## Mathematical Tripos Part II: Michaelmas Term 2014

## Numerical Analysis – Lecture 12

**Method 3.8 (The algebra of Fourier expansions)** Let A be the set of all functions  $f : [-1,1] \to \mathbb{C}$ , which are analytic in [-1,1], periodic with period 2, and that can be extended analytically into the complex plane. Then A is a linear space, i.e.,  $f, g \in A$  and  $a \in \mathbb{C}$  then  $f + g \in A$  and  $af \in A$ . In particular, with f and g expressed in its Fourier series, i.e.,

$$f(x) = \sum_{n = -\infty}^{\infty} \hat{f}_n e^{i\pi nx}, \quad g(x) = \sum_{n = -\infty}^{\infty} \hat{g}_n e^{i\pi nx}$$

we have

$$f(x) + g(x) = \sum_{n = -\infty}^{\infty} (\hat{f}_n + \hat{g}_n) e^{i\pi nx}, \quad a \cdot f(x) = \sum_{n = -\infty}^{\infty} a \hat{f}_n e^{i\pi nx}$$
(3.3)

and

$$f(x) \cdot g(x) = \sum_{n=-\infty}^{\infty} \left( \sum_{m=-\infty}^{\infty} \hat{f}_{n-m} \hat{g}_m \right) e^{i\pi nx} = \sum_{n=-\infty}^{\infty} \left( \hat{f} * \hat{g} \right)_n e^{i\pi nx},$$
(3.4)

where \* denotes the convolution operator, hence  $f(x) \cdot g(x) = (\hat{f} * \hat{g})$ , where  $\check{}$  denotes the inverse Fourier transform. Moreover, if  $f \in A$  then  $f' \in A$  and

$$f'(x) = i\pi \sum_{n = -\infty}^{\infty} n \cdot \hat{f}_n e^{i\pi nx}.$$
(3.5)

Since  $\{\hat{f}_n\}$  decays faster than  $\mathcal{O}(n^{-m})$  for all  $m \in \mathbb{Z}_+$ , this provides that all derivatives of f have rapidly convergent Fourier expansions.

**Example 3.9 (Application to differential equations)** Consider the two-point boundary value problem: y = y(x),  $-1 \le x \le 1$  solves

$$y'' + a(x)y' + b(x)y = f(x), \quad y(-1) = y(1),$$
(3.6)

where  $a, b, f \in A$  and we seek a *periodic solution*  $y \in A$  for (3.6). Substituting y, a, b and f by its Fourier series and using (3.3)-(3.5) we obtain an infinite dimensional system of linear equations for the Fourier coefficients  $\hat{y}_n$ :

$$-\pi^{2}n^{2}\hat{y}_{n} + i\pi\sum_{m=-\infty}^{\infty}m\hat{a}_{n-m}\hat{y}_{m} + \sum_{m=-\infty}^{\infty}\hat{b}_{n-m}\hat{y}_{m} = \hat{f}_{n}, \quad n \in \mathbb{Z}.$$
 (3.7)

Since a, b and  $f \in A$  their Fourier coefficients decrease rapidly, like  $O(n^{-m})$  for every m = 0, 1, 2, ... Hence, we can truncate (3.7) into the *N*-dimensional system

$$-\pi^2 n^2 \hat{y}_n + i\pi \sum_{m=-N/2+1}^{N/2} m \hat{a}_{n-m} \hat{y}_m + \sum_{m=-N/2+1}^{N/2} \hat{b}_{n-m} \hat{y}_m = \hat{f}_n, \quad \text{for } n = -N/2 + 1, \dots, N/2.$$
(3.8)

**Remark 3.10** The matrix of (3.8) is in general dense, but our theory predicts that fairly small values of *N*, hence very small matrices, are sufficient for high accuracy. For instance: choosing  $a(x) = f(x) = \cos \pi x$ ,  $b(x) = \sin 2\pi x$  (which incidentally even leads to a sparse matrix) we get

$$N = 16$$
error of size  $10^{-10}$  $N = 22$ error of size  $10^{-15}$  (which is already hitting the accuracy of computer arithmetic)

## Method 3.11 (Computation of Fourier coefficients (DFT)) We have to compute

$$\hat{f}_n = \frac{1}{2} \int_{-1}^{1} f(\tau) e^{-i\pi n\tau} d\tau, \quad n \in \mathbb{Z}.$$

For this, suppose we wish to compute the integral on [-1,1] of a function  $h \in A$  by means of Riemann sums

$$\int_{-1}^{1} h(\tau) \, d\tau \approx \frac{2}{N} \sum_{k=-N/2+1}^{N/2} h\left(\frac{2k}{N}\right). \tag{3.9}$$

We want to know how good this approximation is. As in Method 1.18, let  $\omega_N = e^{2\pi i/N}$ . Then we have

$$\frac{2}{N}\sum_{k=-N/2+1}^{N/2} h\left(\frac{2k}{N}\right) = \frac{2}{N}\sum_{k=-N/2+1}^{N/2} \sum_{n=-\infty}^{\infty} \hat{h}_n e^{2\pi i n k/N} = \frac{2}{N}\sum_{n=-\infty}^{\infty} \hat{h}_n \sum_{k=-N/2+1}^{N/2} \omega_N^{nk}$$
$$= \frac{2}{N}\sum_{n=-\infty}^{\infty} \hat{h}_n \sum_{k=0}^{N-1} \omega_N^{n(k+1-N/2)} = \frac{2}{N}\sum_{n=-\infty}^{\infty} \hat{h}_n \omega_N^{-n(N/2-1)} \sum_{k=0}^{N-1} \omega_N^{nk}$$

Since  $\omega_N^N = 1$  we have

$$\sum_{k=0}^{N-1} \omega_N^{kn} = \begin{cases} N, & n \equiv 0 \mod N \\ 0, & n \not\equiv 0 \mod N. \end{cases}$$

Moreover, for  $n \equiv 0 \mod N$  also  $\omega_N^{-n(N/2-1)} = 1$  and we deduce that

$$\frac{2}{N}\sum_{k=-N/2+1}^{N/2}h\left(\frac{2k}{N}\right) = 2\sum_{r=-\infty}^{\infty}\hat{h}_{Nr}.$$

Hence, the error committed by the Riemann approximation is

$$\frac{2}{N}\sum_{k=-N/2+1}^{N/2}h\left(\frac{2k}{N}\right) - \int_{-1}^{1}h(\tau)\,d\tau = 2\sum_{r=-\infty}^{\infty}\hat{h}_{Nr} - \int_{-1}^{1}h(\tau)\,d\tau = 2\sum_{r=-\infty}^{\infty}\hat{h}_{Nr} - 2\hat{h}_{0} = 2\sum_{r=1}^{\infty}\hat{h}_{Nr} + \hat{h}_{-Nr}$$

Since  $h \in A$ , its Fourier coefficients decay at spectral rate and hence the error of (3.9) decays spectrally as a function of N.

For  $h(x) = f(x)e^{-i\pi mx}$  we obtain a spectral method for calculating the Fourier coefficients of f:

$$\hat{f}_n \approx \frac{1}{N} \sum_{k=-N/2+1}^{N/2} f\left(\frac{2k}{N}\right) \omega_N^{-nk}, \quad n = -N/2 + 1, \dots, N/2,$$

where the sequence on the right-hand side is called the *discrete Fourier transform (DFT)* of *f*.

**Revision 3.12 (The fast Fourier transform (FFT))** The *fast Fourier transform (FFT)* is a computational algorithm, which computes the leading N Fourier coefficients of a function in just  $O(N \log_2 N)$  operations (cf. Algorithm 1.19). We assume that N is a power of 2, i.e.  $N = 2m = 2^p$ , and for  $y \in \Pi_{2m}$ , denote by

$$y^{(E)} = \{y_{2j}\}_{j \in \mathbb{Z}}$$
 and  $y^{(O)} = \{y_{2j+1}\}_{j \in \mathbb{Z}}$ 

the even and odd portions of y, respectively. Note that  $y^{(E)}, y^{(O)} \in \Pi_m$ . To execute FFT, we start from vectors of unit length and in each *s*-th stage, s = 1...p, assemble  $2^{p-s}$  vectors of length  $2^s$  from vectors of length  $2^{s-1}$  with

$$x_{\ell} = x_{\ell}^{(\mathrm{E})} + \omega_{2^{s}}^{\ell} x_{\ell}^{(\mathrm{O})}, \qquad \ell = 0, \dots, 2^{s-1}.$$
 (3.10)

Therefore, it costs just *s* products to evaluate the first half of *x*, provided that  $x^{(E)}$  and  $x^{(O)}$  are known. It actually costs nothing to evaluate the second half, since

$$x_{2^{s-1}+\ell} = x_{\ell}^{(\mathrm{E})} - \omega_{2^s}^{\ell} x_{\ell}^{(\mathrm{O})}, \qquad l = 0, \dots, 2^{s-1} - 1.$$

Altogether, the cost of FFT is  $p2^{p-1} = \frac{1}{2}N \log_2 N$  products.