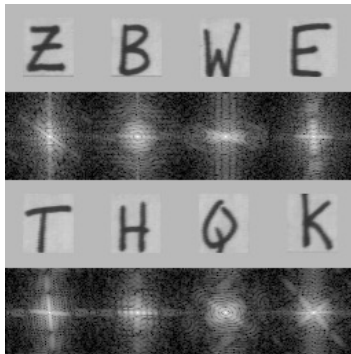


COMS20011 – Data-Driven Computer Science

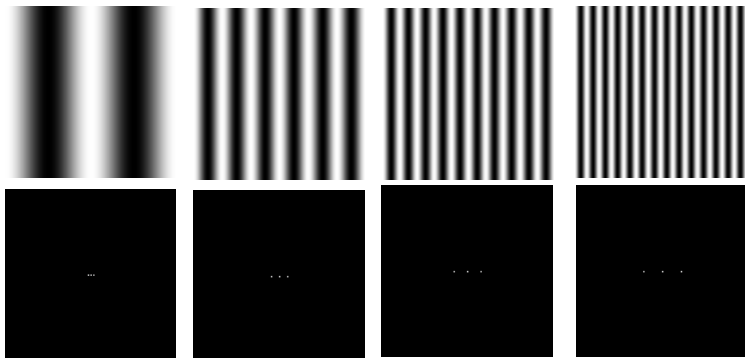


2D Fourier Transform

March 2023

Majid Mirmehdi

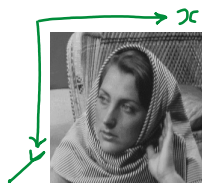
Next in DDCS



Feature Selection and Extraction

- Signal basics and Fourier Series
- 1D and **2D Fourier Transform**
- Another look at features
- Convolutions

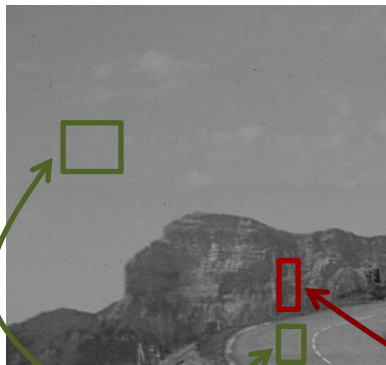
The 2D Fourier Transform



FT \rightarrow straightforward extension to 2D:

- Images are functions of two variables \rightarrow e.g. $f(x, y)$
- Defined in terms of *spatial frequency* \rightarrow 2D frequency.
- Fourier Transform is particularly useful for characterising intensity variations across an image.
- FT identifies the *Rate of change of intensity* along each dimension.

Examples: Spatial Frequency



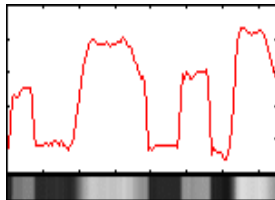
Slowly changing \rightarrow low frequency

Rapidly changing \rightarrow high frequency

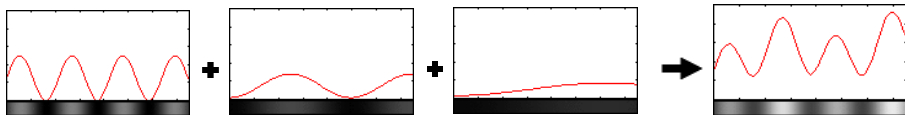
Images are waves!?

(or intuition behind FT)

Consider a single row (or column) of pixels from an image:



Add some regular waves to get one that is close (or as good as) the image



2D Fourier Transform: Continuous Form

The Fourier Transform of a continuous function of two variables $f(x, y)$ is:

$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-j2\pi(ux+vy)} dx dy$$

Conversely, given $F(u, v)$, we can obtain $f(x, y)$ by means of the *inverse* Fourier Transform:

$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u, v) e^{j2\pi(ux+vy)} du dv$$

These two equations are also known as the Fourier Transform Pair.

Note, they constitute a lossless representation of data.

2D Fourier Transform: Discrete Form

The FT of a discrete function of two variables, $f(x, y)$, $x, y = 0, 1, 2, \dots, N - 1$, is:

$$F(u, v) = \frac{1}{N^2} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y) e^{-j2\pi(\frac{ux+vy}{N})} \quad \text{for } u, v = 0, 1, 2, \dots, N - 1.$$

Conversely, given $F(u, v)$, we can obtain $f(x, y)$ by means of the *inverse FT*:

$$f(x, y) = \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} F(u, v) e^{j2\pi(\frac{ux+vy}{N})} \quad \text{for } x, y = 0, 1, 2, \dots, N - 1.$$

These two equations are also known as the Fourier Transform Pair.

Note, they constitute a lossless representation of data.

2D Fourier Transform

- The concept of the frequency domain follows from *Euler's Formula*:

$$e^{-j\theta} = \cos \theta - j \sin \theta$$

- Thus each term of the Fourier Transform is composed of the sum of *all* values of the function $f(x,y)$ multiplied by sines and cosines of various frequencies:

$$F(u, v) = \frac{1}{N^2} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y) \left[\cos\left(\frac{2\pi(ux + vy)}{N}\right) - j \sin\left(\frac{2\pi(ux + vy)}{N}\right) \right]$$

for $u, v = 0, 1, 2, \dots, N - 1$.

We have transformed from a **time domain** to a **frequency domain** representation.

2D Fourier Transform

- The concept of the frequency domain follows from *Euler's Formula*:

$$e^{-j\theta} = \cos \theta - j \sin \theta$$

- Thus each term of the Fourier Transform is composed of the sum of *all* values of the function $f(x,y)$ multiplied by sines and cosines of various frequencies:

$$F(u, v) = \frac{1}{N^2} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x, y)$$

for $u, v = 0, 1, 2, \dots, N - 1$.

The slowest varying frequency component,
i.e. when $u=0, v=0 \rightarrow$ average image
graylevel

We have transformed from a **time domain** to a **frequency domain** representation.

2D Fourier Transform

$F(u, v)$ is a complex number & has real and imaginary parts:

$$F(u, v) = R(u, v) + jI(u, v)$$

Magnitude or spectrum of the FT:

$$|F(u, v)| = \sqrt{R^2(u, v) + I^2(u, v)}$$

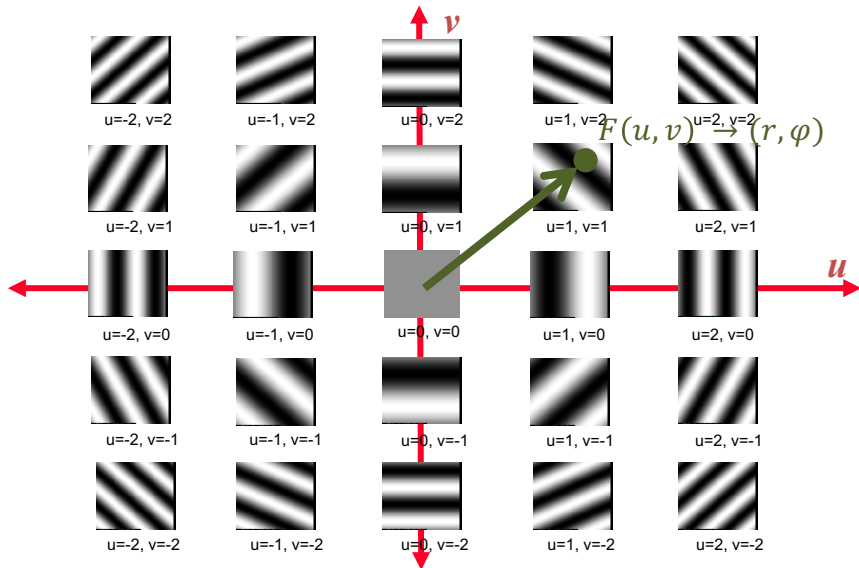
Phase angle or phase spectrum:

$$\varphi(u, v) = \tan^{-1} \frac{I(u, v)}{R(u, v)}$$

Expressing $F(u, v)$ in polar coordinates:

$$F(u, v) = |F(u, v)|e^{j\varphi(u, v)}$$

Another view: The 2D Basis Functions

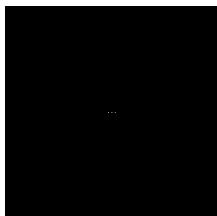
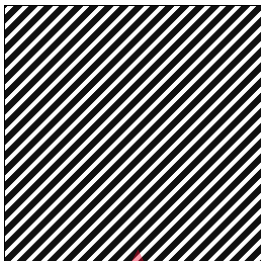


Example I: Image Analysis

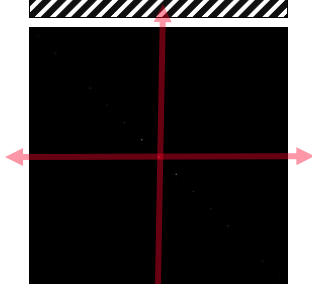
Image



Image

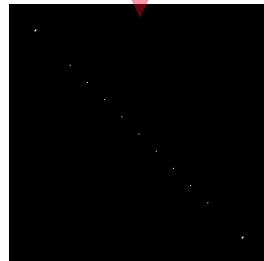
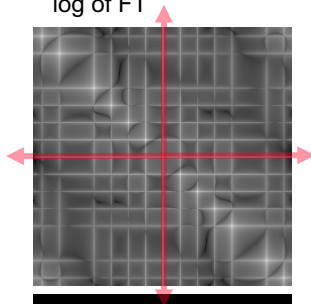


FT



FT

log of FT

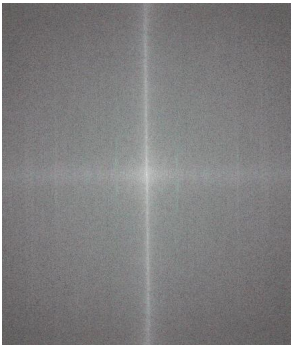


Thresholded log of FT

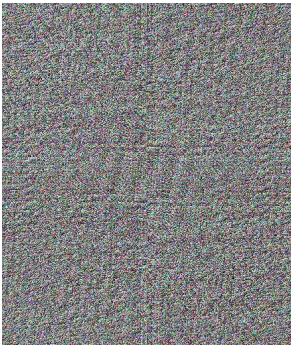
Example II: Magnitude + Phase



I

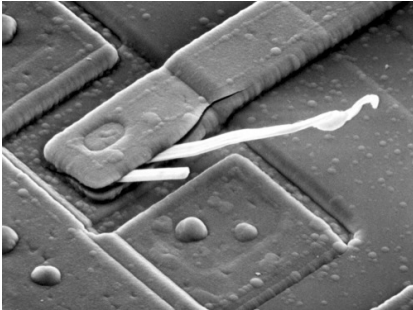


$\log(|F(I)| + 1)$



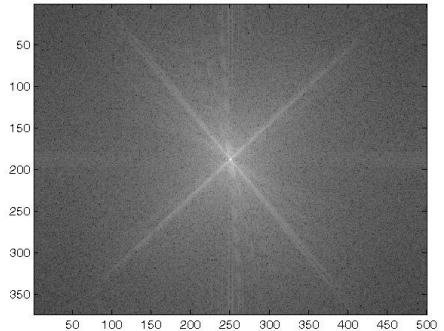
$\phi[F(I)]$

Example IV: Interpreting the FS

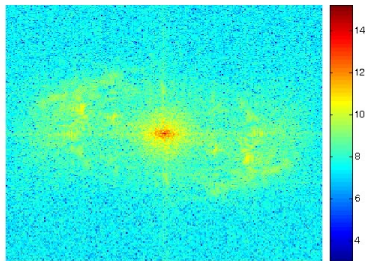


Scanning electron microscope
image of an integrated circuit

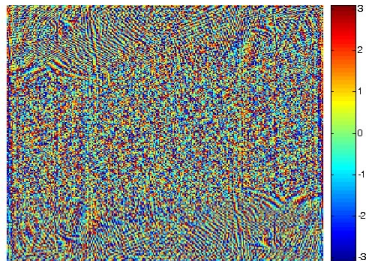
Can we interpret what the bright
components mean?



Viewing Magnitude and Phase

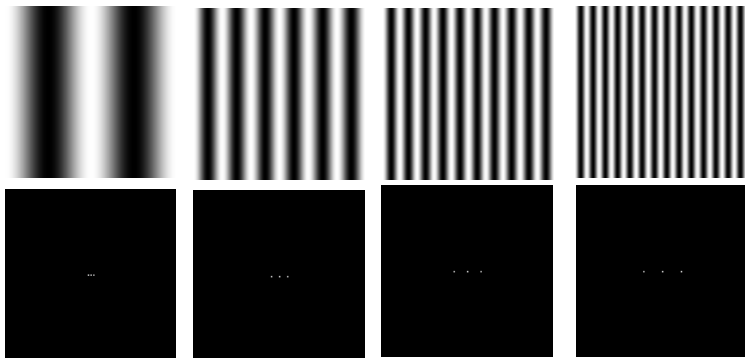


$$\log(|F(I)| + 1)$$



$$\phi[F(I)]$$

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