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Basic Definitions

Definition. (Inner Product) An inner product on V is a function that takes each ordered pair (x,y) of elements of V to a number $\langle x,y\rangle\in\mathbb{F}$ and, $\forall x,y,z\in V,\lambda\in\mathbb{F}$, has the following properties:

$$\begin{array}{c} \textbf{Positivity} & \langle y,y\rangle \geq 0 \\ \textbf{Definiteness} & \langle y,y\rangle = 0 \Longleftrightarrow y = 0 \\ \textbf{Additivity} & \langle x+y,z\rangle = \langle x,z\rangle + \langle y,z\rangle \\ \textbf{Homogeneity} & \langle \lambda x,y\rangle = \lambda \langle x,y\rangle \\ \textbf{Conjugate Symmetry} & \langle x,y\rangle = \langle y,x\rangle^* \end{array}$$

Definition. (Norm) Given a vector space V over $\mathbb{F} = \mathbb{R}$ or \mathbb{C} , a *norm* is a mapping $\|\cdot\|$) : $V \to [0, \infty)$ for which

(i)
$$||f|| \ge 0$$
 and $||f|| = 0 \iff f = 0$
(ii) $||\lambda f|| = |\lambda|||f||$
(iii) $||f + g|| \le ||f|| + ||g||$

for all $f,g\in V,\,\lambda\in\mathbb{F}.$ If V is an inner product space, there is a trivial norm:

$$||f|| = \sqrt{\langle f, f \rangle}$$

Normed Spaces

Theorem. (Triangle Inequality)

$$||x + a|| \le ||x|| + ||a||$$

Theorem. (Reverse Triangle Inequality)

$$||x|| - ||a|| \le ||x - a||$$

Example. (Example Normed Spaces)

$$\begin{split} V &= \mathbb{R}^n, \quad \|x\|_2 = \sqrt{\sum_{j=1}^n x_j^2} \\ V &= \mathbb{R}^2, \quad \|x\|_1 = |x_1| + |x_2| \\ V &= \mathbb{R}^2, \quad \|x\|_\infty = \max\{|x_1|, |x_2|\} \\ V &= C([0,1], \mathbb{R}), \quad \|f\|_p = \left(\int_0^1 |f|^p\right)^{\frac{1}{p}} \end{split}$$

Definition. (Open and Closed Balls) With $(X,\|\cdot\|)$ a normed linear space, with $x_0\in X,$ and r>0:

$$\begin{array}{c|c} \textbf{Open Ball} & \text{With centre } x_0 \text{ and radius } r \text{ as} \\ \mathcal{B}(x_0,r) = \{x \in X : \|x-x_0\| < r\} \\ \textbf{Closed Ball} & \text{With centre } x_0 \text{ and radius } r \text{ as} \\ \mathcal{B}[x_0,r] = \{x \in X : \|x-x_0\| \leq r\} \\ \textbf{Sphere} & \text{With centre } x_0 \text{ and radius } r \text{ as} \\ \mathcal{S}(x_0,r) = \{x \in X : \|x-x_0\| = r\} \\ \end{array}$$

Inner Product Spaces

 ${\bf Example.} \ ({\bf Inner-Product\ Space\ Examples})$

• ℓ^p space:

$$\ell^2 = \ell^2(\mathbb{N}, \mathbb{C}) = \left\{ \left\{ c_k \right\}_{k=1}^\infty : \sum_{k=1}^\infty \left| c_k \right|^2 < \infty \right\}$$

And hence for ℓ^2 (the only ℓ^p space that is a Hilbert space)

$$\langle x,y\rangle_{\ell^2}=\sum x_n\overline{y_n}$$

• $L^2(\mathbb{R})$ space:

$$\langle x, y \rangle_{L^2} = \int_{-\pi}^{\infty} f(x) \overline{g(x)} \, \mathrm{d}x$$

Theorem. (Cauchy-Schwarz Inequality) Given that V is an inner-product space over \mathbb{F} , then for all $f,g\in V$:

$$|\langle f, g \rangle| \le ||f|| ||g||$$

When $\{f,g\}$ is a linearly dependent set, we have equality.

Sequences and Convergence

Definition. (Cauchy Sequence) A sequence $\{a_n\}$ of real numbers is a Cauchy Sequence if

$$\forall \varepsilon>0, \exists N>0: m,n>N \Longrightarrow \|a_m-a_n\|<\varepsilon$$

Given that
$$n>N\Longrightarrow \|x_n-x\|<\varepsilon,$$
 then $\lim_{n\to\infty}x_n=x.$

Definition. (Geometric Series) A geometric series (aka "geometric progression") has sequence elements of the form $a_n = ar^n$, where r is the "common ratio". The partial sum is given by:

$$s_k = a \left(\frac{1 - r^k}{1 - r} \right)$$

and so, given that |r| < 1, the partial sum limits to:

$$\lim_{n \to \infty} \sum_{k=1}^{n} ar^k = \frac{a}{1-r}$$

Operator Theory

Proposition. (Operator Continuity) Let X and Y be normed linear spaces and $T: X \to Y$ be a linear mapping. Then, the following are equivalent:

- (i) T is continuous at every point in X
- (ii) T is continuous at 0
- (iii) T is bounded

Hilbert Spaces

Theorem. (Riesz Representation Theorem) Given a Hilbert space $\mathcal H$ over $\mathbb C$ and a linear functional $F:\mathcal H\to\mathbb C$, there exists a unique $y=y_f,$ for $y_t\in\mathcal H$ such that

$$f(x) = \langle x, y_f \rangle, \quad \forall x \in \mathcal{H}$$

Definition. (Adjoint Operator) Let $T: \mathcal{H}_1 \to \mathcal{H}_2$ be a bounded linear mapping between Hilbert spaces. The mapping $T^*: \mathcal{H}_2 \to \mathcal{H}_1$ is called the *adjoint* of the operator T and, for all $x \in \mathcal{H}_1$, $y \in \mathcal{H}_2$, is defined: $\langle Tx, y \rangle_2 = \langle x, T^*y \rangle$.

T is said to be self-adjoint if $T = T^*$.

Definition. (Unitary Operator) Let $\mathcal H$ be a Hilbert space and $U\in \mathcal B(\mathcal H)$. We say that U is unitary if U is invertible and $U^{-1}=U^*$. Equivalently, $UU^*=U^*U=I$.

Definition. (Bounded Linear Operator) Let $T:X\to Y$ be a bounded linear operator between normed linear spaces. The *norm* of T is defined to be:

$$\|T\| = \inf\{M > 0 : \forall x \in X, \|Tx\| \le M\|x\|\}$$

Note that $\|Tx\| \le \|T\| \|x\|$ and $\|T\|$ is the smallest M such that this is smalled

Proposition. Let $T: X \to Y$ be a bounded linear operator between normed linear spaces. Then

$$\|T\| = \sup\{\|Tx\| : \|x\| = 1\}$$

Definition. (Normal Operator) $T \in \mathcal{L}(V)$ is normal if $TT^* = T^*T$. Furthermore, T is normal if and only if $||Tv|| = ||T^*v||$.

Example. (Orthogonal Sets)

$$\begin{split} \mathcal{H} &= \mathbb{C}^N \text{ with } \left\{ f_j(k) = \frac{1}{\sqrt{N}} e^{2\pi i j k/N} \right\}_{j=0}^{N-1} \\ \mathcal{H} &= L^2 \left(\left[-\frac{1}{2}, \frac{1}{2} \right] \right) \text{ with } \left\{ f_j(x) = e^{2\pi i j x} \right\}_{j=-\infty}^{\infty} \\ \mathcal{H} &= L^2(\mathbb{R}) \text{ with } \left\{ f_j(x) = \chi_{[j,j+1]}(x) \right\}_{j=-\infty}^{\infty} \\ \mathcal{H} &= \ell^2(\mathbb{N}) \text{ with } \left\{ e_j(k) = \delta_{ij} \right\}_{j=1}^{\infty} \end{split}$$

Proposition. For $\{x_n\}$ an orthonormal set in \mathcal{H} :

$$\sum_{n=1}^{\infty} \left| \langle x, x_n \rangle \right|^2 \le \|x\|^2$$

Spectral Theory

Theorem. (Normal Operator Eigenvectors) Suppose $T\in\mathcal{L}(V)$ is normal. Then, eigenvectors of T corresponding to distinct eigenvalues are orthogonal.

Theorem. (Real Spectral Theorem) Suppose $\mathbb{F}=\mathbb{R}$ and $T\in\mathcal{L}(V)$. Then, the following are equivalent:

- T is self-adjoint.
- (ii) V has an orthonormal basis consisting of eigenvectors of T.
- (iii) T has a diagonal matrix with respect to some orthonormal basis of V.

Theorem. (Complex Spectral Theorem) Suppose $\mathbb{F}=\mathbb{C}$ and $T\in\mathcal{L}(V)$. Then, the following are equivalent:

- T is normal.
- (ii) V has an orthonormal basis consisting of eigenvectors of T.
- (iii) T has a diagonal matrix with respect to some orthonormal basis of V.

Theorem. (Diagonalisation) A square $n \times n$ matrix, A, with entries in a field \mathbb{F} is called diagonalisable or non-defective if there exists an $n \times n$ invertible matrix (i.e. an element of the general linear group $\mathrm{GL}_n(\mathbb{F})$), P, such that $P^{-1}AP$ is a diagonal matrix.

Given that Λ is diagonal matrix with eigenvalue entries of A on the main diagonal, then $AP=P\Lambda$ implies that P=V, the matrix whose columns are eigenvectors to A, corresponding to the eigenvalue in the same column of Λ . In this case:

$$A=V\Lambda V^{-1}$$

Note that one of the spectral theorems may guarantee the existence of Λ .

Function Spaces

Definition. (Lebesgue Spaces) Given two measurable functions f and g, we say that $f \sim g$ if f = g almost everywhere.

For $1 \le p <$

$$L^p(\mathbb{R}) = \left\{ f: \mathbb{R} \to \mathbb{C}: \|f\|_p = \left(\int |f|^p\right)^{\frac{1}{p}} < \infty \right\} / \sim$$

for $p = \infty$:

$$L^{\infty}(\mathbb{R}) = \{f: \mathbb{R} \to \mathbb{C}: \|f\|_{\infty} = \text{ess sup } |f(x)| < \infty\}/\sim$$

where

ess sup
$$|f| = \inf\{\alpha \in \mathbb{R} \mid \mu\{x \in \mathbb{R} : |f(x)| > \alpha\} = 0\}$$

Inequalities

Theorem. (Arithmetic-Geometric Mean Inequality)

For positive x and y:

$$4xy \le (x+y)^2$$

Notice that by expanding the RHS:

$$2xy \le x^2 + y^2$$

Trigonometry

Definition.

$$\sin z = \frac{e^{iz} - e^{-i.}}{2i}$$

$$\cos z = \frac{e^{iz} + e^{-i.}}{2}$$

$$\sinh z = \frac{e^z - e^{-z}}{2}$$

$$\cosh z = \frac{e^z + e^{-z}}{2}$$

Definition. (Double Angle Formulae)

$$\begin{array}{l} \sin 2\theta = 2 \sin \theta \cos \theta \\ \cos 2\theta = \cos^2 \theta - \sin^2 \theta \\ = 2 \cos^2 \theta - 1 \\ = 1 - 2 \sin^2 \theta \\ \tan 2\theta = \frac{2 \tan \theta}{1 - \tan^2 \theta} \end{array}$$

Definition. (Half Angle Formulae)

$$\sin^2 \theta = \frac{1 - \cos 2\theta}{2}$$

$$\sin\left(\frac{\theta}{2}\right) = \pm \sqrt{\frac{1 - \cos \theta}{2}}$$

$$\cos^2 \theta = \frac{1 + \cos 2\theta}{2}$$

$$\cos\left(\frac{\theta}{2}\right) = \pm \sqrt{\frac{1 + \cos \theta}{2}}$$

$$\tan\left(\frac{\theta}{2}\right) = \pm \sqrt{\frac{1 - \cos \theta}{1 + \cos \theta}}$$

Definition. (Ptolemy's Identities (Difference Formulas))

$$\sin(\alpha \pm \beta) = \sin \alpha \cos \beta \pm \cos \alpha \sin \beta$$
$$\cos(\alpha \pm \beta) = \cos \alpha \cos \beta \mp \sin \alpha \sin \beta$$

Definition. (Trigonometric Integrals)

$$\int \frac{1}{\sin ax} dx = -\frac{1}{a} \ln|\csc ax + \cot ax| + C$$

$$\int \frac{1}{\cos ax} dx = -\frac{1}{a} \ln|\tan\left(\frac{1}{2}ax + \frac{1}{4}\pi\right)| + C$$

$$\int \tan ax dx = -\frac{1}{a} \ln|\cos ax| + C = \frac{1}{a} \ln|\sec ax| + C$$

$$\int \tan^2 ax dx = \tan x - x + C$$

$$\int \sec ax dx = \frac{1}{a} \ln|\sec ax + \tan ax| + C$$

$$= \frac{1}{a} \ln\left|\tan\left(\frac{1}{2}ax + \frac{1}{4}\pi\right)\right| + C$$

$$= \frac{1}{a} \arctan(\sin ax) + C$$

$$\int \frac{1}{a^2 + a^2} dx = \frac{1}{a} \arctan(\frac{\pi}{a}) + C$$

Fourier Series

Definition. (Fourier Series) For $f \in \mathcal{H}$ and let $\{e_n\}_{n=-\infty}^{\infty}$ be an orthonormal basis in \mathcal{H} , the Fourier series of f is

$$f(x) = \sum_{n=1}^{\infty} \langle x, e_n \rangle e_n$$

Theorem. (Parseval Equality) With $f \in \mathcal{H}$ and $\{e_n\}_{n=1}^{\infty}$ an orthonormal basis for \mathcal{H} , then for $\mathcal{F}: \mathcal{H} \to \ell^2(\mathbb{N})$:

$$(\mathcal{F}f)_n = \langle f, e_n \rangle$$

is unitary. Furthermore, we have

$$\langle f,g\rangle_{\mathcal{H}}=\langle \mathcal{F}f,\mathcal{F}g\rangle_{\ell^2}$$

Theorem. (Plancherel Equality) With $f \in \mathcal{H}$ and $\{e_n\}_{n=1}^{\infty}$ an orthonormal basis for $\mathcal{H}:$

$$\|f\|_{\mathcal{H}}^2 = \sum_{n=1}^\infty \left| \langle f, e_n \rangle \right|^2 = \|\mathcal{F} f\|_{\ell^2}^2$$

L^1 Fourier Transform

Definition. (L^1 Fourier Transform) For $f \in L^1(\mathbb{R})$, the Fourier Transform \hat{f} of f is defined:

$$\mathcal{F}f(\xi) = \hat{f}(\xi) = \int_{-\infty}^{\infty} f(t)e^{-2\pi i \xi t} \, \mathrm{d}t \quad (\xi \in \mathbb{R})$$

Proposition. (Properties)

- $\|\hat{f}\|_{-} \le \|f\|_{1}$
- $\hat{f} \in C(\mathbb{R})$
- $\lim_{|\xi|\to\infty} |\hat{f}(\xi)| = 0$
- $\hat{f} \equiv 0 \iff f = 0$ almost everywhere

Theorem. (In the Schwarz Space) For $f \in \mathcal{S}(\mathbb{R})$:

- F: S → S
- If $f \in \mathcal{S}$ then, for all $t \in \mathbb{R}$:

$$f(t) = \int_{-\infty}^{\infty} \hat{f}(\xi)e^{2\pi i \xi t} d\xi$$

• If $f, g \in \mathcal{S}$ then $\langle f, g \rangle = \langle \hat{f}, \hat{g} \rangle$

Remark. (Differentiation)

$$\mathcal{F}Df(\xi) = 2\pi i \xi \mathcal{F}f(\xi)$$
$$\mathcal{F}(-2\pi i t f) = D\mathcal{F}f$$

Example. With $B(t) = \chi_{[-\frac{1}{2}, \frac{1}{2}]}(t)$:

$$\hat{B}(\xi) = \operatorname{sinc}(\xi) = \frac{\sin(\pi \xi)}{\pi^{\xi}}$$

And for $k \in \mathbb{Z}$, $\hat{B}(k) = \delta_{k0}$

L^2 Fourier Transform

Proposition. (Properties) With \mathcal{F} and \mathcal{F}^{-1} defined on $L^2(\mathbb{R})$, then:

- \$\mathcal{F}\$ and \$\mathcal{F}^{-1}\$ are linear operators.
- \mathcal{F} and \mathcal{F}^{-1} are unitary operator on $L^2(\mathbb{R})$, and so for all $f,g\in L^2(\mathbb{R})$

Equivalently, $\mathcal{F}^* = \mathcal{F}^{-1}$.

Convolution

Definition. (Convolution) For $f, g \in L^1(\mathbb{R})$, we define the convolution:

$$f*g(t) = \int_{-\infty}^{\infty} f(s)g(t-s)\,\mathrm{d}s$$

Theorem. If $f, g \in L^1(\mathbb{R})$ then

- $f * g \in L^1(\mathbb{R})$ and $||f * g||_1 \le ||f||_1 ||g||_1$.
- $(\widehat{f * g})(\xi) = \widehat{f}(\xi)\widehat{g}(\xi)$.

Example. Let
$$f=\chi_{[-\frac{1}{2},\frac{1}{2}]}$$
. Then,
$$f*f(x)=g(x)=\max\{1-|x|,0\}$$

$$\hat{g}(\xi)=\hat{f}(\xi)^2=\frac{\sin^2(\pi\xi)}{2\cdot 2\cdot 2}$$

Covariance

Definition. Given $t_0, \xi_0 \in \mathbb{R}$ we define the following operators on $L^2(\mathbb{R})$: $\tau_{t_0} f(t) = f(t - t_0)$

$$M_{\xi_0} f(t) = e^{-2\pi i \xi_0 t} f(t)$$

$$D_a f(t) = \frac{1}{\sqrt{s}} f\left(\frac{t}{s}\right)$$

Proposition. (Properties)

· Each operator has an inverse, as follows:

$$\tau_{t_0}^{-1} = \tau_{-t_0}$$
 $M_{\xi_0}^{-1} = M_{-\xi_0}$ $D_a^{-1} = D_{a^{-1}}$

- · Each operator is unitary.
- · Each operator has a relationship with the Fourier transform:

$$\begin{split} \mathcal{F}\tau_{t_0} &= M_{t_0}\mathcal{F} \\ \mathcal{F}M_{\xi_0} &= \tau_{-\xi_0}\mathcal{F} \\ \mathcal{F}D_a &= D_{a^{-1}}\mathcal{F} \end{split}$$

Sampling

Definition. (Bandlimit) Given $\Omega > 0$, let

$$B_{\Omega} = \left\{ f \in L^2(\mathbb{R}) : \hat{f}(\xi) = 0 \text{ for } \xi > \frac{\Omega}{2} \right\}$$

Functions in B_{Ω} are said to be bandlimited with bandwidth Ω .

Theorem. (Shannon Sampling Theorem) If $f \in B_{\Omega}$, then

$$f(t) = \sum_{n=-\infty}^{\infty} f\left(\frac{n}{\Omega}\right) \operatorname{sinc}(\Omega t - n)$$

where $sinc(x) = \frac{sin(\pi x)}{x}$

Theorem. (Poisson Summation Formula) If $f,\hat{f}\in L^1(\mathbb{R})$ and

are absolutely convergent for all
$$t, \xi \in \mathbb{R}$$
, then
$$\sum_{n=-\infty}^{\infty} f(t+n) = \sum_{n=-\infty}^{\infty} \hat{f}(m)e^{2\pi i mt}$$

$$\sum_{n=-\infty}^{\infty} f(t+n) = \sum_{m=-\infty}^{\infty} \hat{f}(m)e^{2\pi i m}$$

Remark. This can be used to show that certain infinite sums have some particular value.

Uncertainty

Theorem. (Heisenberg Uncertainty Inequality)

$$\int_{-\infty}^{\infty} (t-t_0)^2 |f(t)|^2 \, \mathrm{d}t \cdot \int_{-\infty}^{\infty} (\xi-\xi_0)^2 \Big| \hat{f}(\xi) \Big|^2 \, \mathrm{d}\xi \geq \frac{\|f\|_2^4}{16\pi^2}$$

Time-Frequency Analysis

Definition. (Short-Time Fourier Transform) Let $\varphi \in L^2(\mathbb{R})$. Given

$$\varphi_{b,\xi}(t) = M_{-\xi}\tau_b\varphi(t) = e^{2\pi i \xi t}\varphi(t-b)$$

Then the short-time Fourier transform $S_{\alpha}f$ of f with respect to φ is defined by

$$S_{\varphi}f(b,\xi) = \left\langle f, \varphi_{b,\xi} \right\rangle$$

Remark. The short-time Fourier transform is also known as the sliding window Fourier transform, the windowed Fourier transform and the Gabor transform (especially in the case where the window φ is a Gaussian). Note that

$$S_{\varphi}f(b,\xi) = \int_{-\infty}^{\infty} f(t)e^{-2\pi i \xi t} \overline{\varphi(t-b)} \, \mathrm{d}t = \left(\mathcal{F}f\overline{\varphi(\cdot-b)}\right)\!(\xi)$$

 $\varphi_{b,\xi}$ is obtained from φ by a translation in time by b and in frequency by ξ , i.e., a translation in "phase space" (the time-frequency plane) by (b, ξ) .

Also, if $\varphi \equiv 1$, $S_{\omega}f(b,\xi) = \hat{f}(\xi)$, but in fact since φ is assumed to have decay properties, $S_{co}f$ represents a windowed version of the Fourier transform around t = b. By the Plancherel formula we may write the

$$S_{\varphi}f(b,\xi) = \int_{-\infty}^{\infty} \hat{f}(s)e^{2\pi i b(s-\xi)}\overline{\hat{\varphi}(s-\xi)}\,\mathrm{d}s$$

 $S_{\omega}f(b,\xi)$ represents time-frequency localised information about f, i.e., the information about f near the time b and the frequency ξ .

Theorem. If $f \in L^2(\mathbb{R})$, then $S_n : L^2(\mathbb{R}) \to L^2(\mathbb{R}^2)$ is a multiple of an

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left| S_{\varphi} f(b,\xi) \right|^2 \mathrm{d}b \, \mathrm{d}\xi = \|\varphi\|_2^2 \int_{-\infty}^{\infty} |f(t)|^2 \, \mathrm{d}t = \|\varphi\|_2^2 \|f\|_2^2$$