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Applied Analysis Problems

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1 Sheet 4

1.1 Fourier Series

The Fourier series of a p periodic function f, integrable on $\begin{bmatrix} -\frac{p}{2}, \frac{p}{2} \end{bmatrix}$ is

$$f(x) = \frac{a_0}{2} + \sum_{n=1}^{\infty} \left(a_n \cos(\frac{2\pi nx}{p}) b_n \sin(\frac{2\pi nx}{p}) \right). \tag{1}$$

The coefficients a_n and b_n are called the Fourier coefficients of f and are given by

$$a_n = \frac{2}{p} \int_{-\frac{p}{2}}^{\frac{p}{2}} f(x) \sin(\frac{2\pi nx}{p}) dx, \quad n \ge 0$$
 (2)

$$b_n = \frac{2}{p} \int_{-\frac{p}{2}}^{\frac{p}{2}} f(x) \cos(\frac{2\pi nx}{p}) dx, \quad n \ge 1$$
 (3)

Let us compute the Fourier series of f(t)=t for $t\in[-\frac{1}{2},\frac{1}{2}].$ The Fourier coefficients are

$$a_n = 2 \int_{-\frac{1}{2}}^{\frac{1}{2}} t \cos(2\pi nt) \ dt = 0 \quad \text{(odd: g(-t) = -g(t))},$$
 (4)

$$b_n = 2 \int_{-\frac{1}{2}}^{\frac{1}{2}} t \sin(2\pi nt) \ dt = \tag{5}$$

$$=2\left(-\frac{1}{2\pi n}\cos(2\pi nt)\right|_{-\frac{1}{2}}^{\frac{1}{2}}+\int_{\frac{1}{2}}^{\frac{1}{2}}\frac{1}{2\pi n}\cos(2\pi nt)\ dt\right)=\tag{6}$$

$$= -\frac{1}{\pi n} \left(-\cos(\pi n) + \frac{1}{\pi n} \sin(\pi n) \right) = \frac{\sin(\pi n) - \pi n \cos(\pi n)}{(\pi n)^2}.$$
 (7)

Thereby the Fourier series of f(t) = t is

$$f(t) = \sum_{n=1}^{\infty} \left(\frac{\sin(\pi n) - \pi n \cos(\pi n)}{(\pi n)^2} \right) \sin(2\pi n t) = t$$

$$\tag{8}$$

1.2 Truncation Error

The truncation error of the trigonometric polynomial (Sf_N) of degree N is

$$\sum_{|k|>N} |\hat{f}(k)|^2 = \|f - S_N\|_2^2 = \int_{-\frac{1}{2}}^{\frac{1}{2}} |E_N(t)|^2 dt.$$
 (9)

Computations for N=3 and N=9 were done in python with a integration error of around 10^{-15} , resulting in the overall truncation errors of

$$\sum_{|k|>3} |\hat{f}(k)|^2 = 0.0053,\tag{10}$$

$$\sum_{|k|>9} |\hat{f}(k)|^2 = 0.0143. \tag{11}$$

To achieve $||E_N||_2^2 < 0.1||f||_2^2$, the number of coefficients needed are about 61. This was done using a while loop and evaluating $||E_N||_2^2$ for N until the above condition is met.

1.3 Orthonormal Bases

Here we will go through the most important properties of orthonormal bases. So let $\{b_n\}_{n\in\mathbb{N}}$ be an ONB of a vector space \mathcal{H} , then for every $x\in\mathcal{H}$ we may write

$$x = \sum_{b_n} \langle b_n, x \rangle b_n, \tag{12}$$

and

$$||x||^2 = \sum_{b_n} |\langle b_n, x \rangle|^2.$$
 (13)

For any $x, y \in \mathcal{H}$ we can write the scalar product as

$$\langle x, y \rangle = \sum_{b_n} \langle b_n, x \rangle \langle b_n, y \rangle,$$
 (14)

Furthermore there exists a linear projection $\Phi: \mathcal{H} \to l^2(\{b_n\}_n)$ such that

$$\langle \Phi(x), \Phi(y) \rangle = \langle x, y \rangle \quad \forall x, y \in \mathcal{H}.$$
 (15)

An example of an orthonormal basis, which spans $L^2([-\frac{p}{2},\frac{p}{2}])$ is $\mathcal{T}_p = \{e_n := \frac{e^{2\pi i \frac{n}{\mu}x}}{\sqrt{p}}\}_{n \in \mathbb{Z}}$. The e_n 's are orthonormal in L^2 which can be easily seen by using the scalar product of L^2 , so for $n,m \in \mathbb{Z}$

$$\langle e_n, e_m \rangle_{L^2([-\frac{p}{2}, \frac{p}{2}))} = \frac{1}{p} \int_{[-\frac{p}{2}, \frac{p}{2}]} e_n \cdot e_m^* dx =$$
 (16)

$$=\frac{1}{p}\int_{\left[-\frac{p}{2},\frac{p}{2}\right]}e^{2\pi i\frac{(n-m)}{p}x}\,dx=\tag{17}$$

$$=\frac{\sin(\pi(n-m))}{\pi(n-m)} = \begin{cases} 0 & n \neq m \\ 1 & n = m \end{cases}$$
 (18)

1.4 Dirichlet Kernel

The function

$$D_t(x) := \sum_{\|k\|_{\infty} \le t} e_k(x), \quad x \in \mathbb{R}^d$$
(19)

is called the Dirichlet Kernel. For $0 < t \in \mathbb{N}$ we have

$$(S_t f)(x) = \int_{I^d} f(y) D_t(x - y) dy, \tag{20}$$

where S_t represents the orthogonal projection onto the trigonometric polynomials Π_t of degree t, by

$$S_t: L^1(\mathbb{T}^d) \to \Pi_t$$
 (21)

$$f \mapsto \sum_{\|k\| \le t} \langle f, e_k \rangle_{L^2(\mathbb{T}^d)} e_k \quad k \in \mathbb{Z}^d$$
 (22)

And furthermore the Dirichlet Kernel satisfies

$$D_t(x) = \prod_{i=1}^d \frac{e_{t+1}(x_i) - e_{-t}(x_i)}{e_1(x_i) - 1}$$
(23)

To show the convolution property, we start off by applying the orthogonal projection into the trigonometric polynomials S_t onto a function $f \in L(\mathbb{T}^d)$

$$(S_t f) = \sum_{|k|_{\infty} \le t} \int_{I^d} f(y) e^{-2\pi i \langle k, y \rangle} dy \ e^{2\pi i \langle k, x \rangle} =$$
 (24)

$$= \int_{I^d} f(y) \sum_{|k|_{\infty} < t} e^{2\pi i \langle k, (x-y) \rangle} dy =$$
 (25)

$$= (f * D_t)(x) = \int_{I^d} f(y)D_t(x - y) \ dy.$$
 (26)

To show the reformulation of the Dirichlet kernel, we need to simply calculate it directly

$$\sum_{\|k\|_{\infty} \le t} e^{2\pi i \langle k, x \rangle} = \prod_{j=1}^{d} \sum_{k_j = -t}^{t} e^{2\pi i k_j x_j} =$$
(27)

$$= \prod_{j=1}^{d} e^{-2\pi i t x_j} \sum_{k_j=0}^{2t} e^{2\pi i k_j x_j} =; \quad \text{(trigonometric series)}$$
 (28)

$$=\prod_{j=1}^{d} e^{-2\pi i t x_j} \frac{e^{2\pi i (2t+1)x_j} - 1}{e^{2\pi i x_j} - 1} =$$
(29)

$$= \prod_{j=1} \frac{e_{t+1}(x_j) - e_{-t}(x_j)}{e_1(x_j) - 1}.$$
(30)