Discrete Fourier Transform and filters

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Lecture 25

DFT vs Matlab FFT

DFT

$$y_k = \frac{1}{N} \sum_{n=0}^{N-1} c_n \exp(i\frac{2\pi(k-1)n}{N})$$
 inverse Fourier transform
$$c_n = \sum_{k=1}^{N} y_k \exp(-i\frac{2\pi(k-1)n}{N})$$
 Fourier transform
$$n = 0, 1, 2, \dots, N-1$$

Matlab FFT

$$y_k = \frac{1}{N} \sum_{n=1}^{N} c_n \exp(i\frac{2\pi(k-1)(n-1)}{N})$$
 inverse Fourier transform
$$c_n = \sum_{k=1}^{N} y_k \exp(-i\frac{2\pi(k-1)(n-1)}{N})$$
 Fourier transform
$$n = 1, 2, \dots, N$$

So do DFT \rightarrow Matlab FFT is equivalent of $n \rightarrow n + 1$ and vice versa $\sim \sim$

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Warning about notation

Since c_0 has a special meaning of a DC component of the signal. I will always use the DFT notation unless mentioned otherwise. People often denote the forward Fourier transform as \mathcal{F} so

$$Y = \mathcal{F}y$$

So Y is the spectrum of the signal y Inverse Fourier transform is denoted as \mathcal{F}^{-1}

$$y = \mathcal{F}^{-1} Y$$

Instead of using c_n coefficient we refer in this notation to Y_n

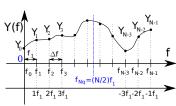
Sampling rate and important physics relationship

Since for DFT we need to have equidistant points and signal repeats itself. We consider signals which start at time 0 and take N points. To deduce the time of the data point we just multiply it's index by the time spacing Δt .

Time series

$y_0 - y_N$ y_1 y_2 y_3 y_4 y_5 y_6 y_{N-1} y_1 y_{N+1} y_{N+1}

Spectrum



Sampling rate is defined as $f_s = 1/\Delta t = f_1 N$ and period $T = N\Delta t$. y_i is taken at time $t_i = i\Delta t = i/f_s$, $y_{i+N} = y_i$. Y_n has the frequency $f_n = f_1(n-1) = f_s(n-1)/N$, $Y_{i+N} = Y_i$.

Nyquist frequency

Provided that we have N data point taken with sampling rate f_s what is the maximum frequency which we can expect to see in our spectrum? Naively, we can say $(N-1)*f_1\approx f_s$ since in spectrum all points are separated by fundamental frequency $f_1=1/T=f_s/N$ However recall that $Y_{N-n}=Y_{-n}$ i.e the higher half of the vector Y contains negative frequency. So at max we can hope to obtain spectrum with the highest frequency smaller than

Nyquist frequency

$$F_{Nq}=f_1\frac{N}{2}=\frac{f_s}{2}$$

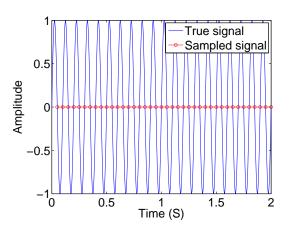
Nyquist criteria

$$f_s > 2f_{signal}$$

You must sample your signal faster than twice the highest frequency component of it. I.e. Nyquist frequency of you sample should be > than the highest signal frequency.

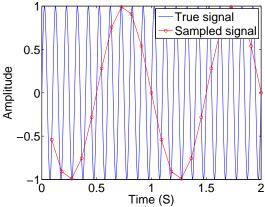
Aliasing: wrong/slow sampling frequency

Sampling with $f_s=2f_{signal}$ i.e. $f_{Nq}=f_{signal}$ Sampled signal appeared to be DC



Aliasing: too slow sampling frequency - reflection

Under sampling $f_s = 1.1 f_{signal}$ Sampled signal seems to be lower frequency.

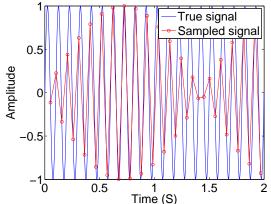


This is case of reflection/folding when frequency higher than the Nyquist frequency appears to be negative and slower one.

$$f_{signal} \rightarrow (f_{signal} - 2f_{Nq})$$

Aliasing: too slow sampling frequency - ghosts

Under sampling $f_s = 1.93 f_{signal}$ Sampled signal seems to be very different



This is also a case of reflection/folding when frequency higher than the Nyquist frequency appears to be negative and slower one.

DFT filters

Once you get a signal you can filter unwanted components out of it. The recipe is the following

- sample the signal
- calculate FT (fft)
- have a look at the spectrum and decide which components are unwanted
- apply filter which attenuate unwanted frequency component (remember that if you attenuate the component of the frequency f by g_f you need to attenuate the component at -f by g_f^* .
- calculate inverse FT (ifft) of the filtered spectrum
- repeat if needed

Applications

- Noise reduction
- Compression