## **About Fourier Interpolation and Differentiation**

Consider an interval [-A,A] and a 2A-periodic function f to be approximated over the interval by a trigonometric polynomial of degree N

$$p(x) = \sum_{k=-N/2}^{N/2-1} p_k v_k(x), v_k(x) = e^{i\frac{k\pi}{A}x}, k = -N/2, -N/2 + 1, ..., 0, 1, ..., N/2 - 1$$

 $\{v_k\}$  is an orthogonal, linearly independent set of functions:

$$\frac{1}{2A} \int_{-A}^{A} v_k(x) \overline{v}_m(x) dx = \delta_{km}$$

(note the complex conjugate). They are also an orthogonal basis for the gridpoint values

$$f(x_m), m = -N/2,..., N/2-1, x_m = \frac{m}{N}2A$$

$$\frac{1}{N} \sum_{j=-N/2}^{N/2-1} v_k(x_j) \bar{v}_m(x_j) = \delta_{km}$$

so the gridpoint values of any 2A-periodic function f admit a representation

$$f(x_m) = p_N(x_m) = \sum_{k=-N/2}^{N/2-1} \hat{f}_k e^{i\frac{k\pi}{A}x_m} = \sum_{k=-N/2}^{N/2-1} \hat{f}_k e^{i\frac{km\pi}{N}}, \hat{f}_k = \frac{1}{N} \sum_{m=-N/2}^{N/2-1} f(x_m) e^{-i\frac{km\pi}{N}}$$

 $p_N$  can be evaluated for any x (just replace  $x_m$  by x) and is the interpolating trigonometric polynomial. The coefficients can be evaluated by the Discrete Fourier Transform.

Matlab's FFT, according to the documentation,

- Computes *N* times the fhat coefficients numbered 0 to *N*-1 (i.e., 1 to *N* with Matlab indexing which starts at 1),
- IFFT divides by N
- the x-values are 0,h,2h,...,(N-1)h, h=1/N.

For length N input vector  $\mathbf{x}$ , the DFT is a length N vector  $\mathbf{X}$ , with elements

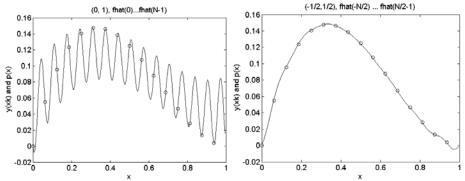
$$N$$
 $X(k) = sum x(n)*exp(-j*2*pi*(k-1)*(n-1)/N), 1 <= k <= N.$ 
 $n=1$ 

The inverse DFT (computed by IFFT) is given by

$$x(n) = (1/N) \text{ sum } X(k) * exp( j*2*pi*(k-1)*(n-1)/N), 1 <= n <= N.$$
 $k=1$ 

**Clearly**, it is necessary to experiment here and understand how to use Matlabs FFT for differentiation. Choose  $f(x) = x(1-x)^2$  so that f(0) = f(1) = 0, but f' and f'' do not match.

First, use the Matlab coefficients and basis functions  $v_k = e^{i2\pi kx}$ , k = 0,1,...,N-1, the interval [0,1) and data points  $x_m = m/N$ , m = 0,1,2,...,N-1. The result is in Fig. 1, left. To the right is the result with  $v_k$ , k = -N/2,...,N/2-1, interval [-1/2,1/2), data points  $x_m = (m-N/2)/N$ , m = 0,1,2,...,N-1

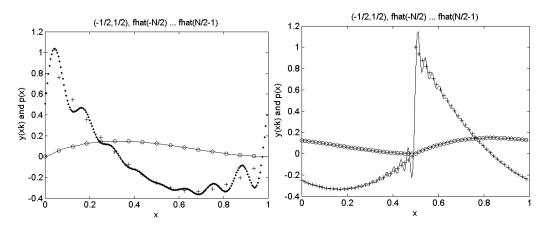


The two trigonometric polynomials behave VERY differently between grid-points. But in fact, they differ only by a multiple of the lowest degree exp. function which vanishes at *all* 

gridpoints, 
$$c_N \sin(\frac{x-a}{b-a}N\pi)$$

**Exercise:** Do the math and find  $c_N$ !

**Clearly**, the symmetric variant looks better, so we use that for differentiation:



The crosses are the exact values of the derivative. The approximation is good in the central parts of the interval, but poor at the end-points. The reason is that f has discontinuous derivatives of order 1 and 2 at x = 0 (and 1): The Gibbs phenomenon. To the right is a plot with 64 data points, and the data shifted 1/2 period so the jump is at x = 0.5. Of course, the interpolation error is also large at x = 0.5, but the discrepancy is better seen in the derivative which is discontinuous.

Here is an approximation to spectral computation of the n:th derivative, using the symmetric variant:

```
function yd = ffd(y,a,b,n)
% Compute yd = n:th derivative of y(x)
% defined by its values y at (a,a+h,...,a+(N-1)h), a+Nh = b
N = length(y);
yhat = fftshift(fft(y));
d = 2*pi*1i*(-N/2:N/2-1)/(b-a);
yd = ifft(fftshift(d.^n.*yhat));
```

**Exercise** Explain the **d**-vector used as differentiation operator in wavenumber space.

**Note:** For the tasks in HW4 it may be more convenient to use the ingredients of **ffd** than the whole code itself.