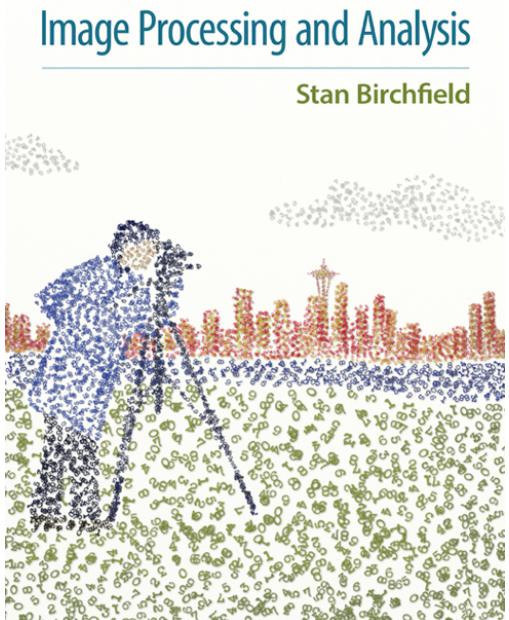


Prof. Kjersti Engan

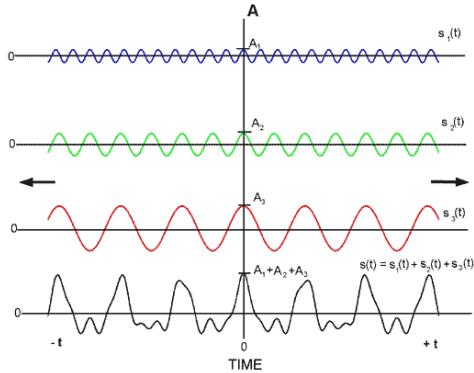
ELE510 Image processing and computer vision

Frequency Domain processing, Fourier transform (Chap 6.1-6.2 Birchfield) 2020

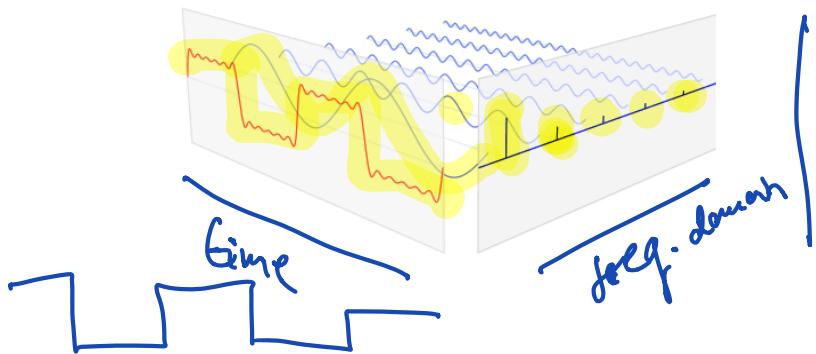


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(6.1) Fourier transform – frequency content



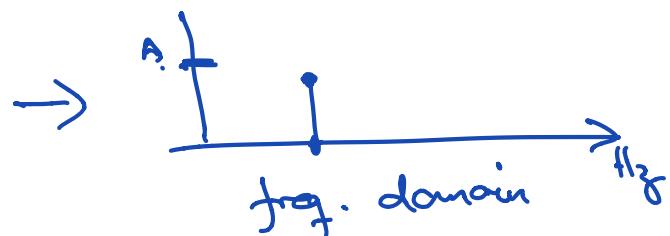
A signal can be regarded as a weighted sum of different frequency components.



Time (space) vs. Freq. domain – main idea



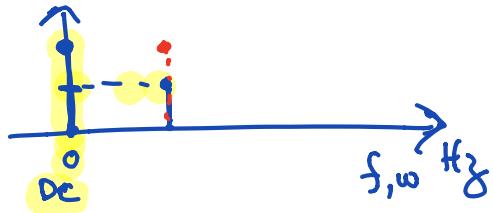
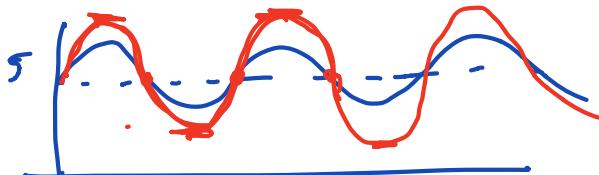
time domain



freq. domain



mean = 0 , DC - level is 0



Fourier transform

- Fourier series: any periodic function can be represented as a weighted sum of sines and cosines.
- Can be extended to aperiodic signals -> Fourier transform
- Discrete signals -> discrete time Fourier transform
- Can we extend to 2D (images)? -> Yes!
- We look at complex sinusoids:

$$e^{jw}$$

\downarrow

$$e^{jw} = \cos w + j \cdot \sin w$$

Eulers formula

\uparrow

Complex

Fourier transform (continuous)

- **Fourier transform $G(f)$:** the integration of the signal after first multiplying by a certain complex exponential:



$$G(f) \equiv \mathcal{F}\{g\} \equiv \int_{-\infty}^{\infty} g(t) e^{-j2\pi ft} dt$$



$$g(t) = \mathcal{F}^{-1}\{G\} \equiv \int_{-\infty}^{\infty} G(f) e^{j2\pi ft} df$$

- If t is measured in seconds, then f is measured in inverse seconds, also known as hertz.
- By applying **Euler's formula:**

$$G(f) = \underbrace{\int_{-\infty}^{\infty} g(t) \cos 2\pi ft dt}_{G_{even}} + j \underbrace{\int_{-\infty}^{\infty} -g(t) \sin 2\pi ft dt}_{G_{odd}}$$

Discrete time Fourier transform (1D)

$$X(e^{jw}) = \mathcal{F}(x(n)) = \sum_{n=-\infty}^{\infty} x(n)e^{-jwn}$$

$\xrightarrow{\hspace{10em}}$

$$\rightarrow x(n) = \left[\frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{jw}) e^{jwn} dw \right]$$

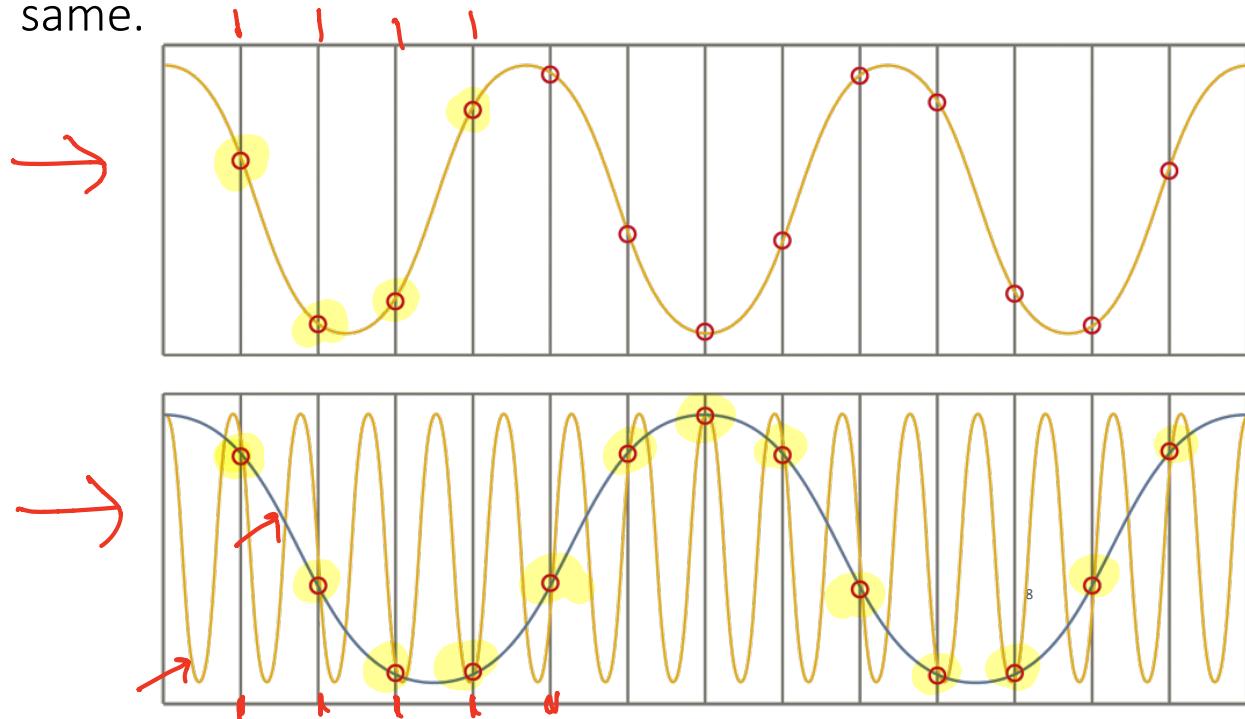
$\xrightarrow{\hspace{10em}}$

Eulers formula:
 $e^{jx} = \cos x + j \sin x$

Dealing with a discrete time signal (or image) we can only represent frequencies between $[-Fs/2, Fs/2]$. Fs is the samplingfrequency.

The frequency is normalized to $[-\pi, \pi]$ rad. Outside this frequency area, the specter repeats itself, i.e. the Discrete time fourier transform is periodic. WHY? - aliasing

Aliasing – for discrete signals, multiple frequencies look the same.

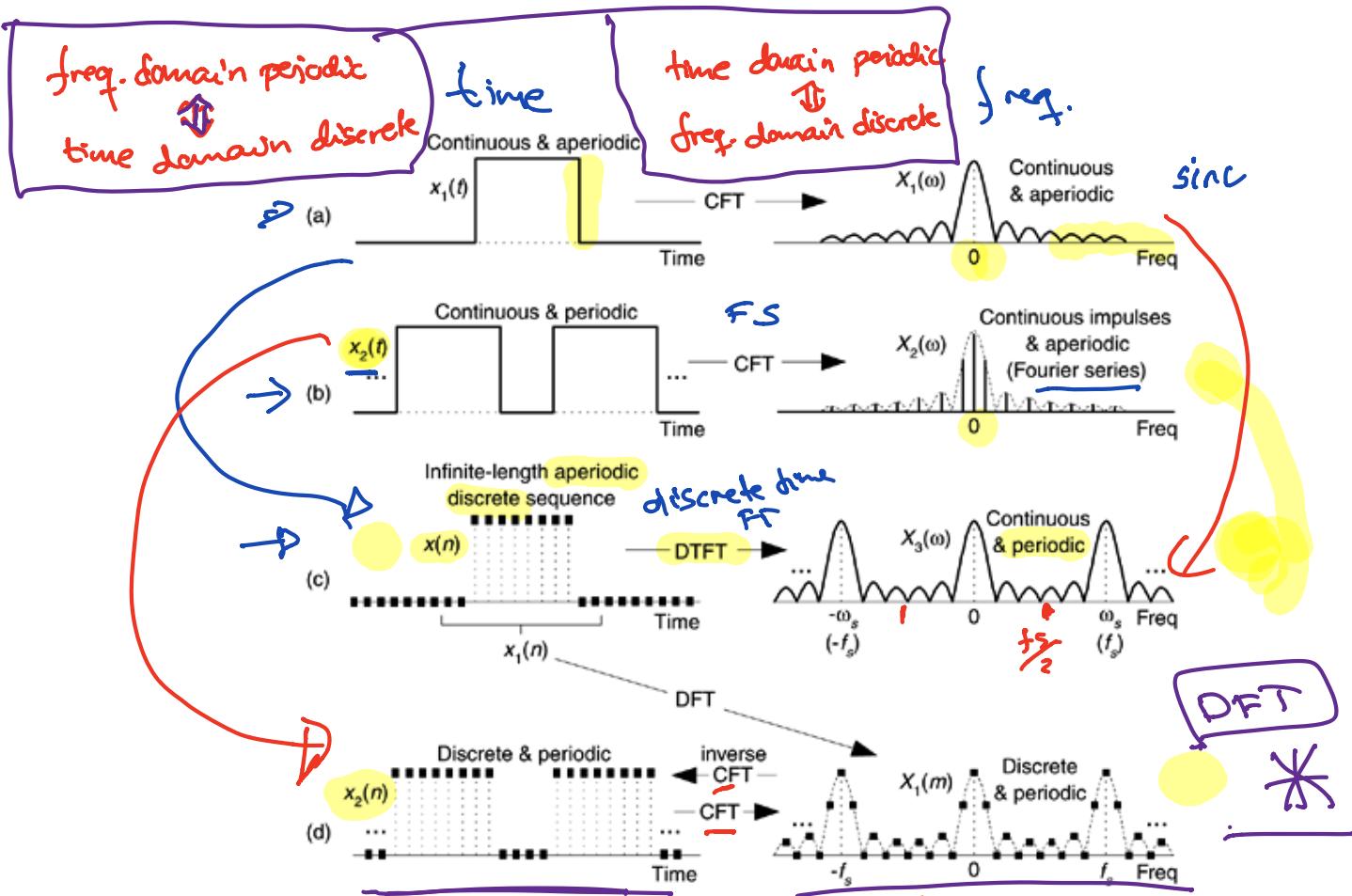


Nyquist rate

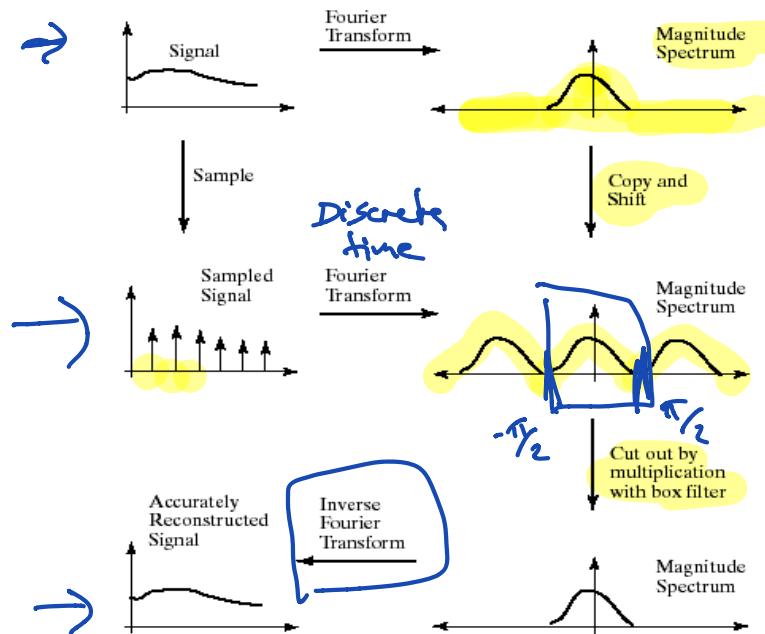
- **Nyquist-Shannon sampling theorem:** if a certain condition holds true, then the discrete samples contain just as much information as the original signal, so that the original signal can be reconstructed *exactly* from the discrete samples.
- The sampling rate must be greater than the **Nyquist rate**, which is twice the highest frequency in the signal.
- **Oversampled:** when the sampling frequency is greater than the Nyquist rate.
 - Perfect reconstruction is possible.
- **Undersampled:** when the sampling frequency is lower than the Nyquist rate.
 - Important information about the signal is irrecoverably lost.
 - When a signal is undersampled, **aliasing** occurs.
- **Critically sampled:** when the sampling frequency is exactly the Nyquist rate.
 - The original signal is also unrecoverable.

Fourier series and transform

		continuous signals (images)	discrete signal
aperiodic signals	Fourier transform (FT) aperiodic	discrete time Fourier transform (DTFT) periodic	
periodic signals	Fourier Series (FS) aperiodic, discrete output.	Discrete Fourier transform (DFT, FFT) periodic, discrete	



Sampling without aliasing



Sampling with aliasing

