

Linear functions: Which functions can be expressed as a matrix-vector product?

In each example, we *assumed* the function could be expressed as a matrix-vector product.

How can we verify that assumption?

We'll state two algebraic properties.

- ▶ If a function can be expressed as a matrix-vector product $\mathbf{x} \mapsto M * \mathbf{x}$, it has these properties.
- ▶ If the function from \mathbb{F}^C to \mathbb{F}^R has these properties, it can be expressed as a matrix-vector product.

Linear functions: Which functions can be expressed as a matrix-vector product?

Let \mathcal{V} and \mathcal{W} be vector spaces over a field \mathbb{F} .

Suppose a function $f : \mathcal{V} \longrightarrow \mathcal{W}$ satisfies two properties:

Property L1: For every vector \mathbf{v} in \mathcal{V} and every scalar α in \mathbb{F} ,

$$f(\alpha \mathbf{v}) = \alpha f(\mathbf{v})$$

Property L2: For every two vectors \mathbf{u} and \mathbf{v} in \mathcal{V} ,

$$f(\mathbf{u} + \mathbf{v}) = f(\mathbf{u}) + f(\mathbf{v})$$

We then call f a *linear function*.

Proposition: Let M be an $R \times C$ matrix, and suppose $f : \mathbb{F}^C \mapsto \mathbb{F}^R$ is defined by $f(\mathbf{x}) = M * \mathbf{x}$. Then f is a linear function.

Proof: Certainly \mathbb{F}^C and \mathbb{F}^R are vector spaces.

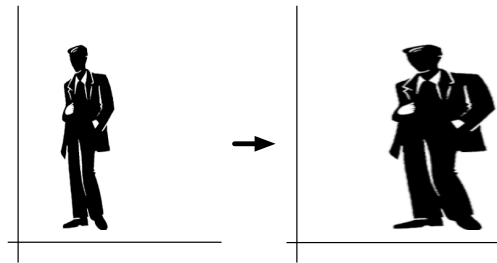
We showed that $M * (\alpha \mathbf{v}) = \alpha M * \mathbf{v}$. This proves that f satisfies Property L1.

We showed that $M * (\mathbf{u} + \mathbf{v}) = M * \mathbf{u} + M * \mathbf{v}$. This proves that f satisfies Property L2.

QED

Which functions are linear?

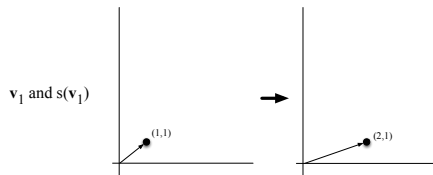
Define $s([x, y]) = \text{stretching by two in horizontal direction}$



Which functions are linear?

Define $s([x, y]) =$ stretching by two in horizontal direction

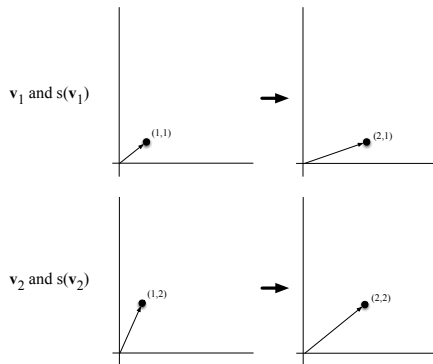
Property L1: $s(\mathbf{v}_1 + \mathbf{v}_2) = s(\mathbf{v}_1) + s(\mathbf{v}_2)$



Which functions are linear?

Define $s([x, y]) = \text{stretching by two in horizontal direction}$

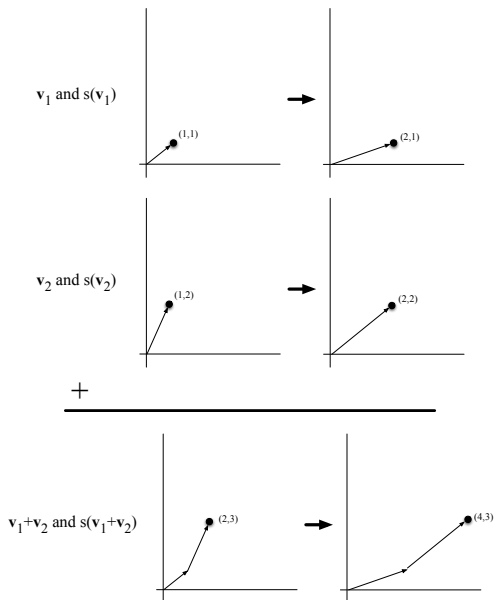
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Which functions are linear?

Define $s([x, y]) = \text{stretching by two in horizontal direction}$

Property L1: $s(\mathbf{v}_1 + \mathbf{v}_2) = s(\mathbf{v}_1) + s(\mathbf{v}_2)$



Which functions are linear?

Define $s([x, y]) = \text{stretching by two in horizontal direction}$

$$\text{Property L1: } s(\mathbf{v}_1 + \mathbf{v}_2) = s(\mathbf{v}_1) + s(\mathbf{v}_2)$$

$$\text{Property L2: } s(\alpha \mathbf{v}) = \alpha s(\mathbf{v})$$

Since the function $s(\cdot)$ satisfies Properties L1 and L2, it is a linear function.

Similarly can show rotation by θ degrees is a linear function.

What about translation?

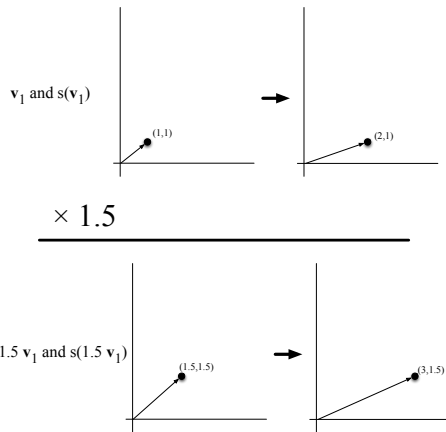
$$t([x, y]) = [x, y] + [1, 2]$$

This function violates Property L1. For example:

$$t([4, 5] + [2, -1]) = t([6, 4]) = [7, 6]$$

but

$$t([4, 5]) + t([2, -1]) = [5, 7] + [3, 1] = [8, 8]$$



Since $t(\cdot)$ violates Property L1 for at least one input, it is **not** a linear function.

Can similarly show that $t(\cdot)$ does not satisfy Property L2.

A linear function maps zero vector to zero vector

Lemma: If $f : \mathcal{U} \longrightarrow \mathcal{V}$ is a linear function then f maps the zero vector of \mathcal{U} to the zero vector of \mathcal{V} .

Proof: Let $\mathbf{0}$ denote the zero vector of \mathcal{U} , and let $\mathbf{0}_{\mathcal{V}}$ denote the zero vector of \mathcal{V} .

$$f(\mathbf{0}) = f(\mathbf{0} + \mathbf{0}) = f(\mathbf{0}) + f(\mathbf{0})$$

Subtracting $f(\mathbf{0})$ from both sides, we obtain

$$\mathbf{0}_{\mathcal{V}} = f(\mathbf{0})$$

QED

Linear functions: Pushing linear combinations through the function

Defining properties of linear functions:

Property L1: $f(\alpha \mathbf{v}) = \alpha f(\mathbf{v})$

Property L2: $f(\mathbf{u} + \mathbf{v}) = f(\mathbf{u}) + f(\mathbf{v})$

Proposition: For a linear function f ,
for any vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ in the domain of f and any scalars $\alpha_1, \dots, \alpha_n$,

$$f(\alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n) = \alpha_1 f(\mathbf{v}_1) + \dots + \alpha_n f(\mathbf{v}_n)$$

Proof: Consider the case of $n = 2$.

$$\begin{aligned} f(\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2) &= f(\alpha_1 \mathbf{v}_1) + f(\alpha_2 \mathbf{v}_2) && \text{by Property L2} \\ &= \alpha_1 f(\mathbf{v}_1) + \alpha_2 f(\mathbf{v}_2) && \text{by Property L1} \end{aligned}$$

Proof for general n is similar.

QED

Linear functions: Pushing linear combinations through the function

Proposition: For a linear function f ,
$$f(\alpha_1 \mathbf{v}_1 + \cdots + \alpha_n \mathbf{v}_n) = \alpha_1 f(\mathbf{v}_1) + \cdots + \alpha_n f(\mathbf{v}_n)$$

Example: $f(\mathbf{x}) = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} * \mathbf{x}$

Verify that $f(10[1, -1] + 20[1, 0]) = 10f([1, -1]) + 20f([1, 0])$

$$\begin{aligned} & \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \left(10[1, -1] + 20[1, 0] \right) \\ &= \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \left([10, -10] + [20, 0] \right) \\ &= \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} [30, -10] \\ &= 30[1, 3] - 10[2, 4] \\ &= [30, 90] - [20, 40] \\ &= [10, 50] \end{aligned}$$

$$\begin{aligned} & 10 \left(\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} * [1, -1] \right) + 20 \left(\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} * [1, 0] \right) \\ &= 10([1, 3] - [2, 4]) + 20(1[1, 3]) \\ &= 10[-1, -1] + 20[1, 3] \\ &= [-10, -10] + [20, 60] \\ &= [10, 50] \end{aligned}$$

From function to matrix, revisited

We saw a method to derive a matrix from a function:

Given a function $f : \mathbb{R}^n \longrightarrow \mathbb{R}^m$, we want a matrix M such that $f(\mathbf{x}) = M * \mathbf{x} \dots$

- ▶ Plug in the standard generators $\mathbf{e}_1 = [1, 0, \dots, 0, 0], \dots, \mathbf{e}_n = [0, \dots, 0, 1]$
- ▶ Column i of M is $f(\mathbf{e}_i)$.

This works correctly whenever such a matrix M really exists:

Proof: If there is such a matrix then f is linear:

- ▶ (Property L1) $f(\alpha \mathbf{v}) = \alpha f(\mathbf{v})$ and
- ▶ (Property L2) $f(\mathbf{u} + \mathbf{v}) = f(\mathbf{u}) + f(\mathbf{v})$

Let $\mathbf{v} = [\alpha_1, \dots, \alpha_n]$ be any vector in \mathbb{R}^n .

We can write \mathbf{v} in terms of the standard generators.

$$\mathbf{v} = \alpha_1 \mathbf{e}_1 + \dots + \alpha_n \mathbf{e}_n$$

so

$$\begin{aligned} f(\mathbf{v}) &= f(\alpha_1 \mathbf{e}_1 + \dots + \alpha_n \mathbf{e}_n) \\ &= \alpha_1 f(\mathbf{e}_1) + \dots + \alpha_n f(\mathbf{e}_n) \\ &= \alpha_1 (\text{column 1 of } M) + \dots + \alpha_n (\text{column } n \text{ of } M) \\ &= M * \mathbf{v} \text{ QED} \end{aligned}$$

Linear functions and zero vectors: Kernel

Definition: *Kernel* of a linear function f is $\{\mathbf{v} : f(\mathbf{v}) = \mathbf{0}\}$

Written $\text{Ker } f$

For a function $f(\mathbf{x}) = M * \mathbf{x}$,

$$\text{Ker } f = \text{Null } M$$

Kernel and one-to-one

One-to-One Lemma: A linear function is one-to-one if and only if its kernel is a trivial vector space.

Proof: Let $f : \mathcal{U} \longrightarrow \mathcal{V}$ be a linear function. We prove two directions.

- ▶ Suppose $\text{Ker } f$ contains some nonzero vector \mathbf{u} , so $f(\mathbf{u}) = \mathbf{0}_{\mathcal{V}}$.
Because a linear function maps zero to zero, $f(\mathbf{0}) = \mathbf{0}_{\mathcal{V}}$ as well,
so f is not one-to-one.

- ▶ Suppose $\text{Ker } f = \{\mathbf{0}\}$.
Let $\mathbf{v}_1, \mathbf{v}_2$ be any vectors such that $f(\mathbf{v}_1) = f(\mathbf{v}_2)$.
Then $f(\mathbf{v}_1) - f(\mathbf{v}_2) = \mathbf{0}_{\mathcal{V}}$
so, by linearity, $f(\mathbf{v}_1 - \mathbf{v}_2) = \mathbf{0}_{\mathcal{V}}$,
so $\mathbf{v}_1 - \mathbf{v}_2 \in \text{Ker } f$.
Since $\text{Ker } f$ consists solely of $\mathbf{0}$,
it follows that $\mathbf{v}_1 - \mathbf{v}_2 = \mathbf{0}$, so $\mathbf{v}_1 = \mathbf{v}_2$.

QED

Kernel and one-to-one

One-to-One Lemma A linear function is one-to-one if and only if its kernel is a trivial vector space.

Define the function $f(\mathbf{x}) = A * \mathbf{x}$.

If $\text{Ker } f$ is trivial (i.e. if $\text{Null } A$ is trivial)

then a vector \mathbf{b} is the image under f of at most one vector.

That is, at most one vector \mathbf{u} such that $A * \mathbf{u} = \mathbf{b}$

That is, the solution set of $A * \mathbf{x} = \mathbf{b}$ has at most one vector.

Linear functions that are onto?

Question: How can we tell if a linear function is onto?

Recall: for a function $f : \mathcal{V} \rightarrow \mathcal{W}$, the *image* of f is the set of all images of elements of the domain:

$$\{f(\mathbf{v}) : \mathbf{v} \in \mathcal{V}\}$$

(You might know it as the “range” but we avoid that word here.)

The image of function f is written $\text{Im } f$

“Is function f is onto?” same as “is $\text{Im } f = \text{co-domain of } f$?”

Example: *Lights Out*

$$\text{Define } f([\alpha_1, \alpha_2, \alpha_3, \alpha_4]) = \left[\begin{array}{|c|c|c|c|} \hline \begin{array}{cc} \bullet & \bullet \\ \bullet & \bullet \end{array} & \begin{array}{cc} \bullet & \bullet \\ & \bullet \end{array} & \begin{array}{cc} \bullet & \\ \bullet & \bullet \end{array} & \begin{array}{cc} & \bullet \\ \bullet & \bullet \end{array} \\ \hline \end{array} \right] * [\alpha_1, \alpha_2, \alpha_3, \alpha_4]$$

$\text{Im } f$ is set of configurations for which 2×2 *Lights Out* can be solved,
so “ f is onto” means “ 2×2 *Lights Out* can be solved for every configuration”

Can 2×2 *Lights Out* be solved for every configuration? What about 5×5 ?

Each of these questions amounts to asking whether a certain function is onto.

Linear functions that are onto?

“Is function f is onto?” same as “is $\text{Im } f = \text{co-domain of } f$?”

First step in understanding how to tell if a linear function f is onto:

- ▶ study the image of f

Proposition: The image of a linear function $f : \mathcal{V} \longrightarrow \mathcal{W}$ is a vector space