

CS 726 : Recap

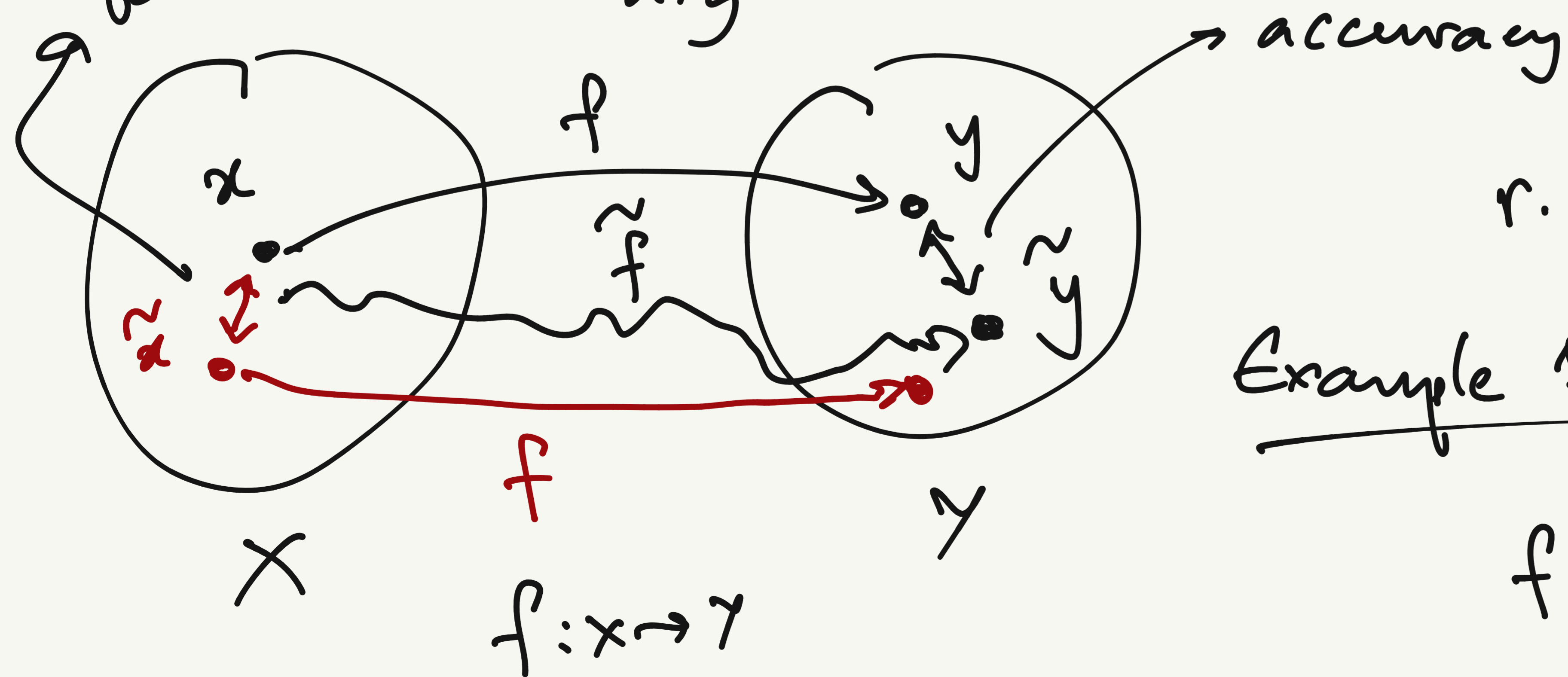
floating point numbers

$$\underline{\underline{x \in M \times \beta^E}} \rightarrow \pm (d_0 . d_1 d_2 \dots d_{n-1})$$

$$\mathbb{Z} \quad \{E_{\min}, \dots, E_{\max}\}$$

- for any $x \in \mathbb{R}$, $fl(x) = x(1 + \epsilon)$ for some $|\epsilon| \leq \epsilon_m$
- for any $x, y \in \mathbb{F}$, $x \otimes y = (x * y)(1 + \epsilon)$ for some $|\epsilon| \leq \epsilon_m$

backward stability



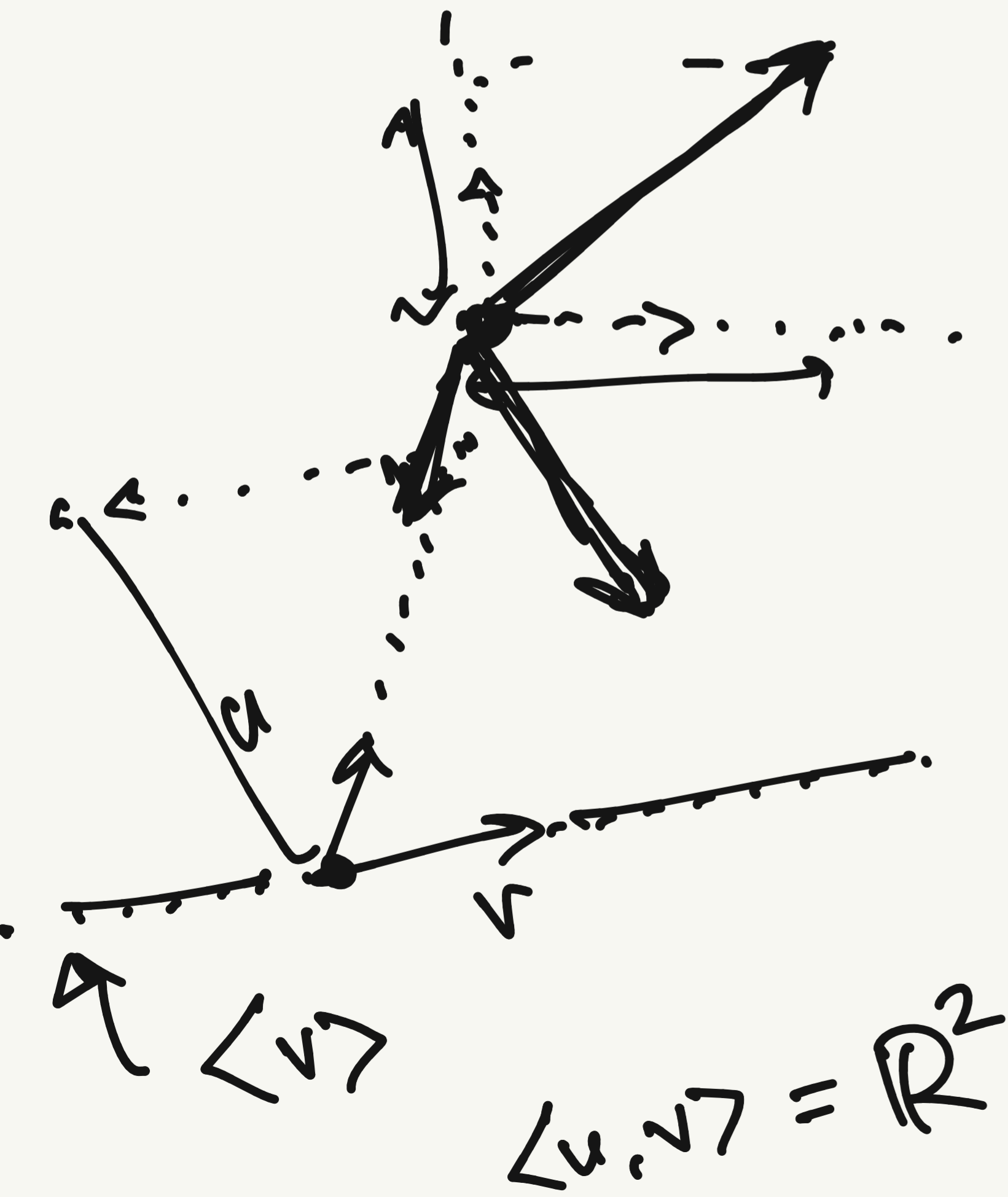
$$r.f.e. \leq (\kappa(x) + o(1)) r.b.e.$$

Example 1: $f: \mathbb{R}^n \setminus \{0\} \rightarrow \mathbb{R}^n$

$$f(\vec{x}) = \frac{\vec{x}}{\|\vec{x}\|}$$

Linear algebra

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n \text{ or } \mathbb{C}^n$$



vector space V over a field \mathbb{K} is a set with
operations $+$: $V \times V \rightarrow V$
 \cdot : $\mathbb{K} \times V \rightarrow V$ such that... $(\mathbb{R} \text{ or } \mathbb{C})$

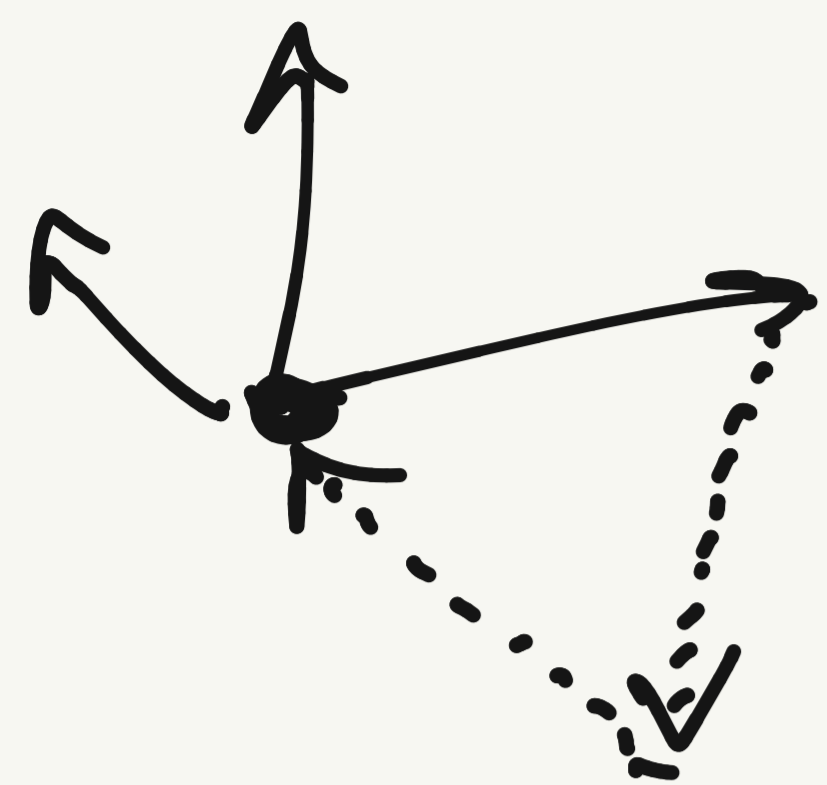
$$\{v_1, v_2, \dots, v_n\} \in V$$

$$\rightarrow v = \underbrace{s_1 v_1 + s_2 v_2 + \dots + s_n v_n}_{\text{linear combination}} \text{ for some } s_1, \dots, s_n \in \mathbb{K}$$

linear combination

Span = set of all l.c.'s = $\langle v_1, v_2, \dots, v_n \rangle \subseteq V$

$\{v_1, \dots, v_n\}$ is linearly independent if no nontrivial l.c. is zero. $s_1 v_1 + \dots + s_n v_n = \vec{0} \Leftrightarrow s_1 = \dots = s_n = 0$



$\{v_1, \dots, v_n\}$ are not lin. indep. $\Leftrightarrow \exists i$ s.t.

$$v_i = s_1 v_1 + \dots + s_n v_n$$

(not including i)

basis of V is a

linearly independent set that spans V .

$$\langle b_1, \dots, b_n \rangle = V$$

$$\beta = \{\vec{b}_1, \dots, \vec{b}_n\}$$

for any $\vec{x} \in V$, $\exists s_1, \dots, s_n$ s.t. $\vec{x} = s_1 \vec{b}_1 + \dots + s_n \vec{b}_n$
and s_1, \dots, s_n are unique

\Rightarrow any vector $\vec{x} \in V$ can be represented uniquely as

$$\begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix}$$

$$\in \mathbb{K}^n$$

Standard basis = $\left\{ \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \right\}$ coordinate vector for \vec{x}

$$\vec{e}_1 \quad \vec{e}_2 \quad \dots \quad \vec{e}_n$$

$$[\vec{x}]_\beta$$

$$\mathbb{R}^n$$

$$\mathbb{C}^n$$

$$\vec{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} + \dots = x_1 \vec{e}_1 + x_2 \vec{e}_2 + \dots + x_n \vec{e}_n$$

$[\vec{x}]_{\vec{e}} = \vec{x}$

$T: U \rightarrow V$
 $\downarrow \quad \downarrow$
 $\mathbb{K}^n \quad \mathbb{K}^m$

T is a linear transformation if
 $T(u_1 + u_2) = T(u_1) + T(u_2)$ and $T(su) = sT(u)$

$(\Rightarrow) T(s_1 u_1 + \dots + s_n u_n) = s_1 T(u_1) + \dots + s_n T(u_n)$

$\vec{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \rightarrow T(\vec{x}) = y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{pmatrix} = \begin{pmatrix} | \\ | \\ \vec{a}_1 \\ | \\ | \end{pmatrix} \dots \begin{pmatrix} | \\ | \\ \vec{a}_n \\ | \\ | \end{pmatrix} \in \mathbb{K}^m$

$A = \begin{pmatrix} | & & | \\ \vec{a}_1 & \dots & \vec{a}_n \\ | & & | \\ m & \times & n \end{pmatrix}$

$$T(\vec{x}) = T(x_1 \vec{e}_1 + \dots + x_n \vec{e}_n) = x_1 T(\vec{e}_1) + \dots + x_n T(\vec{e}_n) = x_1 \vec{a}_1 + \dots + x_n \vec{a}_n$$

$$y = T(\vec{x}) = x_1 \vec{a}_1 + \dots + x_n \vec{a}_n \quad \text{where } \vec{a}_j = T(\vec{e}_j) \quad A = \begin{bmatrix} | & & | \\ \vec{a}_1 & \dots & \vec{a}_n \\ | & & | \end{bmatrix}$$

$$= \sum x_j \vec{a}_j$$

$$y_1 = ?$$

$$y_2 = ?$$

$$\vdots$$

$$y_i = \sum x_j a_{ij} \longrightarrow \text{matrix-vector multiplication}$$

$$\underbrace{\begin{bmatrix} | \\ \vec{a}_1 \\ | \end{bmatrix}}_A \underbrace{\begin{bmatrix} | \\ \vec{a}_n \\ | \end{bmatrix}}_A \underbrace{\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}}_{\vec{x}} = \sum x_j \vec{a}_j$$

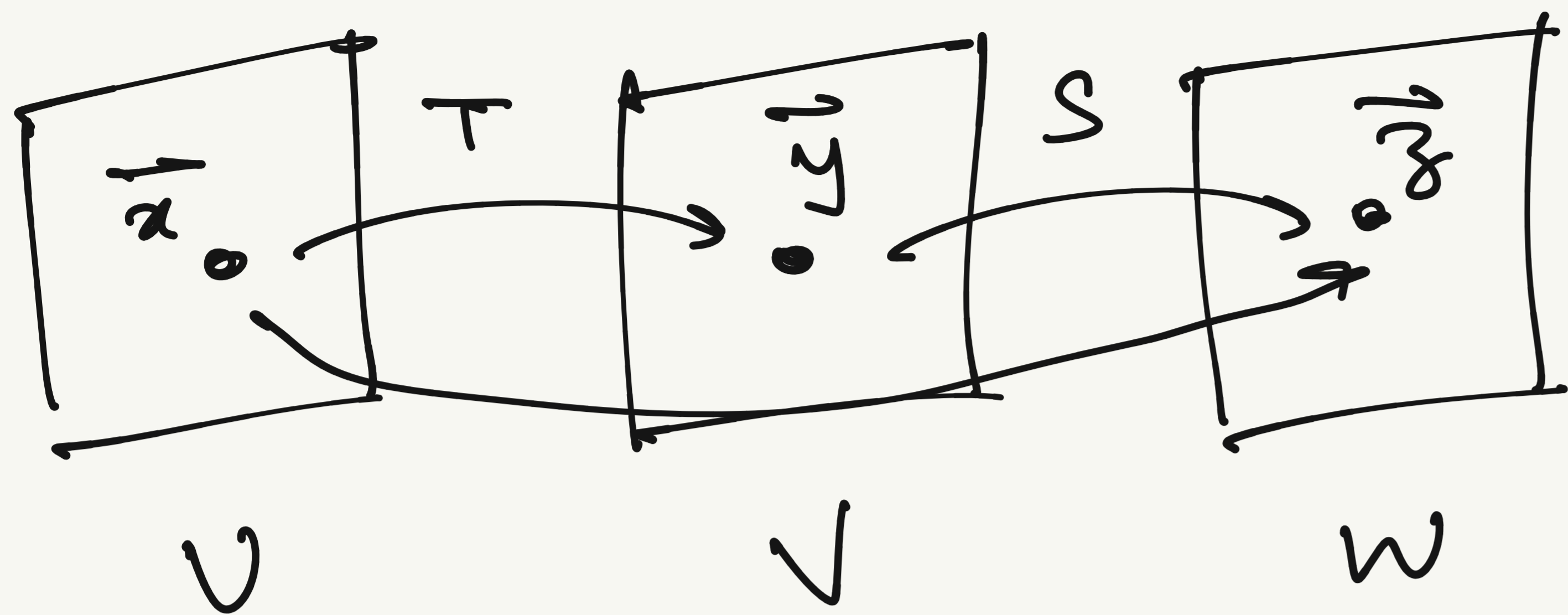
①

② j^{th} col. of $A = A\vec{e}_j = \underline{\underline{T(\vec{e}_j)}}$

$$T: \mathbb{K}^n \rightarrow \mathbb{K}^m$$

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} \in \mathbb{K}^{m \times n}$$

$$T(\vec{x}) = A\vec{x}$$



$$T(\vec{x}) = B\vec{x}$$

$$S(\vec{y}) = A\vec{y}$$

$$S \circ T : U \rightarrow W$$

$$(S \circ T)(\vec{x}) = A(B\vec{x})$$

$$C\vec{x} = (S \circ T)(\vec{x}) = A(B\vec{x})$$

$$C\vec{e}_j = A(B\vec{e}_j)$$

$$\vec{c}_j = A\vec{b}_j$$

$$\begin{bmatrix} \vec{c}_1 & \dots & \vec{c}_n \end{bmatrix}$$

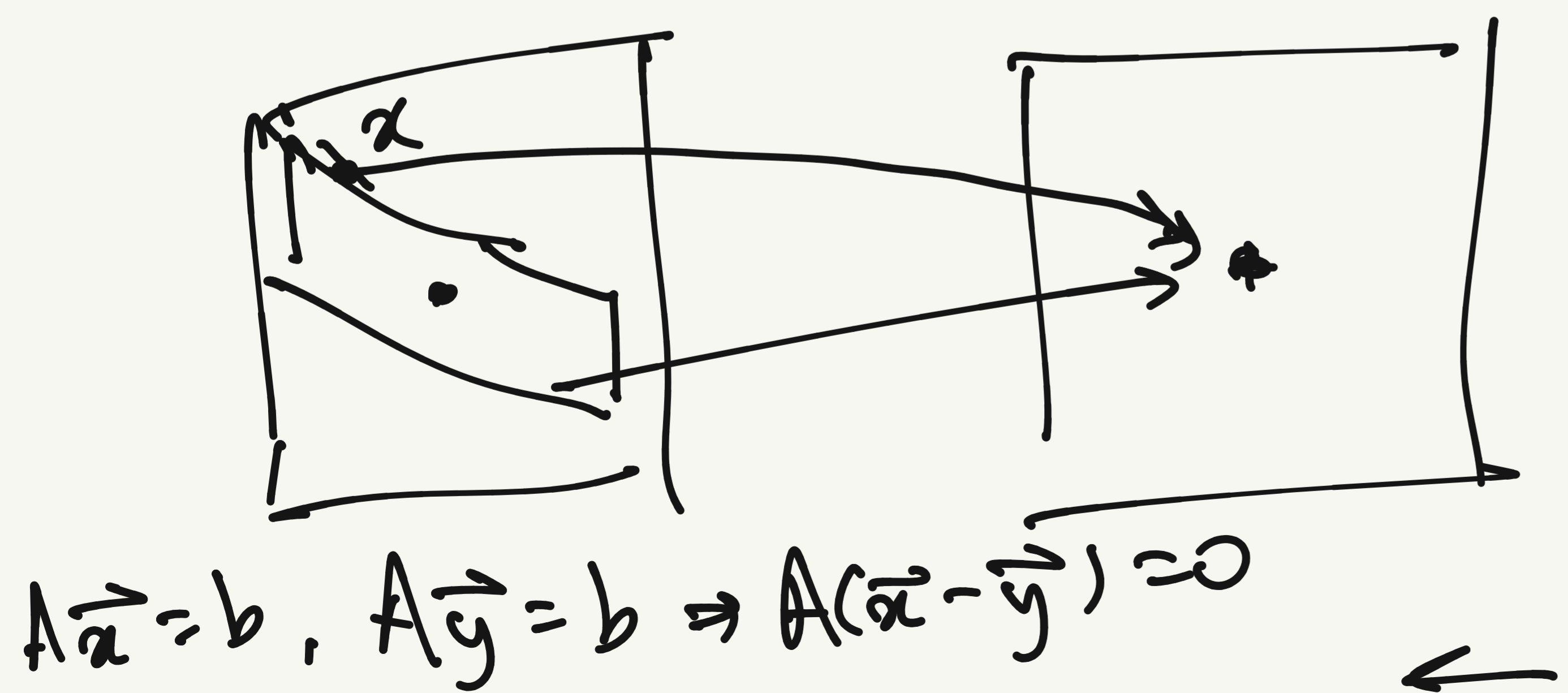
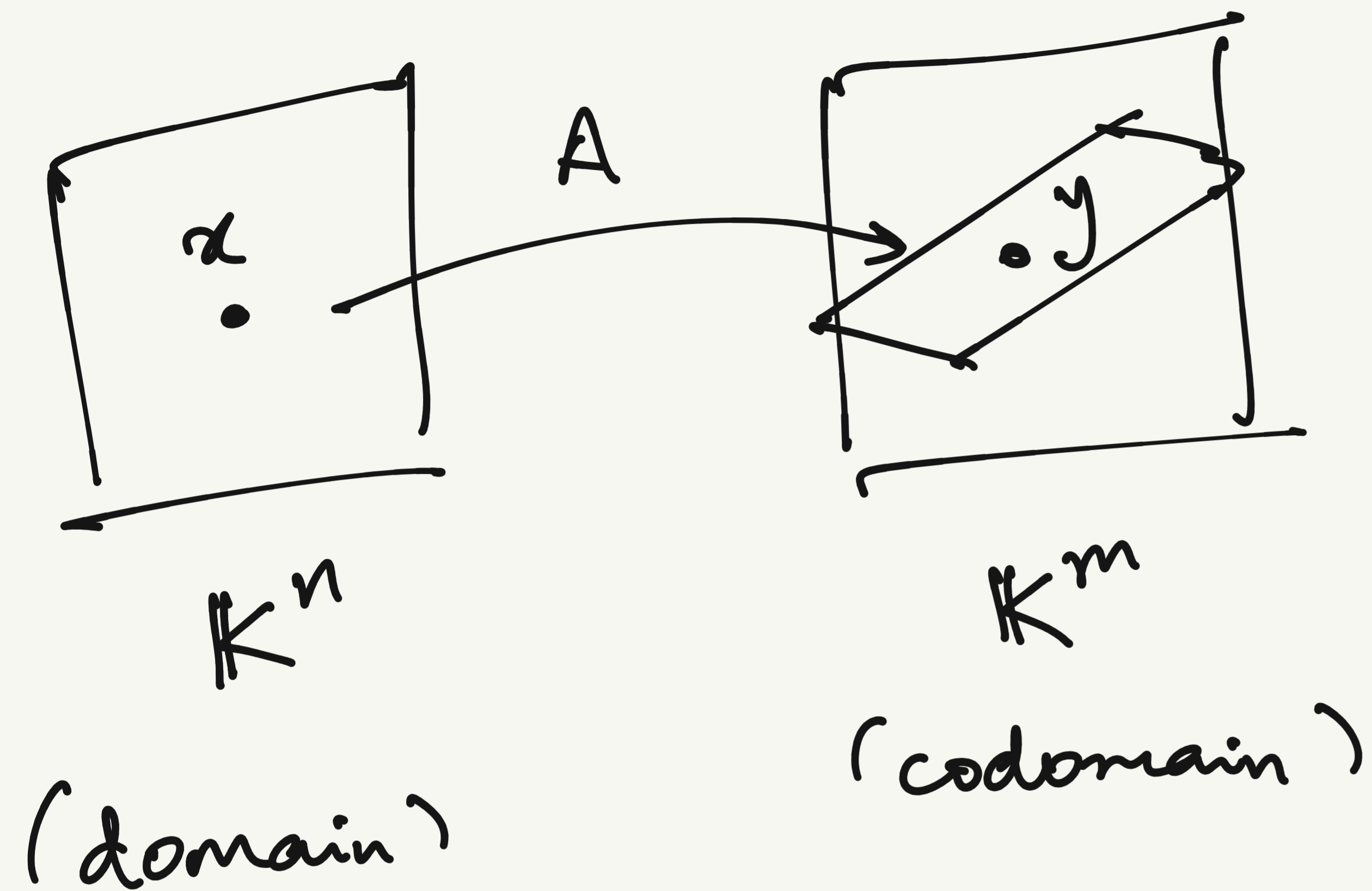
$$= \begin{bmatrix} A\vec{b}_1 & \dots & A\vec{b}_n \end{bmatrix}$$

$$C = AB$$

$$\begin{bmatrix} \vec{c}_1 \\ \vdots \\ \vec{c}_m \end{bmatrix} = \begin{bmatrix} a_{11}B \\ \vdots \\ a_{m1}B \end{bmatrix}$$

$$c_{ik} = \sum a_{ij} b_{jk}$$

$$A \in \mathbb{K}^{m \times n}$$



$$\underline{\underline{\text{range}(A)}} = \{ \underline{A\vec{x}} : \vec{x} \in \mathbb{K}^n \}$$

$$A = \begin{bmatrix} | & & | \\ \vec{a}_1 & \dots & \vec{a}_n \\ | & & | \end{bmatrix}$$

$$= \langle \vec{a}_1, \dots, \vec{a}_n \rangle$$

$$= \text{Col}(A) \subseteq \text{codomain}$$

$$\vec{a}_1 x_1 + \dots + \vec{a}_n x_n$$

$$\underline{\underline{\text{null}(A)}} = \{ \vec{x} : A\vec{x} = \vec{0} \} \subseteq \text{domain}$$

$$\text{Solve } A\vec{x} = \vec{b}$$

1. Does sol. exist? iff $\vec{b} \in \text{range}(A)$

2. If \vec{x} is a sol., what are all other sol.?

$A\vec{x} = \vec{b}$ has sol. $\Leftrightarrow \vec{b} \in \text{range}(A)$

If \vec{x}, \vec{y} are sol. $\Rightarrow \vec{x} - \vec{y} \in \text{null}(A)$

\vec{x} sol. and $\vec{z} \in \text{null}(A) \Leftrightarrow \vec{x} + \vec{z}$ is also sol.

col rank = # lin. indep. cols. of A
= $\dim(\text{range}(A))$

row rank = # lin indep. rows = col rank

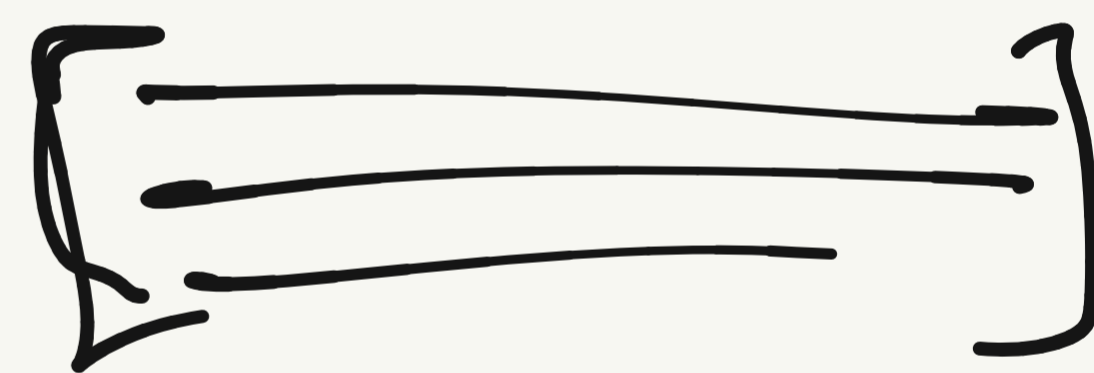
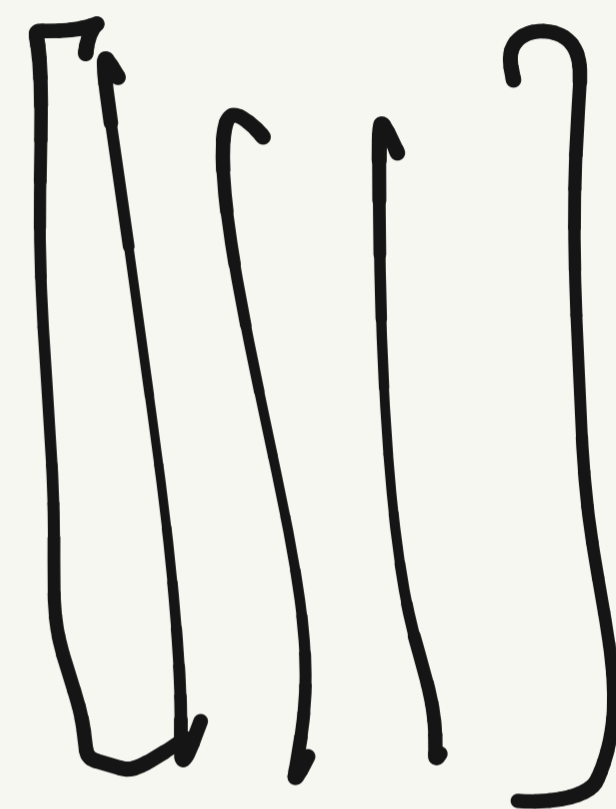
$\text{rank}(A) \leq \min(m, n)$

If $\text{rank}(A) = \min(m, n)$:

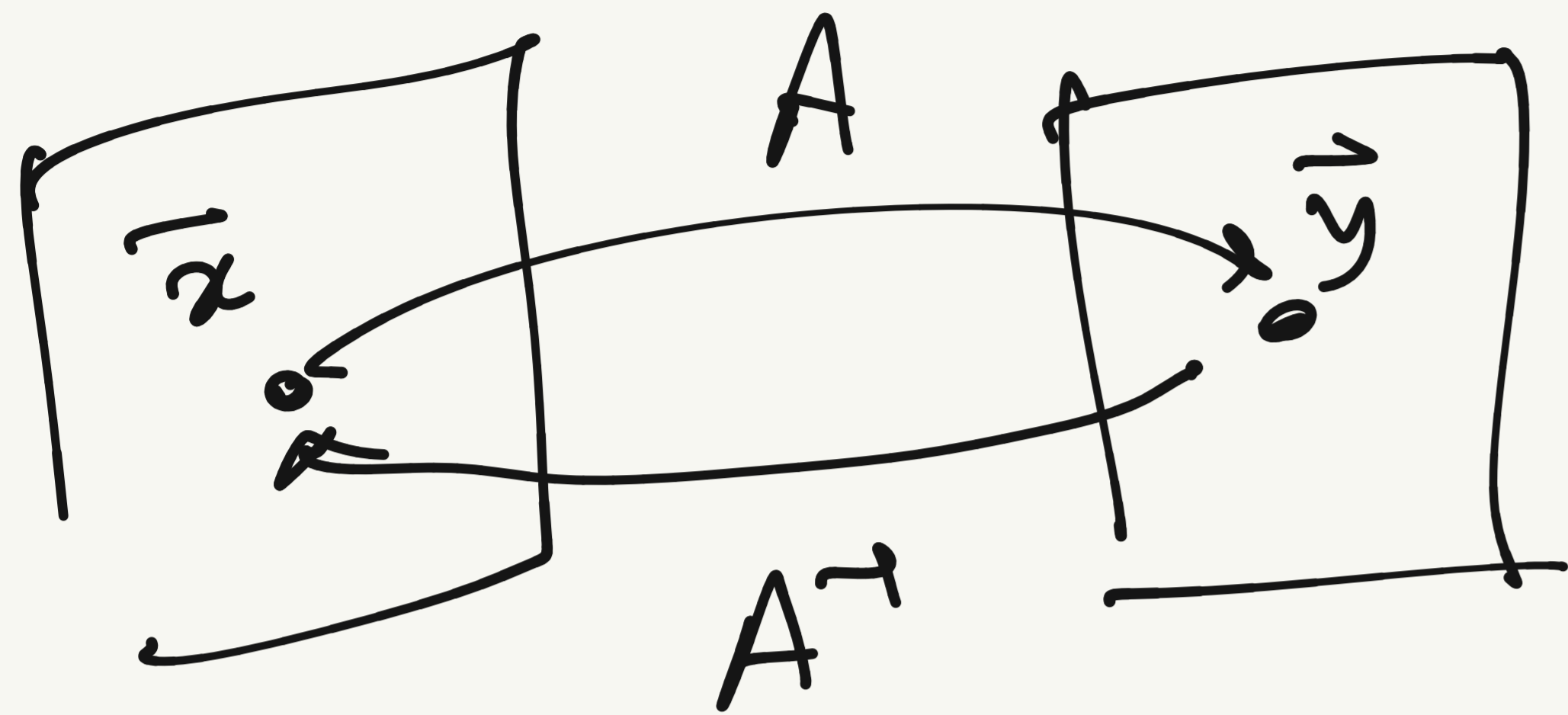
A has full rank

Thm: A "tall" matrix $A \in K^{m \times n}$, $m \geq n$ is full rank

iff. $\vec{x} \mapsto A\vec{x}$ is injective ie. $A\vec{x}_1 = A\vec{x}_2 \Rightarrow \vec{x}_1 = \vec{x}_2$.



Inverse of matrix A : $A^{-1}(A\vec{x}) = \vec{x}$ exists only if $m=n$



if A is not invertible,

A is Singular

$$T(\vec{x}) = \vec{y} = A\vec{x}$$

$$T^{-1}(\vec{y}) = \vec{x} = A^{-1}\vec{y}$$

$$A^{-1}(A\vec{x}) = \vec{x}$$

$$A(A^{-1}\vec{y}) = \vec{y}$$

if $A \in \underline{K^{m \times m}}$,

A^{-1} exists

$$\Leftrightarrow \text{rank}(A) = m$$

$$\Leftrightarrow \text{range}(A) = K^m$$

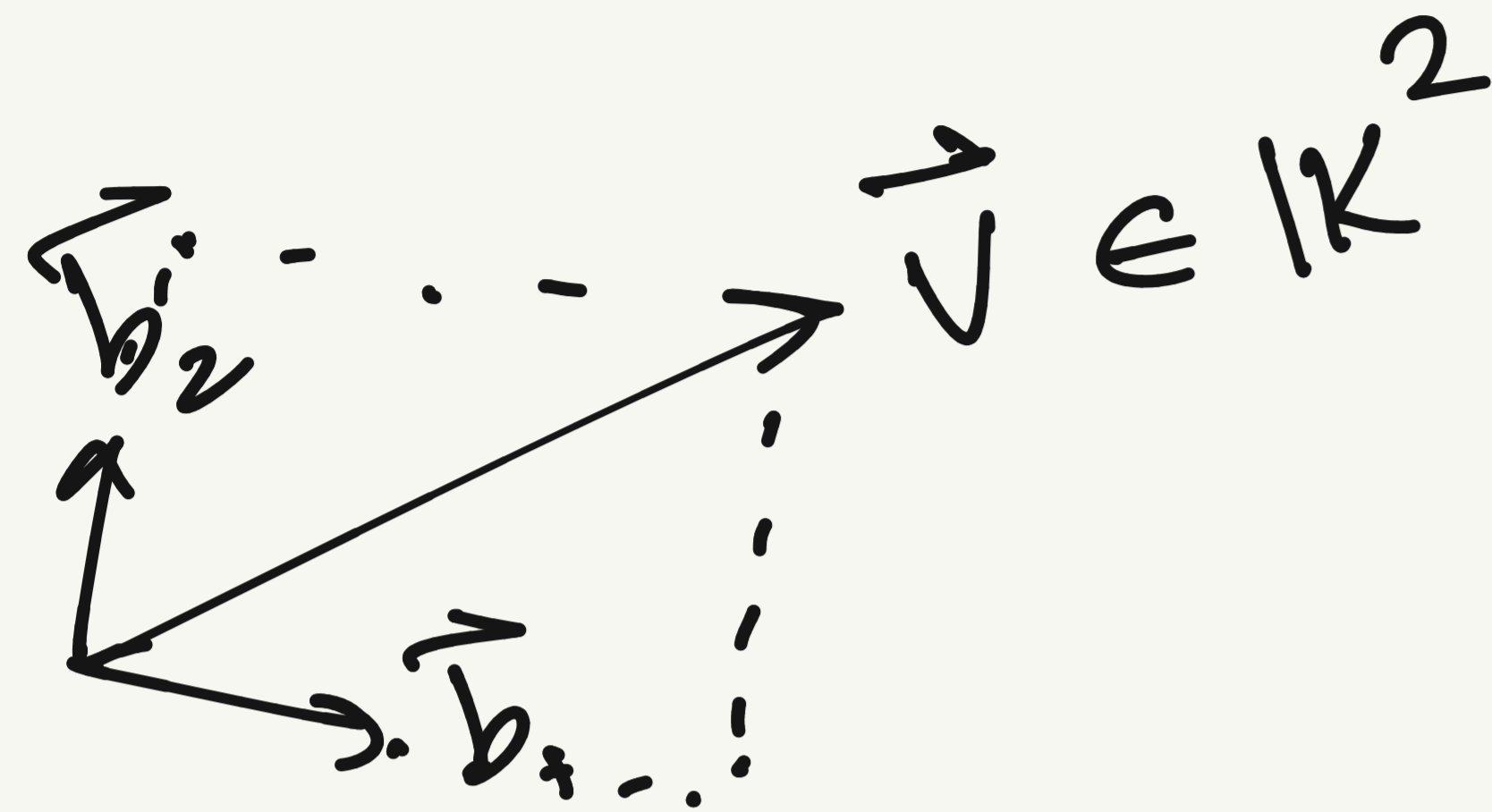
$$\Leftrightarrow \text{null}(A) = \{\vec{0}\}$$

0 is not eigenvalue of A

$0 \not\approx$ not singular value of A

$$\Leftrightarrow \det(A) \neq 0$$

$$\beta = \{\vec{b}_1, \dots, \vec{b}_m\} \Rightarrow B = \left[\begin{array}{c|c|c} \vec{b}_1 & \dots & \vec{b}_m \end{array} \right]$$



$$\vec{v} = \alpha_1 \vec{b}_1 + \alpha_2 \vec{b}_2 + \dots + \alpha_m \vec{b}_m$$

$$= B\vec{\alpha}$$

$$\vec{\alpha} = B^{-1} \vec{v} = [\vec{v}]_{\beta}$$

$$\vec{\alpha} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$$

$B \rightarrow B^{-1} \rightarrow B^{-1} \vec{v}$
unstable !!

Solve for $\vec{\alpha}$ s.t. $B\vec{\alpha} = \vec{v}$

$$A = \begin{bmatrix} 1.00 & 2.01 \\ 1.01 & 2.03 \end{bmatrix}$$

$$b = \begin{bmatrix} 1.01 \\ 1.02 \end{bmatrix}$$

- Solve $A\vec{x} = \vec{b}$
- ① by A^{-1}
 - ② by elimination