

Finite Analysis

John M. Dusel

The Setup

The Discrete
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The
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Finite Analysis and the Discrete Fourier Transform

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Let G be a finite abelian group

$$G \simeq \mathbb{Z}_{m_1} \oplus \cdots \oplus \mathbb{Z}_{m_r}.$$

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Definition (The function space $L^2(G)$)

$$L^2(G) = \{f : G \rightarrow \mathbb{C}\}$$

$$\langle f, g \rangle_{L^2(G)} = \sum_{x \in G} f(x) \bar{g}(x)$$

$$\|f\|_{L^2(G)}^2 = \langle f, f \rangle_{L^2(G)}$$

Proposition

$L^2(G)$ is a $|G|$ -dimensional \mathbb{C} -vector space with an orthonormal basis consisting of the delta functions

$$\delta_a(x) = \begin{cases} 1 & \text{if } x = a \\ 0 & \text{if } x \neq a. \end{cases}$$

Proof.

Obvious. □

Definition

For $f, g \in L^2(G)$

$$\tau_s f(x) = f(s + x)$$

$$f * g(x) = \sum_{a \in G} f(a)g(x - a).$$

Convolving with a delta function amounts to translation:

$$f * \delta_a(x) = \sum_{y \in G} f(y)\delta_a(x - y) = f(x - a) = \tau_{-a}f(x)$$

A **character** of G is a group homomorphism from G into the 1-torus (*i.e.* unit circle), $\mathbb{T} \subset \mathbb{C}$, under multiplication.

Definition

The set of all characters of G under pointwise multiplication is called **the dual group** \widehat{G}

$$\widehat{G} = \{\chi \in \text{Hom}_{\text{Grp}}(G, \mathbb{T})\}$$

$$\chi_1 \chi_2(x) = \chi_1(x) \chi_2(x)$$

$$e_{\widehat{G}} = 1(x) \equiv 1$$

$$\chi^{-1}(x) = \overline{\chi}(x) = \chi(-x)$$

Proposition (Orthogonality of characters)

If $\chi, \psi \in \widehat{G}$ then

$$\langle \chi, \psi \rangle = \begin{cases} |G| & \text{if } \chi = \psi \\ 0 & \text{otherwise} \end{cases}$$

Proof.

First observe that $\langle \chi, \psi \rangle = \langle \chi\psi^{-1}, 1 \rangle$, then consider $\langle \chi, 1 \rangle$. \square

Definition (L^2 on the dual group)

The space of functions $L^2(\widehat{G})$ is defined by

$$\begin{aligned}L^2(\widehat{G}) &= \{F : \widehat{G} \rightarrow \mathbb{C}\} \\ \langle F, G \rangle_{L^2(\widehat{G})} &= \sum_{\chi \in \widehat{G}} F(\chi) \overline{G(\chi)} \\ \|F\|_{L^2(\widehat{G})}^2 &= \langle F, F \rangle_{L^2(\widehat{G})}\end{aligned}$$

For $a = (a_1, \dots, a_r), x = (x_1, \dots, x_r) \in G$ there is a character

$$e_a(x) = \prod_{j=1}^r e_{a_j}(x_j)$$

given by componentwise complex exponentiation

$$\mathbb{Z}_{m_j} \ni x_j \xrightarrow{e_{a_j}} \exp\left(\frac{2\pi i a_j}{m_j} x_j\right) \in \mathbb{T}$$

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For all $\xi, x \in G$

$$e_\xi(x) = e_x(\xi)$$

$$\overline{e_\xi(x)} = e_{-\xi}(x) = e_\xi(-x)$$

Definition

The **discrete Fourier transform** (DFT) is the map $L^2(G) \rightarrow L^2(\widehat{G})$ defined by

$$\widehat{f}(\chi) = \langle f, \chi \rangle_{L^2(G)} = \sum_{x \in G} f(x) \overline{\chi(x)}.$$

The **inverse DFT** is defined by

$$f^\vee(\chi) = \sum_{x \in G} f(x) \chi(x).$$

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- For all $a \in G$,

$$\widehat{\delta}_a(\chi) = \sum_{x \in G} \delta_a(x) \overline{\chi}(x) = \overline{\chi}(a)$$

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$$\widehat{\delta}_a(\chi) = \sum_{x \in G} \delta_a(x) \overline{\chi}(x) = \overline{\chi}(a)$$

- $\overline{\widehat{f}} = \check{f}$ and $\widehat{\check{f}} = \overline{f}$, because

$$\overline{\widehat{f}}(\chi) = \overline{\sum_{x \in G} f(x) \overline{\chi}(x)} = \sum_x \overline{f(x) \overline{\chi}(x)} = \sum_x \overline{f(x)} \chi(x) = \check{f}(\chi)$$

and

$$\widehat{\check{f}}(\chi) = \sum_{x \in G} \overline{f(x) \overline{\chi}(x)} = \sum_x \overline{f(x) \overline{\chi}(x)} = \overline{f(x) \overline{\chi}(x)} = \overline{f(x)} \chi(x) = \overline{f}(\chi)$$

Lemma

For all $f, g \in L^2(G)$,

$$\sum_{x \in G} \widehat{f}(x)g(x) = \sum_{x \in G} f(x)\widehat{g}(x).$$

Proof.

Switch the order of summation, re-arrange, then use the property that $e_{-x}(\xi) = e_{-\xi}(x)$. □

Proposition

Let $f, g \in L^2(G)$ and $x \in G$.

- 1 The DFT $\mathcal{F} : L^2(G) \rightarrow L^2(\widehat{G})$ is bijective and linear

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- 3 The DFT is invertible:

$$\frac{1}{|G|}(\widehat{f})^\vee = f = \frac{1}{|G|}(f^\vee)^\wedge$$

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- ④ Plancherel/Parseval:

$$\frac{1}{|G|}\langle f, h \rangle_{L^2(G)} = \langle \hat{f}, \hat{h} \rangle_{L^2(\hat{G})}$$

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- ④ Plancherel/Parseval:

$$\frac{1}{|G|}\langle f, h \rangle_{L^2(G)} = \langle \widehat{f}, \widehat{h} \rangle_{L^2(\widehat{G})}$$

- ⑤ Translation transforms to scalar multiplication: $(\tau_s f)^\wedge(\chi) = \chi(s)\widehat{f}(\chi)$

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- The Laplacian $\Delta = \frac{d^2}{dx^2}$ on \mathbb{R}/\mathbb{Z} can be approximated by an operator known as the “combinatorial Laplacian” on a certain type of graph.

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- Let $f \in C^2(\mathbb{R})$ and suppose we have sampled its values at n points x_0, \dots, x_{n-1} of equal spacing $\delta = \frac{1}{n}$. For simplicity denote $f(x_j) = f(j)$.

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- Let $f \in C^2(\mathbb{R})$ and suppose we have sampled its values at n points x_0, \dots, x_{n-1} of equal spacing $\delta = \frac{1}{n}$. For simplicity denote $f(x_j) = f(j)$.
- Assume in addition that f satisfies periodic boundary conditions; thereby we may regard f as being defined on the “finite circle” \mathbb{Z}_n , interpreting all of f 's arguments mod n .

Definition (Finite Difference Operators)

The “first derivative” is

$$D_1 f(j) = \frac{f(j+1) - f(j)}{\delta},$$

the “second derivative” is

$$D_2 f(j) = D_1(D_1 f(j)) = [n^2(A - 2I)] f(j+1).$$

Where I is the identity operator, $If(j) = f(j)$ and A is the adjacency operator

$$Af(j) = f(j-1) + f(j+1)$$

Definition

The **combinatorial Laplacian** Δ_c is defined by

$$\Delta_c = A - 2I$$

so that

$$D_2 f(j) = n^2 \Delta_c f(j+1).$$

Note $\lim_{n \rightarrow \infty} n^2 \Delta_c = \Delta$, which is to say that $n^2 \Delta_c$ approximates the second derivative on the circle.

- On \mathbb{R}/\mathbb{Z} , Δ has eigenfunctions

$$f_a(x) = \exp(2\pi i a x)$$

for $a \in \mathbb{Z}, x \in \mathbb{R}$ with eigenvalues

$$\lambda_a = -4\pi^2 a^2$$

- On \mathbb{Z}_n we should expect $n^2 \Delta_c$ to have eigenfunctions

$$f_{a,n}(j) = \exp\left(\frac{2\pi i a}{n} j\right)$$

for $a = 0, \dots, n-1$ and $j \in \mathbb{Z}_n$ with eigenvalues

$$\lambda_{a,n} \approx -4\pi^2 a^2$$

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Theorem (Continuous case)

If $f \in C^1[a, b]$ then

$$\int |f(x)|^2 dx \leq (2\pi)^{-2} \int |f'(x)|^2 dx.$$

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Theorem (Discrete case)

If $z \in \mathbb{C}^k$ then

$$\sum_{j=0}^{k-1} |z_j|^2 \leq (2 \sin(\pi/k))^{-2} \sum_{j=0}^{k-1} |z_{j+1} - z_j|^2$$

Theorem (Continuous Versions)

① Let $f \in \mathcal{S}(\mathbb{R})$. For all $b, \beta \in \mathbb{R}$

$$\frac{\|f\|_2^4}{16\pi^2} \leq \|(x - b)f\|_2^2 \|(\xi - \beta)\hat{f}\|_2^2$$

② If $f, \hat{f} \in C_c^\infty(\mathbb{R}^n)$ then $f \equiv 0$

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② If $f, \hat{f} \in C_c^\infty(\mathbb{R}^n)$ then $f \equiv 0$

Theorem (Discrete Version)

If $f : G \rightarrow \mathbb{C}$ is not identically zero then

$$|\text{supp } f| |\text{supp } \hat{f}| \geq |G|.$$

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- 1 The Gaussian $f(x) = e^{-\pi x^2}$ is a well-known example for which equality holds in the continuous version.

It is also proportional to its Fourier transform and so is a kind of fixed point.

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$$\delta_H = \begin{cases} 1 & \text{if } x \in H \\ 0 & \text{if } x \notin H \end{cases}$$

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- 2 In the discrete case, if $H < G$ then equality holds for

$$\delta_H = \begin{cases} 1 & \text{if } x \in H \\ 0 & \text{if } x \notin H \end{cases}$$

- 3 The Legendre symbol from number theory is a fixed point of the DFT that does **not** saturate the uncertainty principle!

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Terras [3] for the DFT. Folland [1], Katznelson [2] for the continuous fourier transform.



Gerald B. Folland, *Real analysis: Modern techniques and their applications*, John Wiley & Sons, 1999.



Yitzhak Katznelson, *An introduction to harmonic analysis*, 3 ed., Cambridge Mathematical Library, Cambridge University Press, 2004.



Audrey Terras, *Fourier analysis on finite groups and applications*, London Mathematical Society Student Texts, Cambridge University Press, 1999.