

# Linear Algebra Basics

A **vector** is a tuple of values

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_d \end{pmatrix}$$

$$\mathbf{x} = (x_1 \quad x_2 \quad \dots \quad x_d)$$

Linear algebra provides a way to do calculations with vectors

# Linear Algebra Basics

A **vector space** over  $\mathbb{R}$  is a set of vectors  $V$  closed under the operations vector addition and scalar multiplication.

**vector addition:**

$$\forall \mathbf{x}, \mathbf{y} \in V: \quad \mathbf{x} + \mathbf{y} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \end{pmatrix} + \begin{pmatrix} y_1 \\ y_2 \\ \vdots \end{pmatrix} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \end{pmatrix} \in V$$

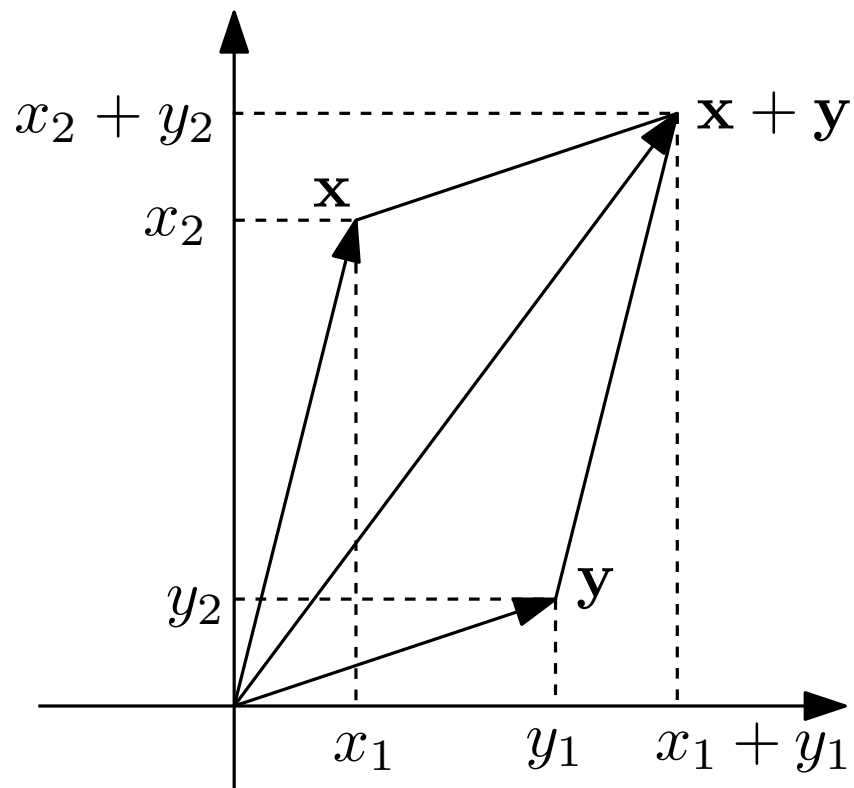
**scalar multiplication:**

$$\forall \lambda \in \mathbb{R}, \mathbf{x} \in V: \quad \lambda \mathbf{x} = \lambda \begin{pmatrix} x_1 \\ x_2 \\ \vdots \end{pmatrix} = \begin{pmatrix} \lambda x_1 \\ \lambda x_2 \\ \vdots \end{pmatrix} \in V$$

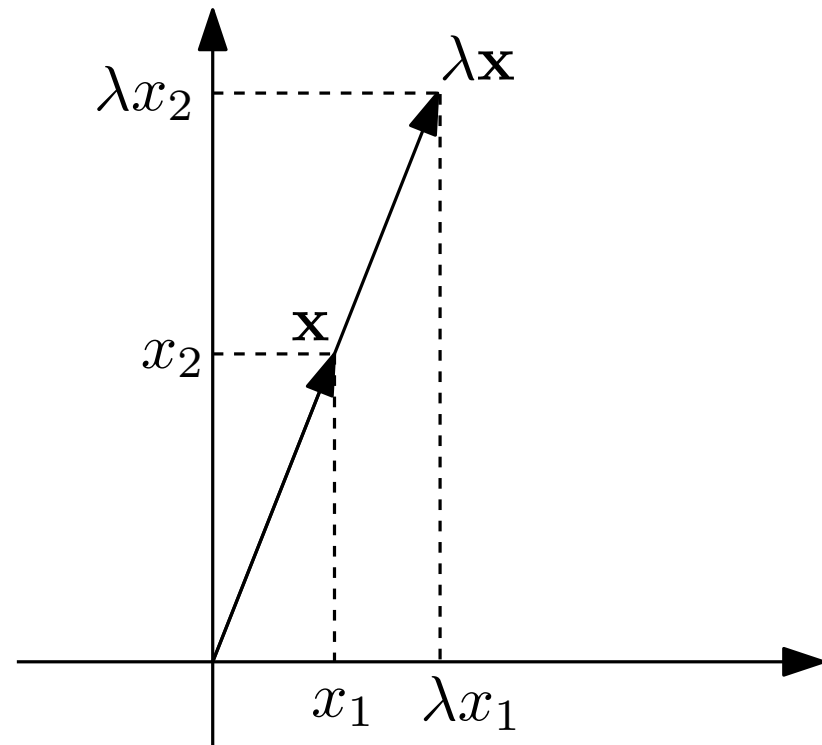
# Linear Algebra Basics

## Geometric interpretation

### vector addition:



### scalar multiplication:



# Linear Algebra Basics

A **basis** of the vector space is a set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_d\}$ , such that every  $\mathbf{x} \in V$  can be expressed as a linear combination of the basis vectors:

$$\mathbf{x} = x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \dots + x_d \mathbf{v}_d$$

Standard basis:  $\mathbf{v}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \mathbf{v}_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} \dots \mathbf{v}_d = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$

# Linear Algebra Basics

The **inner product** of two vectors:

$$\langle \mathbf{x}, \mathbf{y} \rangle := \sum_{i=1}^d x_i y_i$$

The **outer product** of two vectors:

$$\mathbf{x} \otimes \mathbf{y} := \begin{pmatrix} x_1 y_1 & x_1 y_2 & \cdots & x_1 y_d \\ x_2 y_1 & x_2 y_2 & \cdots & x_2 y_d \\ \vdots & \ddots & & \\ x_d y_1 & x_d y_2 & \cdots & x_d y_d \end{pmatrix}$$

# Linear Algebra Basics

Transpose of a matrix switches row and column indices

$$\begin{pmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \\ a_{4,1} & a_{4,2} & a_{4,3} \end{pmatrix} \Rightarrow \begin{pmatrix} a_{1,1} & a_{2,1} & a_{3,1} & a_{4,1} \\ a_{1,2} & a_{2,2} & a_{3,2} & a_{4,2} \\ a_{1,3} & a_{2,3} & a_{3,3} & a_{4,3} \end{pmatrix}$$

$\mathbf{A}$   $\mathbf{A}^T$

It holds that  $(\mathbf{A} \cdot \mathbf{B})^T = \mathbf{B}^T \cdot \mathbf{A}^T$

# Linear Algebra Basics

Matrix Multiplication  $\mathbf{A} \cdot \mathbf{B} = \mathbf{C}$

Each entry  $c_{ij}$  of  $\mathbf{C}$  is defined as

$$c_{ij} = \sum_{\ell} a_{i,\ell} \cdot b_{\ell,j} = \langle \mathbf{a}_i, \mathbf{b}_j \rangle$$

$\mathbf{a}_i$  is the  $i$ 'th row vector of  $\mathbf{A}$

$\mathbf{b}_j$  is the  $j$ 'th column vector of  $\mathbf{B}$

$$\mathbf{A} \begin{pmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \\ a_{4,1} & a_{4,2} & a_{4,3} \end{pmatrix} \cdot \mathbf{B} \begin{pmatrix} b_{1,1} & b_{1,2} & b_{1,3} & b_{1,4} \\ b_{2,1} & b_{2,2} & b_{2,3} & b_{2,4} \\ b_{3,1} & b_{3,2} & b_{3,3} & b_{3,4} \end{pmatrix} = \mathbf{C} \begin{pmatrix} c_{1,1} & c_{1,2} & c_{1,3} & c_{1,4} \\ c_{2,1} & c_{2,2} & c_{2,3} & c_{2,4} \\ c_{3,1} & c_{3,2} & c_{3,3} & c_{3,4} \\ c_{4,1} & c_{4,2} & c_{4,3} & c_{4,4} \end{pmatrix}$$

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# Linear Algebra Basics

The **span** of a set of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k$  is the set of all possible linear combinations of these vectors

$$\left\{ \sum_{i=1}^k \lambda_i \mathbf{v}_i \ : \ \lambda_i \in \mathbb{R} \right\}$$

The **rank** of a matrix is the dimension of the space spanned by its column vectors (or row vectors)

# Linear Algebra Basics

A **linear map** is a mapping between two vector spaces  $f : V \rightarrow W$  that satisfies the following two conditions:

$$(1) f(\mathbf{x} + \mathbf{y}) = f(\mathbf{x}) + f(\mathbf{y})$$

$$(2) \forall \gamma \in \mathbb{R} : f(\gamma \mathbf{x}) = \gamma f(\mathbf{x})$$

Any matrix  $\mathbf{A}$  defines a **linear map**:

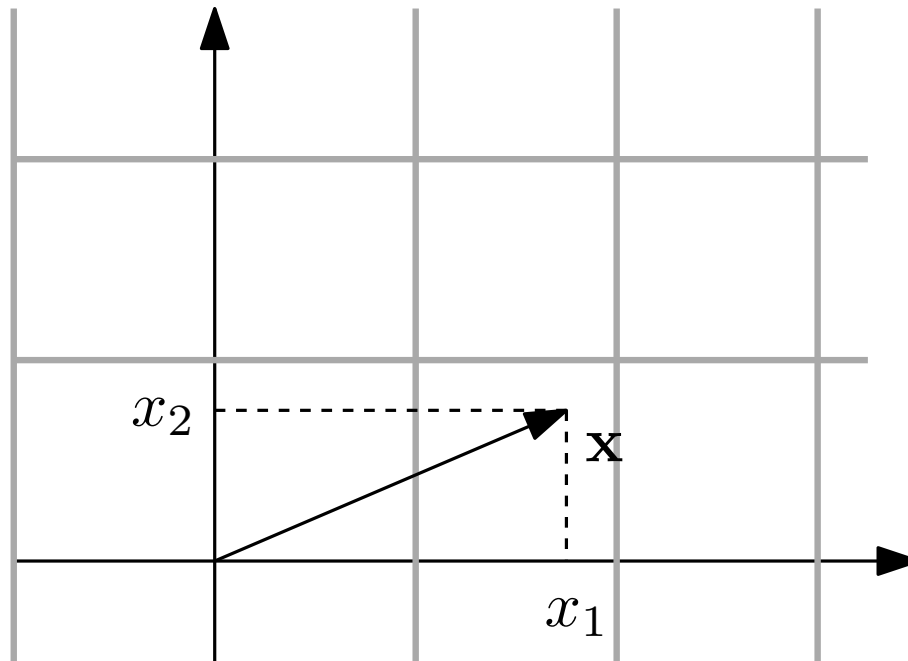
$$f : V \rightarrow W$$

$$f(\mathbf{x}) = \mathbf{A} \cdot \mathbf{x}$$

# Linear Algebra Basics

Using linearity, we can expand  $\mathbf{A} \cdot \mathbf{x}$  using the standard basis:

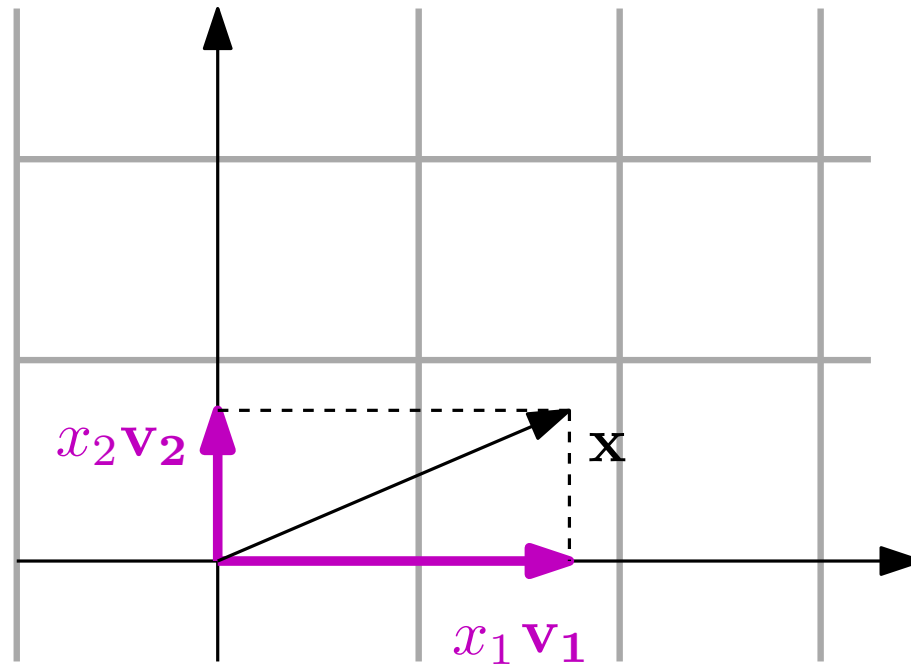
$$\mathbf{A}\mathbf{x} = \mathbf{A}(x_1\mathbf{v}_1 + x_2\mathbf{v}_2) = x_1\mathbf{A}\mathbf{v}_1 + x_2\mathbf{A}\mathbf{v}_2$$
$$\begin{pmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{pmatrix} \left( x_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right) = x_1 \begin{pmatrix} a_{1,1} \\ a_{2,1} \end{pmatrix} + x_2 \begin{pmatrix} a_{1,2} \\ a_{2,2} \end{pmatrix}$$



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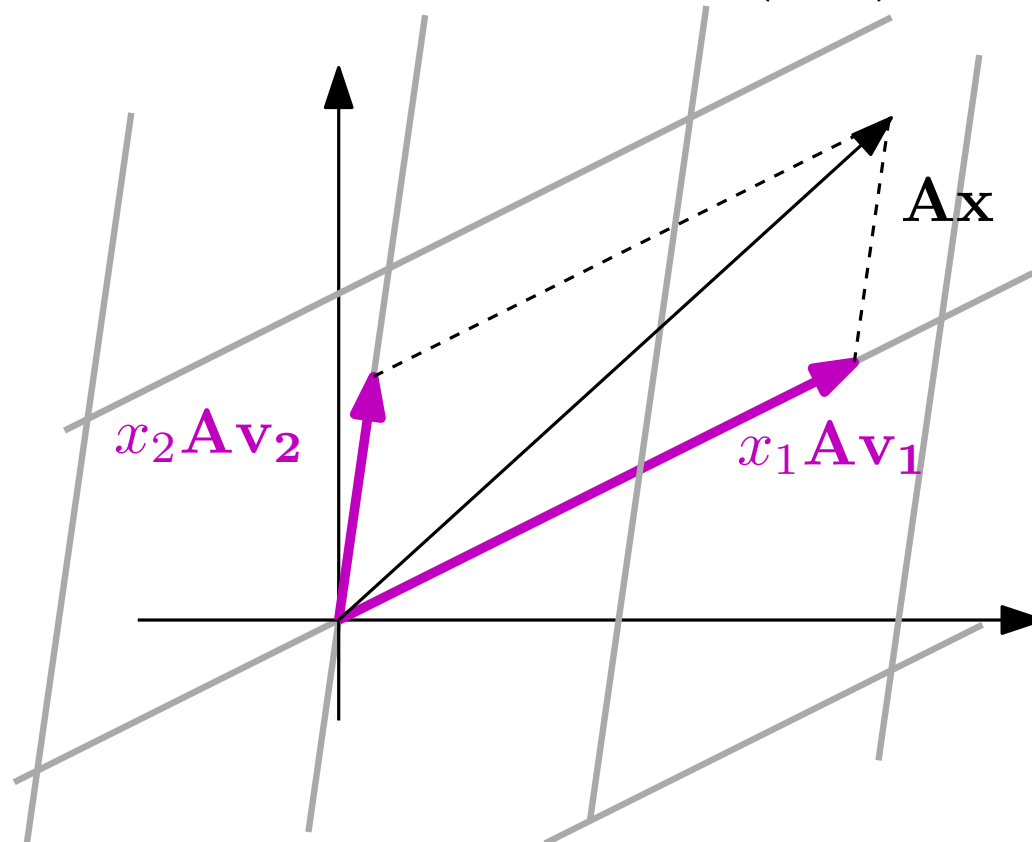
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# Linear Algebra Basics

The **length** of a vector:

$$\|\mathbf{x}\| := \sqrt{\sum_{i=1}^d x_i^2}$$

Vector  $\mathbf{x}$  has **unit length** if and only if  $\langle \mathbf{x}, \mathbf{x} \rangle = 1$

The **angle** between two vectors  $\mathbf{x}$  and  $\mathbf{y}$ :

$$\cos(\alpha) = \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\| \|\mathbf{y}\|}$$

Vectors  $\mathbf{x}$  and  $\mathbf{y}$  are **orthogonal** if and only if  $\langle \mathbf{x}, \mathbf{y} \rangle = 0$

# Linear Algebra Basics

Geometric interpretation

$$\langle \mathbf{x}, \mathbf{y} \rangle = \|\mathbf{x}\| \|\mathbf{y}\| \cos(\alpha)$$

