

PHYSICAL FORMULATION

MATHEMATICS AS THE LANGUAGE OF PHYSICS

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OPERATORS

OPERATORS AS THE BRICKS OF EQUATIONS

This is NOT a PDE course.

If interested you can check this:

[Introduction to PDE | MIT](#)

So, what will we do today, then?

- 1) Understand the concept of operator
- 2) Look at operator properties

An operator is a mapping:

It takes an input from a vector space and returns an output in another, not necessarily the same, vector space.

It generalizes the concept of function.

∇ , $\nabla \cdot$ and ∇^2 are your breakfast operators!

In mathematical terms:

$$A := U \rightarrow V$$

Vector spaces:

A vector space is a set of objects (vectors) over a field of scalars that satisfy vector addition and scalar multiplication.

We are familiar with n-dimensional vectors:

$$\vec{u} = (u_1 \hat{e}_1 + u_2 \hat{e}_2 + \dots + u_n \hat{e}_n)$$

Where the basis is defined as:

$$(\hat{e}_1, \hat{e}_2, \dots, \hat{e}_n) = \{(1, 0, \dots, 0), (0, 1, \dots, 0), (0, 0, \dots, 1)\}$$

Vector spaces:

As far as we satisfy vector addition and scalar multiplication* we can choose a different basis such as:

$$(\hat{e}_1, \hat{e}_2, \dots, \hat{e}_n) = \{\sin(\pi x), \sin(2\pi x), \dots, \sin(n\pi x)\}$$

*There are actually 8 axioms that need to be satisfied to formally prove that a set of vectors and a field form a vector space. More [here](#).

Vector spaces:

You can look at “normal” vectors as a function where the input space is the position of the vector. Like in a tabulated function, every index corresponds with a value. For an n -dimensional vector:

$$\hat{e}_1 = \{1, 0, \dots, 0\}$$

$$\hat{e}_1 : \{\mathbb{N} \leq n\} \rightarrow \mathbb{R}$$

Note that the size of the input space is n , which is the number of components that the space has.

Vector spaces:

When you have a given vector:

$$\vec{u} = (u_1 \hat{e}_1 + u_2 \hat{e}_2 + \dots + u_n \hat{e}_n)$$

You end up having something that:

$$\vec{u} : \{\mathbb{N} \leq n\} \rightarrow \mathbb{R}$$

Note that the dimension of u is not necessarily the same as n (i.e., a 2D plane in a 3D space)

$$\vec{u}_{plane} = x_{plane} \{1, 1, 0\} + y_{plane} \{0, 0, 1\}$$

Vector spaces:

“Function” vectors define a set of values as well, but instead of a “table”, they use a function. We say that the independent variable is the input and the dependent one is the output.

$$\hat{e}_1 = \sin(\omega_1 t)$$
$$\hat{e}_1 : \mathbb{R} \rightarrow \mathbb{R}$$

Note that the input space are the real numbers and thus the **size of the input space is infinity**.

Vector spaces:

When you have a given vector:

$$\vec{u} = (u_1 \hat{e}_1 + u_2 \hat{e}_2 + \dots + u_n \hat{e}_n)$$

You end up having something that:

$$\vec{u} : \mathbb{R} \rightarrow \mathbb{R}$$

So we end up with another function!

Note that, like in the Fourier series, we may need an infinite sum of base functions to represent a particular one.

Vector spaces:

Finally, note that you already knew several examples of this: Fourier and Taylor series are just ways to represent a particular function in a different basis!!!

$$f(x) = \sum_{n=0}^{\infty} a_n \sin(\omega_n x) + b_n \cos(\omega_n x)$$

$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n$$

Matrix multiplication

$$Mx$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$A = M \quad U = \mathbb{R}^3 \quad V = \mathbb{R}^2$$

Power

$$(x)^n$$

$$x^{1/2} = \sqrt{x}$$

$$A = ()^n \quad U = \mathbb{R} \quad V = \mathbb{C}$$

Integration

$$\int_{\Omega} f dx$$

$$\int \sin(n\pi x) dx = \frac{-1}{n\pi} \cos(n\pi x) \quad \nabla \sin(n\pi x) = n\pi \cos(n\pi x)$$

$$A = \int dx$$

$$U = \sin(n\pi x) \quad V = \cos(n\pi x) \quad \forall n \in \mathbb{Z}$$

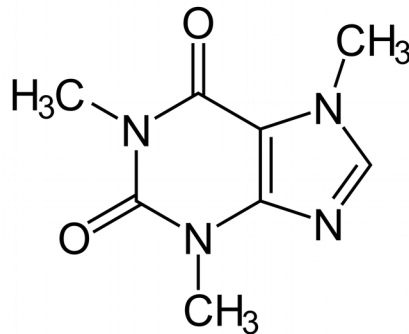
$$A = \nabla$$

$$U = \sin(n\pi x) \quad V = \cos(n\pi x) \quad \forall n \in \mathbb{Z}$$

Operators: Examples

PhD Student

Thesis Operator



$A = you$ $U = substances$ $V = science(?)$

Hold on, am I an operator?

A particular operator may require particular properties of either U or V .

Most of the times we will require our spaces to be equipped with a **dot product**, to have a **metric** (measure) and to be **differentiable** spaces (manifolds).

We will see how this plays a role in the coming slides and avoid mathematical formalism.

Quantum Mechanics

$$\hat{H} \Psi$$

$$\left(\frac{-\hbar}{2m} \nabla^2 + V\right) \Psi = Ek + Ep$$

$$A = \hat{H}$$

$$U = \mathbb{C} \quad V = \mathbb{R}$$

Fluid Mechanics

$$NS \begin{bmatrix} u \\ P \end{bmatrix}$$

$$\nabla \cdot u = 0$$

$$\rho \frac{Du}{Dt} + \nabla P = \nabla \cdot \sigma$$

$$A = NS$$

$$U = \mathbb{R}^3 \times \mathbb{R} \quad V = \mathbb{R} \times \mathbb{R}^3$$

Operator properties provide **qualitative insight into the output**.

Is this something you want to preserve? Or not?

We will look at:

- Linearity
- Kernel
- Symmetry
- Normalization
- Conservation
- Norm-conservation
- Definitiveness

Preliminary concepts

Endomorphism:

$$A: U \rightarrow U$$

Dot product:

$$(\cdot)(\cdot): U \times U \rightarrow \mathbb{R}$$

Norm:

$$\|\cdot\|: U \rightarrow \mathbb{R}$$

A dot product induces a norm!

Linearity

Conditions:

$$A(x + y) = A(x) + A(y)$$

$$A(\alpha x) = \alpha A(x)$$

Linear operator

Example: variables

$$M(x + y) = Mx + My \rightarrow OK$$

$$(x + y)^n \neq x^n + y^n \rightarrow NOT \text{ linear !}$$

Example: functions

$$\int (f + g) dx = \int f dx + \int g dx \rightarrow OK$$

$$\nabla (f + g) = \nabla f + \nabla g \rightarrow OK$$

Kernel

Definition:

$$\ker(A) = \{u \mid Au = \emptyset\}$$

Tells us what is the space that maps to 0.

Example: variables

Matrix:

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

$$\ker(A) = c \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \quad \forall c \in \mathbb{R}$$

Kernel

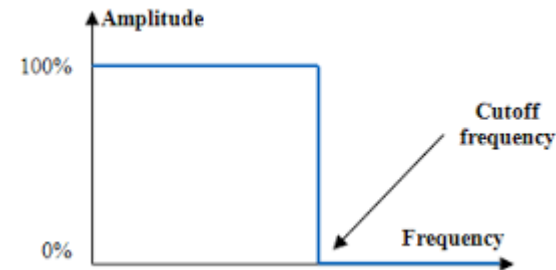
Definition:

$$\ker(A) = \{u | Au = \emptyset\}$$

Tells us what is the space that maps to 0.

Example: functions

Low pass filter:



$$F : S \rightarrow S$$

$$S = \{\sin(\omega_i t)\}$$

$$\ker(F) = \{S | \omega_i > \omega_{cutoff}\}$$

Symmetry

Conditions:

$$A : U \rightarrow U$$

$$\langle Ax | y \rangle = \langle x | Ay \rangle$$

Bracket notation!!

Literally means:

DOT PRODUCT!!

$$\langle Ax | y \rangle := (Ax) \cdot y$$

Example: variables

Matrix multiplication

$$(Mx) \cdot y = x \cdot (My) ?$$

We note that:

$$a \cdot b = a^T b \quad (Ax)^T = x^T A^T$$

So:

$$x^T M^T y = x^T M y \Leftrightarrow M^T = M$$

Standard matrix transpose!

Symmetry

We need to define a
DOT product for
functions!

$$\langle u|v \rangle := \int_{\Omega} uv$$

IMPORTANT:

REMEMBER!!!

Example: functions

Laplacian

$$\langle \nabla^2 f | g \rangle = \langle f | \nabla^2 g \rangle ?$$

We note that:

$$\int_{\Omega} \nabla^2 f g = [\nabla f g]_{\partial\Omega} - \int_{\Omega} \nabla f \nabla g$$

What if $[\nabla f g]_{\partial\Omega} = 0$?

Can we do that?

Symmetry

We need to define a
DOT product for
functions!

$$\langle u|v \rangle := \int_{\Omega} uv \, dx$$

IMPORTANT:

REMEMBER!!!

Example: functions

Laplacian

In that case:

$$\int_{\Omega} \nabla^2 f \, g = - \int_{\Omega} \nabla f \cdot \nabla g$$

Which is the same as:

$$\int_{\Omega} f \, \nabla^2 g = - \int_{\Omega} \nabla f \cdot \nabla g$$

Symmetric!

Normalization

Conditions:

$$A: U \rightarrow U$$

$$A 1_U = 1_U$$

Constants are preserved!

Example: variables

Average matrix:

$$A 1_U = \begin{bmatrix} 1/2 & 1/2 & 0 \\ 1/3 & 1/3 & 1/3 \\ 0 & 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

Matrices whose rows sum 1

Normalization

Conditions:

$$A: U \rightarrow U$$

$$A 1_U = 1_U$$

Constants are preserved!

Example: functions

Helmholtz filter:

$$HF = I + \alpha \nabla^2$$

$$HF 1_U = I 1_U + \alpha \nabla^2 1_U \stackrel{=0}{=} 1_U$$

$$1_U \in \ker(\nabla^2)$$

Conservation

Conditions:

$$A: U \rightarrow U$$

$$\langle Ax | 1_U \rangle = \langle x | 1_U \rangle$$

Conserves dot product over unitary space vector.

Example: variables

Shift matrix

$$\Gamma x = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$\langle \Gamma x | 1_U \rangle = x^T \Gamma 1_U = x^T 1_U$$

Conservation

Conditions:

$$A: U \rightarrow U$$

$$\langle Ax | 1_U \rangle = \langle x | 1_U \rangle$$

Conserves integral.

Example: function

Shift operator:

$$\Gamma x = \exp(\theta) x \quad \forall x \in \exp(i \omega_j t)$$

$$\int_{\Omega} \exp(i \omega_j t + \theta) 1_U = \int_{\Omega} \exp(i \omega_j t + \theta)$$

$$\int_{\Omega} \exp(i \omega_j t) 1_U = \int_{\Omega} \exp(i \omega_j t) 1_U$$

$$\int_{\Omega} \exp(i \omega_j t + \theta) = \int_{\Omega} \exp(i \omega_j t) 1_U$$

Conservation

We can actually prove that a normalized and symmetric operator is conservative:

$$A:U \rightarrow U \quad A 1_U = 1_U \quad \langle Ax|y \rangle = \langle x|Ay \rangle$$

$$\langle Ax|1_U \rangle = \langle x|A 1_U \rangle = \langle x|1_U \rangle$$

$$\langle Ax|1_U \rangle = \langle x|1_U \rangle$$

Norm - Conservation

Conditions:

$$A: U \rightarrow U$$

$$\|Ax\| = \|x\| \quad \forall x$$

$$\langle Ax | Ax \rangle = \langle x | x \rangle$$

Conserves the norm of the input!

Depends on the norm!

Example: variables

Rotation matrix:

$$R = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

$$x^T R^T R x = x^T x \Leftrightarrow R^T R = I$$

$$R^T R = \begin{bmatrix} \cos^2 + \sin^2 & 0 \\ 0 & \sin^2 + \cos^2 \end{bmatrix}$$

Norm - Conservation

Conditions:

$$A: U \rightarrow U$$

$$\|Ax\| = \|x\| \quad \forall x$$

$$\langle Ax | Ax \rangle = \langle x | x \rangle$$

Conserves the norm of the input!

Depends on the norm!

Example: function

Shift operator

$$SH : \sin(\omega_i t) \rightarrow \sin(\omega_i t + \theta)$$

$$\int_{\Omega} \sin^2(\omega_i t) = \int_{\Omega} \sin^2(\omega_i t + \theta)$$

Definitiveness

Definition:

$$A: U \rightarrow U$$

$$\langle Ax | x \rangle = s \|x\| \quad \forall x \in U$$

$$\text{sign}(s) \begin{cases} > 0 \text{ positive} \\ \geq 0 \text{ non-negative} \\ \leq 0 \text{ non-positive} \\ < 0 \text{ negative} \end{cases}$$

Example: variables

Rotation matrix:

$$R = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

$$x^T R^T x = s x^T x$$

$$x^T R x = (x_1^2 + x_2^2) \cos(\theta)$$

It will depend on θ !

Definitiveness

Definition:

$$A:U \rightarrow U$$

$$\langle Ax|x \rangle = s \|x\| \quad \forall x \in U$$

$$\text{sign}(s) \begin{cases} > 0 \text{ positive} \\ \geq 0 \text{ non-negative} \\ \leq 0 \text{ non-positive} \\ < 0 \text{ negative} \end{cases}$$

Example: functions

Laplacian:

$$\langle \nabla^2 f | f \rangle = - \langle \nabla f | \nabla f \rangle$$

$$\langle \nabla^2 f | f \rangle = - \|\nabla f\|^2$$

Negative definite!

Review of our every day operators:

name	grad	div	lap
symbol	∇	$\nabla \cdot$	∇^2
linear	yes	yes	yes
kernel			
symm.			yes
norm.			
cons.			
def.			-

Let's fill the gaps!

Note that *grad* and *div* are **NOT endomorphisms!**

Symmetry, normalization, conservation, norm-conservation and definitiveness DO NOT APPLY!

Let's fill the gaps!

We can, however, find useful relations between them:

$$\langle \nabla \cdot (s \vec{f}) | 1_U \rangle = \langle \nabla \cdot \vec{f} | s \rangle + \langle \vec{f} | \nabla s \rangle$$

1 million \$ identity!

$$\int_{\Omega} \nabla \cdot (s \vec{f}) = \int_{\partial \Omega} s \vec{f} = 0$$

$$\langle \nabla \cdot \vec{f} | s \rangle = - \langle \vec{f} | \nabla s \rangle$$

We say that they are negative adjoint!

Let's fill the gaps!

We have several useful calculus identities as well:

$$\nabla \cdot \nabla \times A = 0 \quad \forall A \in VS(\text{Vector Space})$$

$$\nabla \times \nabla \theta = 0 \quad \forall \theta \in \mathbb{R}$$

$$\nabla \cdot \nabla = \nabla^2$$

Operators: Your friends

Review:

name	grad	div	lap
symbol	∇	$\nabla \cdot$	∇^2
linear	yes	yes	yes
kernel	$f(\vec{x}) = C$	$\nabla \times A$	$f(\vec{x}) = a\vec{x} + \vec{b}$
symm.	*	*	yes
norm.	*	*	no
cons.	*	*	no
def.	*	*	-

*: only apply to endomorphisms