

LPC 23

Linear System's

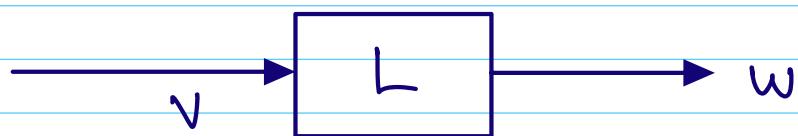
- An APPreciation.

- * we want to see how the Fourier Transform applies to linear systems.
- * The main ideas are about the impulse response and the Transfer Function,
- * Complex exponentials appearing as eigen functions of LTI systems.

BASIC Definition's

So a linear system for us is a method of associating an output to an input that satisfies the principle of superposition.

* A mapping from inputs to outputs
that satisfies the superposition principle.



$$\begin{aligned} \textcircled{1} \quad L(v_1 + v_2) &= L(v_1) + L(v_2) \\ \textcircled{2} \quad L(\alpha v) &= \alpha L(v) \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{Linearity}$$

$$L\left(\sum_{i=1}^n \alpha_i v_i\right) = \sum_{i=1}^n \alpha_i L(v_i)$$

We can extend this idea to infinite sums & integrals, but generally that requires assumptions on the operator L

* We assume some kind of continuity.

Ex: Linear System

* it is in the relationship with direct proportionality.

$$Lv = \alpha v$$

$$\Rightarrow L(v_1 + v_2) = \alpha (v_1 + \alpha v_2) = \alpha (v_1 + Lv_2)$$

* all the linear system's essentially can be understood in terms of direct proportionality.

\Rightarrow all linear system's have been back somehow to the operation of direct proportionality.

\Rightarrow direct proportionality also known as multiplication

\Rightarrow any linear system that is given by multiplication (or any system that is given by multiplication) is a linear system

\Rightarrow a little more generally is a multiplication

\Rightarrow if the input signal is a function of t or x , then we can multiply by another function.

$$\mathcal{L} V(t) = d(t) V(t) \text{ (linear)}$$

Ex:

Switching on and off for a duration a

$$\mathcal{L} V(t) = T_{\alpha}(t) V(t)$$

+ Sampling is a Linear Operation.

$$\mathcal{L} V(t) = V(t) \underline{\Pi}_p(t)$$

$$= \sum_{k=-\infty}^{\infty} v(kp) \delta(t - kp)$$

A slight But important generalization

in ① direct proportion + adding

i.e matrix multiplication

Ansm

$$\Rightarrow \mathbf{w} = \mathbf{Av} \quad \text{where } \mathbf{v} \in \mathbb{R}^m \\ \mathbf{w} \in \mathbb{R}^n$$

$$\textcircled{1} \quad \mathbf{A}(\mathbf{v}_1 + \mathbf{v}_2) = \mathbf{Av}_1 + \mathbf{Av}_2$$

$$\textcircled{2} \quad \mathbf{A} \alpha \mathbf{v}_1 = \alpha \mathbf{Av}_1$$

Ex: EE 263 $\dot{x} = Ax$

$$x(0) = v \quad \text{initial condition}$$

$$\text{then } x(t) = e^{At} v = e^{At} x(0)$$

* Linear system's with special properties

derived from special properties of matrices

① if A is symmetric $A = A^T$

that's a special type of system,

② A is unitary or orthogonal. (A non)

$$AA^* = I \quad QQ^T = I$$

Special linear system

we often look for eigenvalues & eigenvectors of matrix A .

\Rightarrow v is an eigenvector, $v \neq 0$ if

$$Av = \lambda v.$$

Now it may be if we have the whole family of eigenvectors which span the set of all possible inputs (eigenvectors)

are basis for input function)

* if we have eigenvectors

v_1, v_2, \dots, v_n with corresponding

eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$, that form a

Basis for all the inputs. Then we can analyze the action of A easily. Because let's v be any input then, we can write

$$v = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n$$

then A operating on v , by linearity

$$Av = \alpha_1 Av_1 + \alpha_2 Av_2 + \dots + \alpha_n Av_n$$

$$= \sum_{i=1}^n \alpha_i A v_i$$

$$= \sum_{i=1}^n \alpha_i \lambda_i v_i$$

\Rightarrow The action of A on a arbitrary input v is directly proportional
+ adding.

Question is when do linear system's have a basis of eigen vectors?

When can we do this? That's when
these special properties come in.
Orthogonal, Symmetric etc..

- * The spectral theorem in finite dimension for matrices says if A is a Hermitian operator or Symmetric operator (in great case) then it has basis of eigen vectors.
- * When we talked about Fourier series, Complex exponential's forming orthogonal basis for the signal's. And the whole idea of Fourier Transform & diagonalizing the F.T and finding eigen vectors & Eigen function in that case is exactly in matrix form

* These are simple idea's with what we started with) The idea of superposition.
Sum of the input's goes to the sum of the output's. Scaled input's go to the scaled output's.

All these thing's have some analog in continuous case , in infinite dimensional continuous case.

One more property in finite dimensional (matrix multiplication)

* its not just direct proportionality as an example of linear systems , direct proportionality is the only example of linear system.

* It's Not Just That matrix

multiplication in a good (natural)
example of finite dimensional Linear
system. It's the only example.

- * Any Linear Operator on a finite dimensional space can be realized as matrix multiplication.
- * Any finite dimensional (finite number of D.O.F , finite number of describing any input, described by finite set of vector's) Linear system can be realized as matrix multiplication.

(one of the fundamental results of linear algebra)

Example: All polynomial's of degree $\leq n$

input : $a_0 + a_1x + a_2x^2 + \dots + a_nx^n$

a familiar linear operator in differentiation

$L = \frac{d}{dx} \Rightarrow L$ can be described

as $(n+1) \times (n+1)$ matrix.

\Rightarrow Any Linear operator on a finite dimensional space can be described as matrix multiplication.

* and in fact we are going to see that same statement hold's in the infinite dimensional Continuous Case.

* we will see very similar (analogous) statement for infinite dimensional Continuous Case.

* The example that generalizes the matrix multiplication in integration against kernel.

+ Inputs in a function $v(x)$

+ Kernel in a function $k(x,y)$

$$w(x) = L v(x) = \int_{-\infty}^{+\infty} k(x,y) v(y) dy$$

integration against kernel

* the kernel $k(x,y)$ defines the operations define's the linear system, because the integration is linear.

L is Linear System

* we can think of this, infinite dimensional continuous analog of matrix multiplication

It's like to think we somehow an infinite continuous column vector.

$$\int_{-\infty}^{\infty} k(x,y) v(y) dy$$

Column vector

Like a matrix
(a doubly infinite continuous matrix)

x : index of row

y : index of column

Integration is like sum.

What else is true?

* In finite dimensional case there are spectral linear systems, that are characterized by special properties of the matrix

* So too in the infinite dimensional continuous case there are spectral properties of system that are characterized by spectral properties of kernel.

* Special linear systems are solved by special extra assumption's on the kernel.

$$\textcircled{1} \quad \rightarrow k(x,y) = k(y,x) \quad (\text{symmetric kernel})$$

one special property

$$\textcircled{2} \quad k(x,y) = \overline{k(y,x)} \quad (\text{Hermitian property})$$

Example:

\textcircled{1} Fourier transform

$$\mathcal{F}f(s) = \int_{-\infty}^{\infty} e^{-2\pi i st} f(t) dt$$

$$k(s,t) = e^{-2\pi i st} \quad (\text{kernel})$$

\textcircled{2} Convolution.

fix a function $h(t)$

$$L_V(t) = (h * v)(t)$$

$$L v(t) = \int_{-\infty}^t h(t-\tau) v(\tau) d\tau$$

this is spection one , because $h(t-\tau)$, kernel does not depend on separately on t, τ , it depends on its difference.

- * for convolution the kernel = $h(t-\tau)$ depends on $t-\tau$, But not on t, τ separately. (Time invariance property)

* if we shift t, τ by the same amount

$$t \rightarrow t-a$$

$$\tau \rightarrow \tau-a$$

then $t-\tau \rightarrow t-\tau$ (unchanged)

so the convolution is unchanged

- * It's Not Just this is a good idea, it's not just that this is a good example of Linear System's. It is the only example.