

# Linear Mappings

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## 01 Algebra

### Vector Space

#### DEFINITION 1

A vector space is any collection of objects that can be added together and multiplied by a scalar, and the resulting object is still a vector.

This property can be specified by considering two vector spaces  $V, W$  and a mapping  $\Phi : V \rightarrow W$  that preserves the structure of the vector space if and only if

$$\Phi(x + y) = \Phi(x) + \Phi(y)$$

$$\Phi(\lambda x) = \lambda \Phi(x)$$

for all  $x, y \in V$  and  $\lambda \in \mathbb{R}$ . From this we can define a linear mapping

### Linear Mapping

#### DEFINITION 2

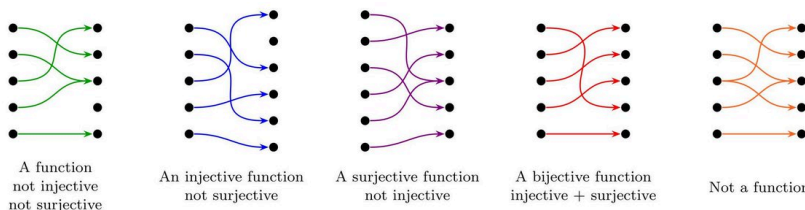
A linear mapping, also known as a *vector space homomorphism* or *linear transform*  $\Phi : V \rightarrow W$  is defined for

$$\forall x, y \in V \forall \lambda, \psi \in \mathbb{R} : \Phi(\lambda x + \psi y) = \lambda \Phi(x) + \psi \Phi(y)$$

which essentially reads for all vectors  $x, y$  in  $V$  and all  $\lambda, \psi$  in  $\mathbb{R}$  the linear mapping  $\Phi$  satisfies the scalar multiplication and additivity constraints.

These definitions are useful because we can define certain special mappings.

- **Injective** if  $\forall x, y \in V : \Phi(x) = \Phi(y) \Rightarrow x = y$
- **Surjective** if  $\Phi(V) = W$
- **Bijective** if  $\Phi$  is both injective and surjective



This specifies that if  $\Phi$  is surjective, then every element in  $W$  can be “reached” from  $\Phi(V)$ . For a bijective  $\Phi$ , we can furthermore define an inverse mapping  $\Psi : W \rightarrow V$  such that  $\Psi \circ \Phi(x) = x$ .

Then, we can also introduce some special cases of linear mappings

- **Isomorphism:**  $\Phi : V \rightarrow W$  is linear and bijective
- **Endomorphism:**  $\Phi : V \rightarrow V$  is linear
- **Automorphism:**  $\Phi : V \rightarrow V$  linear and bijective

## 02 Matrix Representation

Based on this we can state the theorem that *finite-dimensional vector spaces  $V$  and  $W$  are isomorphic if and only if  $\dim(V) = \dim(W)$* . From hence we can define a matrix from first principles as representing an  $n$ -dimensional vector space isomorphic to  $\mathbb{R}^n$ . Specifically, we consider a basis  $\{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  of an  $n$ -dimensional vector space  $V$  isomorphic to  $\mathbb{R}^n$ . As we are familiar with in linear algebra, this basis represents a linear transformation on a coordinate system.

**REMARK:** An **ordered** basis is a special form of basis in which the order of vectors matters. Our canonical linear algebra intuition relies on this constraint, since matrix  $[\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3] \neq [\mathbf{v}_3, \mathbf{v}_2, \mathbf{v}_1]$ . However it is rarely necessary for understanding vector spaces. We specify ordered basis  $B = (\mathbf{b}_1, \dots, \mathbf{b}_n)$  and unordered basis  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$

### Transformation Matrix

### DEFINITION 3

Consider vector spaces  $V, W$  with ordered bases  $B = (\mathbf{b}_1, \dots, \mathbf{b}_n)$  and  $C = (\mathbf{c}_1, \dots, \mathbf{c}_m)$ , scalars  $\alpha \in \mathbb{R}^n$  and a linear mapping  $\Phi : V \rightarrow W$  such that

$$\Phi(\mathbf{b}_j) = \alpha_{1j}\mathbf{c}_1 + \dots + \alpha_{mj}\mathbf{c}_m = \sum_{i=1}^m \alpha_{ij}\mathbf{c}_i$$

The result of this in matrix form will be the *transformation matrix* corresponding to  $\Phi$  (with respect to the chosen bases  $B$  and  $C$ ) which we call  $\mathbf{A}_\Phi$  such that for  $\mathbf{y} = \Phi(\mathbf{x}) \in W$ ,

$$\mathbf{A}_\Phi \hat{\mathbf{x}} = \hat{\mathbf{y}}$$

And just like that we've gone from algebra to linear algebra.

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