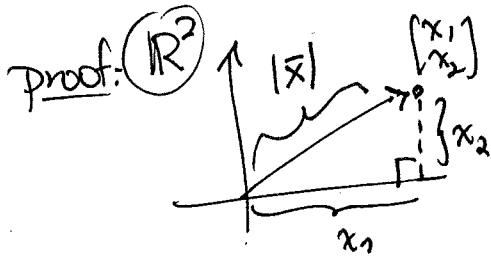
Pythagorean Theorem:   
 has  $c^2 = a^2 + b^2$

proof:   
 $(a+b)^2 = \text{area of big square}$   
 $= c^2 + 4 \cdot \frac{1}{2} ab$   
 $a^2 + 2ab + b^2 = c^2 + 2ab$   
 $a^2 + b^2 = c^2$   $\square$

Law of Cosines: More generally   
 has  $c^2 = a^2 + b^2 - 2ab \cos \theta$

proof:   
 $c^2 \stackrel{\text{Pythagoras}}{=} (a - b \cos \theta)^2 + (b \sin \theta)^2$   
 $= a^2 - 2ab \cos \theta + b^2 \cos^2 \theta + b^2 \sin^2 \theta$   
 $= a^2 + b^2 - 2ab \cos \theta$   $\square$

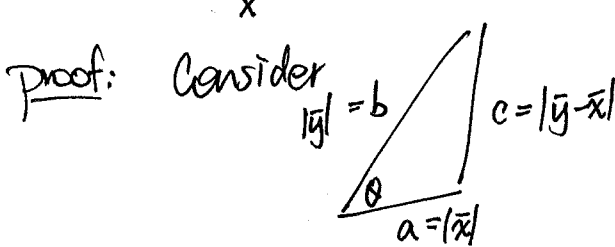
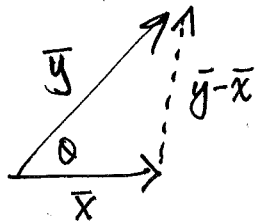
(17)  $\mathbb{R}^1$  and  
 COR: In  $\mathbb{R}^2, \mathbb{R}^3, \dots, \mathbb{R}^n$ ,  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$  has length squared =  $\vec{x} \cdot \vec{x} = |\vec{x}|^2$   
 (and as definition in  $\mathbb{R}^n$ )  $= \sum_{i=1}^n x_i^2$



Thus  $|\vec{x}| := \text{length of } \vec{x} = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2} = \sqrt{\vec{x} \cdot \vec{x}}$

COR (PROP 1.43) More generally  
 For any  $\vec{x}, \vec{y} \in \mathbb{R}^n$ , their dot product  $\vec{x} \cdot \vec{y} = |\vec{x}| \cdot |\vec{y}| \cos \theta$   
 ( $= \sum_{i=1}^n x_i y_i$ )

if  $\theta$  is the angle between them:



and law of cosines:  
 $c^2 = a^2 + b^2 - 2ab \cos \theta$

$(\vec{y} - \vec{x}) \cdot (\vec{y} - \vec{x}) = \vec{x} \cdot \vec{x} + \vec{y} \cdot \vec{y} - 2|\vec{x}||\vec{y}| \cos \theta$   
 $= \vec{y} \cdot \vec{y} - \vec{x} \cdot \vec{y} - \vec{y} \cdot \vec{x} + \vec{x} \cdot \vec{x}$

distributivity of dot product  
 (= matrix multiplication)

$\Rightarrow -2\vec{x} \cdot \vec{y} = -2|\vec{x}||\vec{y}| \cos \theta$

$\vec{x} \cdot \vec{y} = |\vec{x}||\vec{y}| \cos \theta$

9/19/2016 >

So  $\vec{x} \cdot \vec{y} = \begin{cases} 0 \Leftrightarrow \vec{x} \perp \vec{y} \text{ perpendicular/orthogonal } (\cos \theta = 0) \\ > 0 \Leftrightarrow \vec{x}, \vec{y} \text{ acute } (\text{so } \cos \theta > 0) \\ < 0 \Leftrightarrow \vec{x}, \vec{y} \text{ obtuse } (\text{so } \cos \theta < 0) \end{cases}$