

Chapter 7 - Fast Fourier Transformation

#### DFT in a nutshell



Standard FT (without normalization):

$$f(k) = \int_{-\infty}^{\infty} e^{-ikx} f(x) dx$$

■ FT of periodic functions of period *L*: (In numerical representation we are limitied to an interval, *f* has to be quasi-periodic)

$$f(k_n) = \int_0^L e^{-ik_n x} f(x) dx$$

Sampling of f on a (discrete) grid yields:

$$f(k_n) = \Delta x \sum_{j=0}^{N-1} e^{-ik_n j \Delta x} f(x_j)$$

- N grid points to cover interval of length L results in two limits:
  - Due to the finite length L there is a lowest wave-number  $k_{min} = \Delta k = 2\pi/L$
  - Sampling Theorem: To uniquely identify an oscillation we need two sampling points per period, i.e.  $\lambda_{min}=2\Delta x$ , thus  $k_{max}=\pi/\Delta x$

### DFT in a nutshell



$$f(k_n) = \Delta x \sum_{j=0}^{N-1} e^{-ik_n j \Delta x} f(x_j) \longrightarrow f(k_n) = \Delta x \sum_{j=0}^{N-1} e^{-i\frac{2\pi}{N\Delta x} n j \Delta x} f_j$$
$$k_n = \frac{2\pi}{N\Delta x} n, \quad n = 0, \dots, N-1$$

- Forward DFT:  $f(k_n) = \Delta x \sum_{j=0}^{N-1} e^{-2\pi i n j/N} f_j$
- This is nothing but a projection of  $f(x_i)$  on a new basis:

$$\hat{\mathbf{f}} = \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \dots & \omega^{N-1} \\ 1 & \omega^2 & \omega^4 & \dots & \omega^{2(N-1)} \\ \vdots & & & & & \\ 1 & \omega^{N-1} & \omega^{2(N-1)} & \dots & \omega^{(N-1)^2} \end{pmatrix} \begin{pmatrix} f(x_0) \\ f(x_1) \\ f(x_2) \\ \vdots \\ f(x_{N-1}) \end{pmatrix}$$

$$\omega = e^{-2\pi i/N}$$

■ Fast Fourier Transformation is a clever way to compute this matrix-vector multiplication in  $\mathcal{O}(N\log(N))$  instead of  $\mathcal{O}(N^2)$ 

### DFT in a nutshell

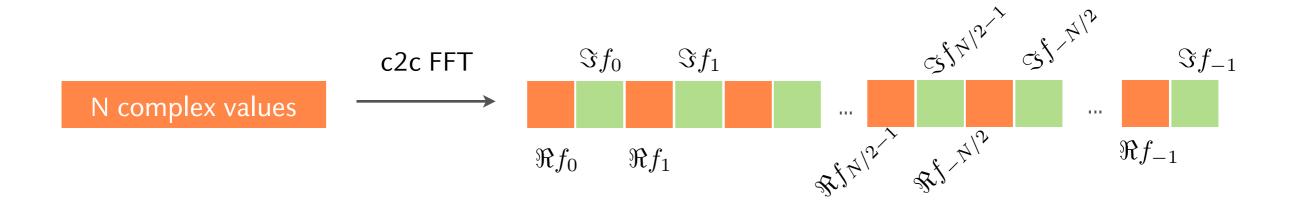


$$f(k_n) = \Delta x \sum_{j=0}^{N-1} e^{-2\pi i n j/N} f_j$$
 is periodic with period N:  $f(k_{N-n}) = f(k_{-n})$ 

- There are two ways in which we can identify the frequencies:
  - All frequencies are positive, ranging from  $k_{min}=0$  to  $k_{max}=k_{N-1}$
  - There are negative and positive frequencies, symmetric about 0:  $k_{min} = k_{-N/2}$  to  $k_{max} = k_{N/2-1}$

# Frequency mapping





N complex values = 2N doubles

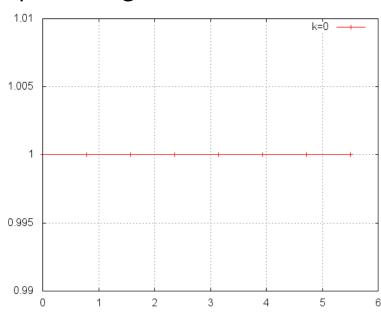
## Example: N = 8

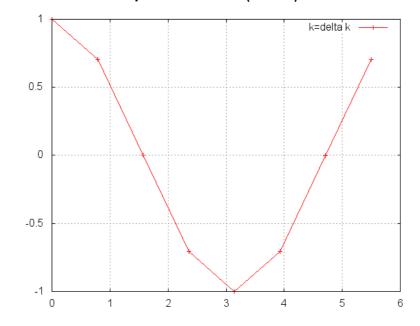


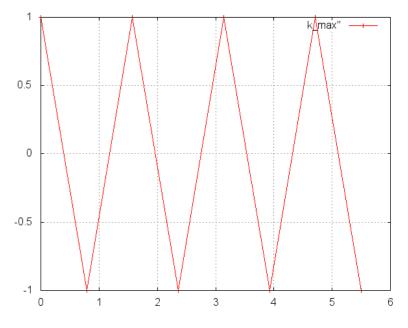
$$I = [0, 2\pi], \quad L = 2\pi, \quad N = 8, \quad \Delta k = \frac{2\pi}{L} = 1, \quad k_{max} = \frac{\pi}{\Delta x}$$

$$\Delta x = \frac{L}{N} = \frac{2\pi}{8} \longrightarrow x_{max} = 7\Delta x = \frac{7}{8}2\pi \neq 2\pi$$

periodic grid: we do not store the redundant point at f(x=L)!







	<u>Re</u>	<u>lm</u>	
$f_0$	8	0	
	0	0	
	0	0	
	0	0	
	0	0	
	0	0	
	0	0	
	0	0	

	<u>Re</u>	<u>lm</u>
	0	0
	0	0
	0	0
<b>a</b>	0	0
$f_{-N/2}$	8	0
	0	0
	0	0
	0	0

# Using FFTW to caclulate the DFT



- FFTW provides three interfaces:
  - Basic interface (we will use this one)
  - Advanced interface
  - Guru interface
- Results from FFTW are usually without normalization, i.e. one might need to normalize results by 1/N to obtain agreement with analytical results

#### C2C transformation with FFTW



```
#include <fftw3.h>
#include <cmath>
#include <iostream>
//----
using namespace std;
//----
int main()
    const int N=8;
   fftw_plan fw,fw2;
   fftw_complex* inC = (fftw_complex*) fftw_malloc(sizeof(fftw_complex) * N);
   fftw_complex* outC = (fftw_complex*) fftw_malloc(sizeof(fftw_complex) * N);
   const double L=2*M_PI;
   const double dx=L/N;
   double x;
   // inC = exp(ii*x)
   for(int i=0; i<N; i++){
       x=i*dx;
    inC[i][0] = cos(x); // Re
    inC[i][1]= sin(x); // Im
   fw = fftw_plan_dft_1d(N, inC, outC, FFTW_FORWARD, FFTW_ESTIMATE);
   fftw_execute(fw);
   (...)
```