

Highlights from §1.8: Linear Transformations

Think of $A\mathbf{x} = \mathbf{b}$ as “Matrix A acting on vector \mathbf{x} to produce a vector \mathbf{b} ”

Or “Multiply A by \mathbf{x} to get a new product \mathbf{b} ”

If A is $m \times n$ and \mathbf{x} is $n \times 1$ then $\mathbf{b} = A\mathbf{x}$ is $m \times 1$.

Recall: The vector $\mathbf{b} = A\mathbf{x}$ is a linear combination of the columns of A with weights the components of \mathbf{x} .

We can create a function T from \mathbb{R}^n to \mathbb{R}^m by $T(\mathbf{x}) = A\mathbf{x}$.

Definitions: A **transformation** T from \mathbb{R}^n to \mathbb{R}^m is a rule which assigns to each vector \mathbf{x} in \mathbb{R}^n some vector $T(\mathbf{x})$ in \mathbb{R}^m . The set \mathbb{R}^n is the **domain** of T and the set \mathbb{R}^m is the **codomain** of T . The vector $T(\mathbf{x})$ is the **image** of \mathbf{x} under the action of T . The set of all images of T is called the **range** of T .

Some Examples with $n = 2$.

If $T(\mathbf{x}) = A\mathbf{x}$ for some matrix A , then T is called a **matrix transformation**.

Definition: A transformation T from \mathbb{R}^n to \mathbb{R}^m is called a linear transformation if for all \mathbf{u} and \mathbf{v} in the domain of T and all scalars c , both

- (i) $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$, and
- (ii) $T(c\mathbf{u}) = c T(\mathbf{u})$

Note: **Every matrix transformation is a linear transformation.**

Not all transformations are linear

Theorem: If T is a linear transformation, then

$$T(\mathbf{0}) = \mathbf{0} \text{ and}$$

$$T(c\mathbf{u} + d\mathbf{v}) = c T(\mathbf{u}) + d T(\mathbf{v})$$

Special Transformations: **shear, dilation, contraction**

