

10-2-03

Fourier Analysis

Fourier series $f = \frac{1}{2} \sum \langle f, e_n \rangle e_n$

General orthonormal systems

coefficients; convergence; completeness/uniqueness

Bessel's inequality

Best Approximation Lemma

→ Completeness

Equivalent conditions for completeness

$\text{span} \{g_n\}$

$\overline{\text{span} \{g_n\}}$

Fourier transforms

Fourier inversion

convolution

Smoothness vs. Decay.

Sampling Theorem.

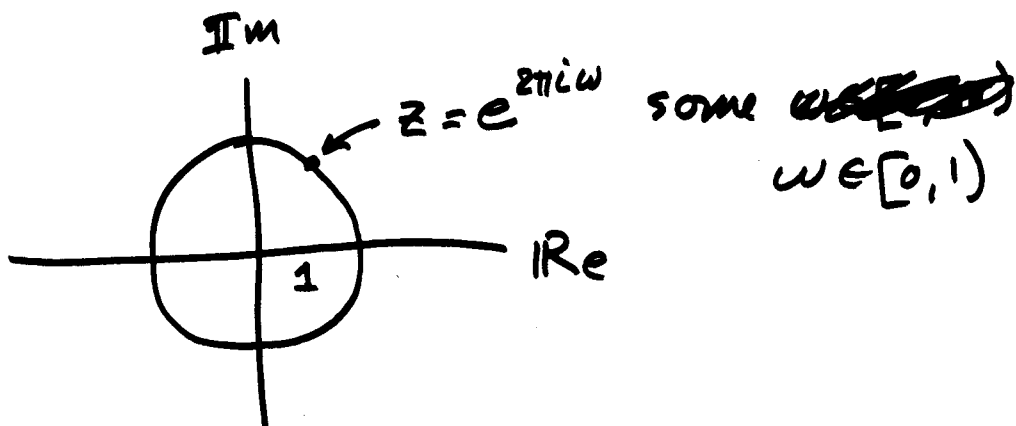
e.g. $x(n) = \frac{1}{n^2}$; $x(0) = 0$

$$\sum_{n \in \mathbb{Z}} \frac{1}{n^2} < \infty$$

e.g. $x(n) = \begin{cases} a^n & n \geq 0 \\ 0 & n < 0 \end{cases}$ some $|a| < 1$

$$\sum_{n \in \mathbb{Z}} x(n) = \sum_{n=0}^{\infty} a^n = \frac{1}{1-a}$$

e.g. $x(n) = \delta(n) = \begin{cases} 1 & n=0 \\ 0 & n \neq 0. \end{cases}$



$$\hat{x}(\omega) = \sum x(n) e^{-2\pi i n \omega} = \mathcal{X}(e^{2\pi i \omega}) \equiv \mathcal{X}(z)$$

where $|z|=1$ in \mathbb{C}

$$\mathcal{X}(z) = \sum_{n \in \mathbb{Z}} x(n) z^{-n} \leftarrow \text{power series in } z^{-1}$$

Power series have a radius of convergence $R > 0$. In our case

$$\sum_{n=0}^{\infty} x(n) z^{-n} \text{ will converge if } |z^{-1}| < R$$

or $|z| > \frac{1}{R}$

$$\text{and } \sum_{n=-\infty}^{-1} x(n) z^{-n} = \sum_{n=1}^{\infty} x(-n) z^n$$

will converge for $|z| < R'$

So $\bar{X}(z)$ will converge on some annulus $\frac{1}{R} < |z| < R'$

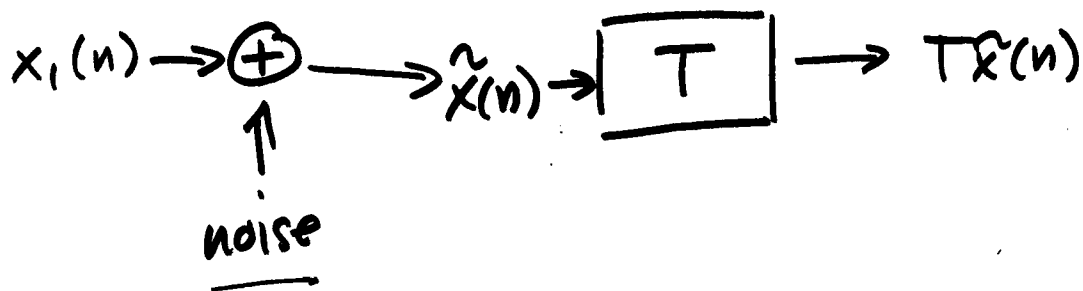
We hope that $\frac{1}{R} < 1 < R'$

Then $\bar{X}(e^{2\pi i \omega})$ makes sense.

$$x(n) \rightarrow \boxed{T} \rightarrow Tx(n)$$

linear

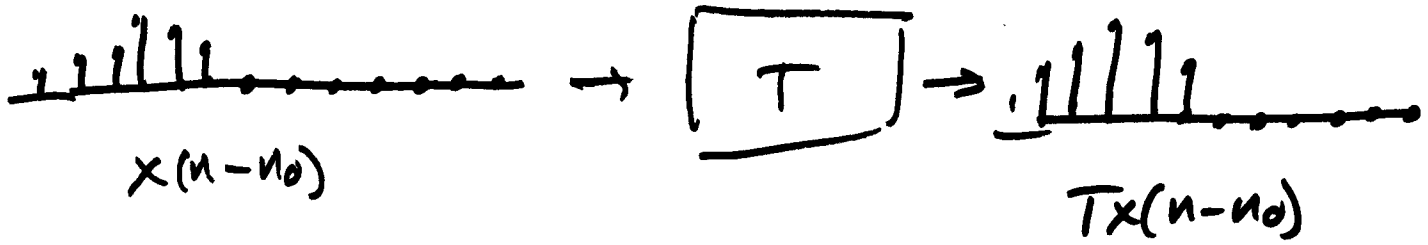
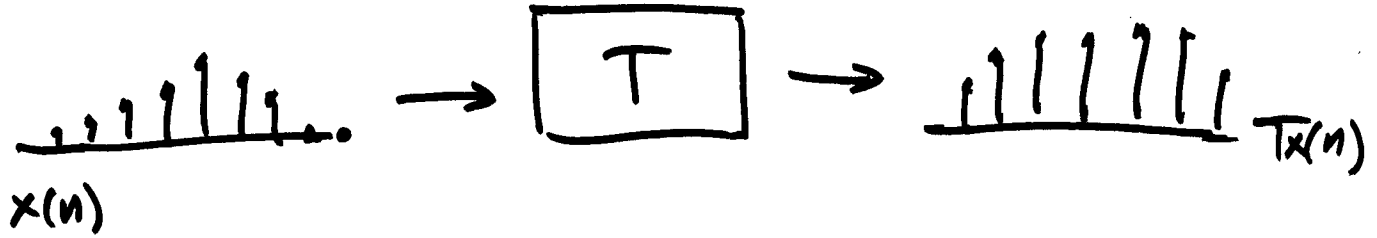
$$\left[\begin{array}{l} x_1(n) + x_2(n) \rightarrow \boxed{T} \rightarrow Tx_1(n) + Tx_2(n) \\ a x_1(n) \rightarrow \boxed{T} \rightarrow a Tx_1(n) \end{array} \right.$$



$$\text{noise intensity} = \sum |\tilde{x}(n) - x(n)|^2$$

$$\begin{aligned} \text{error} &= \sum |T\tilde{x}(n) - Tx(n)|^2 \\ &= \sum |T(\tilde{x} - x)(n)|^2 \end{aligned}$$

$$\leq c \sum |\tilde{x}(n) - x(n)|^2 = \text{noise intensity}$$



PF: (a) $\sum_n |y(n)| = \sum_n |x_1 * x_2(n)|$

$$= \sum_n \left| \sum_k x_1(k) x_2(n-k) \right|$$

$$\leq \sum_n \sum_k |x_1(k)| |x_2(n-k)|$$

$$= \sum_k |x_1(k)| \underbrace{\sum_n |x_2(n-k)|}_{= \sum_n |x_2(n)|}$$

$$= \sum_k |x_1(k)| \sum_n |x_2(n)| < \infty$$

(c) $T_h x(n) = x * h(n)$

stable: $\sum_n |Tx(n)| \leq \underbrace{\sum_n |h(n)|}_C \sum_n |x(n)|$

linear: obvious

Translation invariant: $T_h(T_{n_0} x)(n)$

$$= \sum_k T_{n_0} x(k) h(n-k) = \sum_k x(k-n_0) h(n-k)$$

$$m = k - n_0$$

$$k = m + n_0$$

$$= \sum_m x(m) h(n-m-n_0)$$

$$= \sum_m x(m) h((n-n_0)-m) = T_h x(n-n_0) = T_{n_0} T_h x(n)$$

PR: Let $h(n) = T\delta(n)$, $\delta(n) = \begin{cases} 1 & n=0 \\ 0 & n \neq 0 \end{cases}$

Note that for any $x(n)$

$$x(n) = (x * \delta)(n) = \sum_k x(k) \delta(n-k)$$

$$\begin{aligned} \therefore Tx(n) &= T\left(\sum_k x(k) \delta(n-k)\right) \\ &= \sum_k x(k) T(\tau_k \delta)(n) \\ &= \sum_k x(k) \tau_k (T\delta)(n) \\ &= \sum_k x(k) \tau_k h(n) \\ &= \sum_k x(k) h(n-k) \end{aligned}$$

Why frequency response?

Let $x(n) = e^{2\pi i n \omega}$, i.e. x is a pure frequency.

~~with~~

$$\begin{aligned}Tx(n) &= \sum_k x(k) h(n-k) \\&= \sum_k e^{2\pi i k \omega} h(n-k) \\&= \sum_k e^{2\pi i (n-k) \omega} h(k) \\&= e^{2\pi i n \omega} \sum_k e^{-2\pi i k \omega} h(k) \\&= x(n) \hat{h}(\omega)\end{aligned}$$

Think of $x(n) = \int_0^1 \hat{x}(\omega) e^{2\pi i n \omega} d\omega$

$$\begin{aligned}Tx(n) &= \int_0^1 \hat{x}(\omega) T(e^{2\pi i n \omega}) d\omega \\&= \int_0^1 \hat{x}(\omega) \hat{h}(\omega) e^{2\pi i n \omega} d\omega\end{aligned}$$

$$\therefore \hat{Tx}(\omega) = \hat{x}(\omega) \hat{h}(\omega)$$

PR: $(x * h)(n) = \sum_k x(k) h(n-k)$

$$|(x * h)(n)| = \left| \sum_k x(k) h(n-k) \right|$$

$$\leq \sum_k |x(k)| |h(n-k)|$$

$$\leq \max_{0 \leq k \leq N-1} |x(k)| \sum_k |h(n-k)|$$

$$= \max |x(k)| \sum_k |h(k)| < \infty$$

$$x * h(n+N) = \sum_k x(k) h(n+N-k)$$

$$= \sum_k x(k) h(n - (k-N))$$

$$= \sum_m x(m+N) h(n-m) \quad \begin{array}{l} m = k - N \\ k = m + N \end{array}$$

$$= \sum_m x(m) h(n-m) = x * h(n)$$

$$\hat{x}(\omega) = \sum_k x(k) e^{-2\pi i k \omega} \leftarrow \left(\frac{n}{N}\right)$$

$$\hat{x}(n+N) = \sum_{j=0}^{N-1} x(j) e^{-2\pi i j (n+N)/N}$$

$$= \sum_{j=0}^{N-1} x(j) e^{-2\pi i j n/N} e^{-2\pi i j} \xrightarrow{1}$$

$$= \hat{x}(n)$$

PF: $\frac{1}{N} \sum_{n=0}^{N-1} \hat{x}(n) e^{2\pi i n j/N}$

$$= \frac{1}{N} \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} x(k) e^{-2\pi i n k/N} e^{2\pi i n j/N}$$

$$= \sum_{k=0}^{N-1} x(k) \underbrace{\frac{1}{N} \sum_{n=0}^{N-1} e^{2\pi i n (j-k)/N}}_{= 1 \text{ if } j=k}$$

$$\rightarrow \frac{1}{N} \sum_{n=0}^{N-1} (e^{2\pi i (j-k)/N})^n = \frac{1}{N} \frac{1 - e^{2\pi i (j-k)}}{1 - e^{2\pi i (j-k)/N}} = 0 \text{ if } j \neq k$$

$$= x(j)$$

$$\begin{aligned}
 \underline{\text{PA:}} \quad & \sum_{j=0}^{N-1} x(k) h(j) e^{-2\pi i n j / N} \\
 &= \sum_{j=0}^{N-1} \sum_k x(k) h(j-k) e^{-2\pi i n j / N} \\
 &= \sum_{j=0}^{N-1} \sum_k h(k) x(j-k) e^{-2\pi i n j / N} \\
 &= \sum_k h(k) e^{-2\pi i n k / N} \sum_{j=0}^{N-1} x(j-k) e^{-2\pi i n (j-k) / N} \\
 &= \hat{h}\left(\frac{n}{N}\right) \hat{x}(n)
 \end{aligned}$$

Note: $\hat{h}\left(\frac{n}{N}\right)$ is a period N sequence and is the DFT of something. That something is $\tilde{h}(n) = \sum_k h(n+kN)$,

the N -periodization of h . Another calculation shows that

$$(x * h)(n) = \sum_{k=0}^{N-1} x(k) \tilde{h}(n-k)$$

which inspires the definition of circular convolution.

Rem! Period N signals can be identified with vectors in \mathbb{R}^N .

$$x(n) \longleftrightarrow \begin{bmatrix} x(0) \\ \vdots \\ x(N-1) \end{bmatrix}$$

Circ. convolution ~~is~~ can be thought of as a linear transformation from \mathbb{R}^N to \mathbb{R}^N .

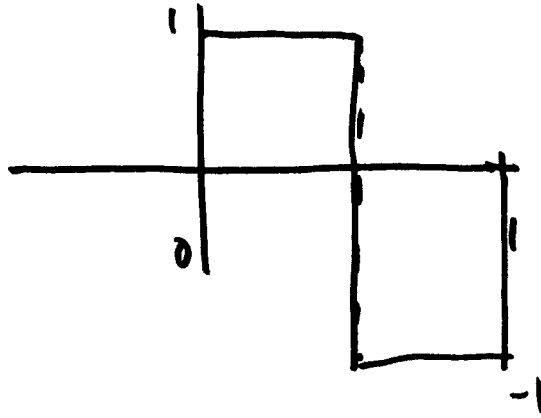
$$\begin{aligned}
 \text{Pf: (a)} \quad \overline{\Sigma(2-k)} &= \sum_{n=1}^N x(n) e^{-2\pi i (2-k-1)(n-1)/N} \\
 &= \sum_{n=1}^N x(n) e^{2\pi i (1-k)(n-1)/N} \\
 &= \sum_{n=1}^N x(n) e^{-2\pi i (k-1)(n-1)/N} = \Sigma(k)
 \end{aligned}$$

Note: $\Sigma(1)$ is always real.

(b) By (a) I know $\Sigma(k) = \overline{\Sigma(2-k)}$.
 So sufficient to show $\overline{\Sigma(k)}$ is real.

$$\begin{aligned}
 \overline{\Sigma(k)} &= \sum_{n=1}^N x(n) e^{-2\pi i (k-1)(n-1)/N} \\
 &= \sum_{n=1}^N x(n) e^{2\pi i (k-1)(n-1)/N} \\
 &\quad \begin{array}{l} m = 2-n \\ n = 2-m \\ n=1 \rightarrow m=1 \\ n=N \rightarrow m=2-N \end{array} \\
 &= \sum_{m=2-N}^1 x(2-m) e^{2\pi i (k-1)(1-m)/N} \\
 &= \sum_{m=1}^N x(m) e^{-2\pi i (k-1)(m-1)/N} \\
 &= \sum_{m=1}^N x(m) e^{-2\pi i (k-1)(m-1)/N} = \Sigma(k)
 \end{aligned}$$

wavelet:



Good

Small support
good localization
in time

fast algorithm
orthonormal.

Bad

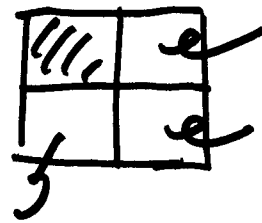
Discontinuous

Smooth fns are
poorly approximated
by ~~fast~~ partial sums
of Haar series.

e.g. Partial reconstructions
of images are blocky

GOAL: Design new
wavelets that have
most of good properties
of Haar but not all
the bad.

MAIN IDEA: Multiresolution
Analysis.



Haar coeffs will "find"
Jump discontinuities but
not corners.