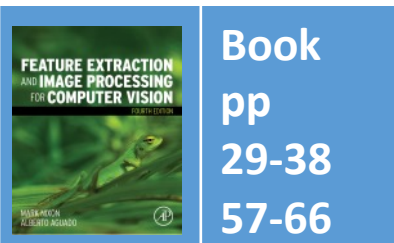


# Lecture 2 Image Formation

COMP3204 Computer Vision

**What is inside an image?**



Department of  
Electronics and  
Computer Science

UNIVERSITY OF  
**Southampton**  
School of Electronics  
and Computer Science

# Content

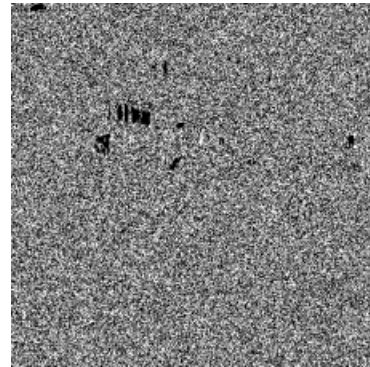
1. How is an image formed?
2. What restrictions are there on image formation?
3. Go to a different space - Fourier .....

# Decomposing an image into its bits

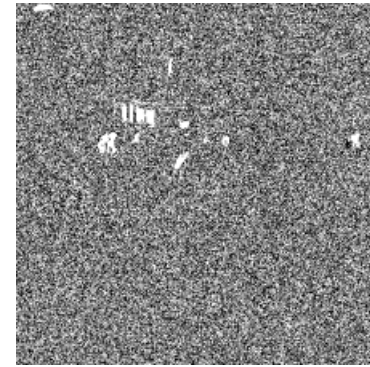
The **Most Significant Bit** carries the **most information** where as bit 0 is **noise**



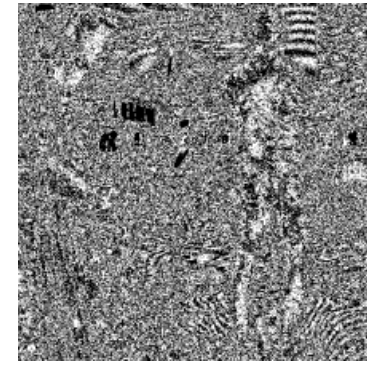
(a) original image



(b) bit 0 (LSB)



(c) bit 1



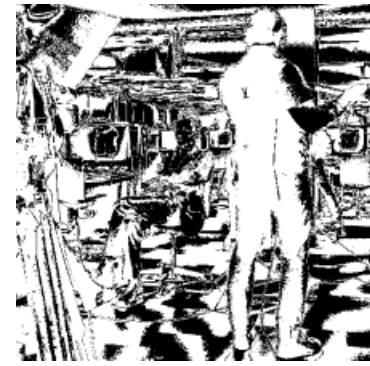
(d) bit 2



(e) bit 3



(f) bit 4



(g) bit 5

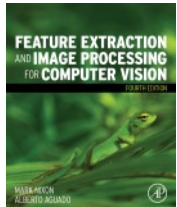


(h) bit 6



(i) bit 7 (MSB)

... and here, bit 4 is the **lighting**



# Effects of differing image resolution



(a)  $64 \times 64$



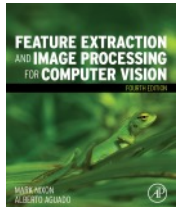
(b)  $128 \times 128$



(c)  $256 \times 256$

Low resolution lose information but  $N \times N$  points implies much storage

How do we choose an appropriate value for  $N$ ?



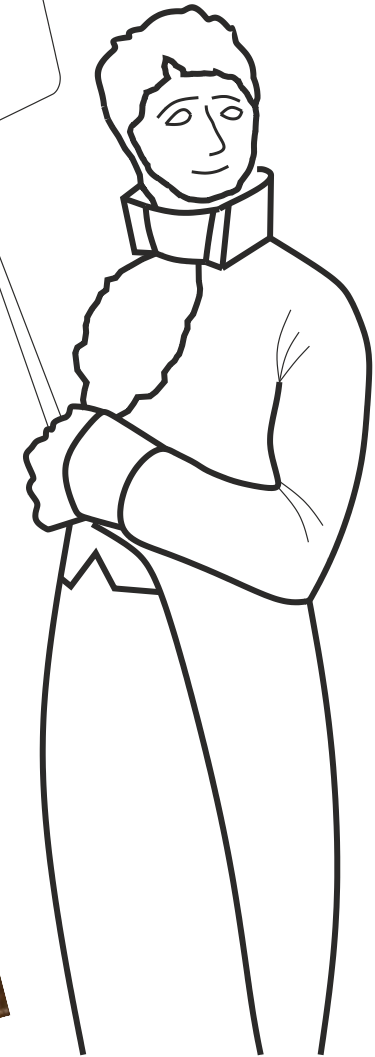
# Joseph Fourier

- Any periodic function is the result of adding up sine and cosine waves of different frequencies
- Sceptical? Yeah, so were Lagrange and Laplace. Good company eh?
- “Fourier’s treatise is one of the very few scientific books that can never be rendered antiquated by the progress of science”  
James Clerk Maxwell 1878



$$Fp(\omega) = \int_{-\infty}^{+\infty} p(t)e^{-j\omega t} dt$$

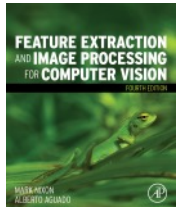
Yeah!





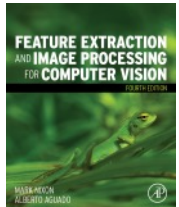
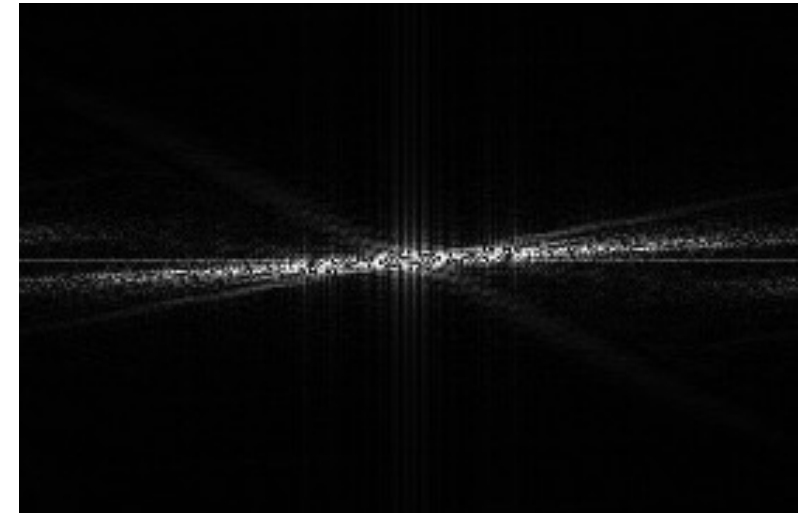
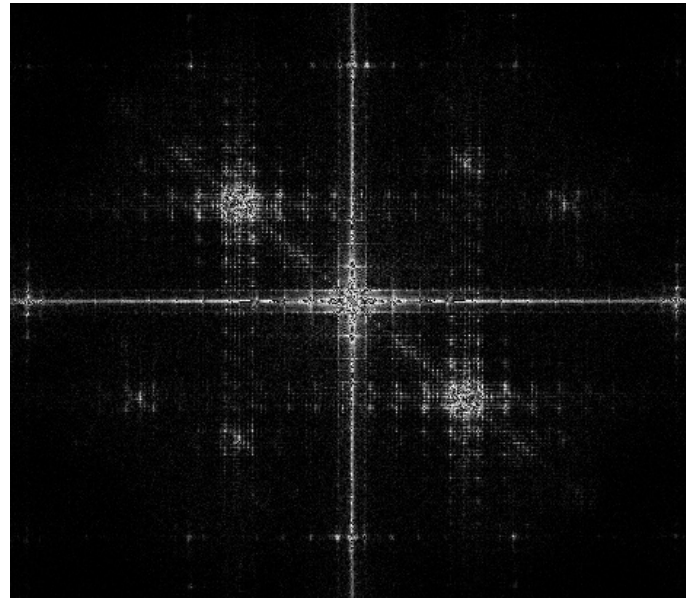
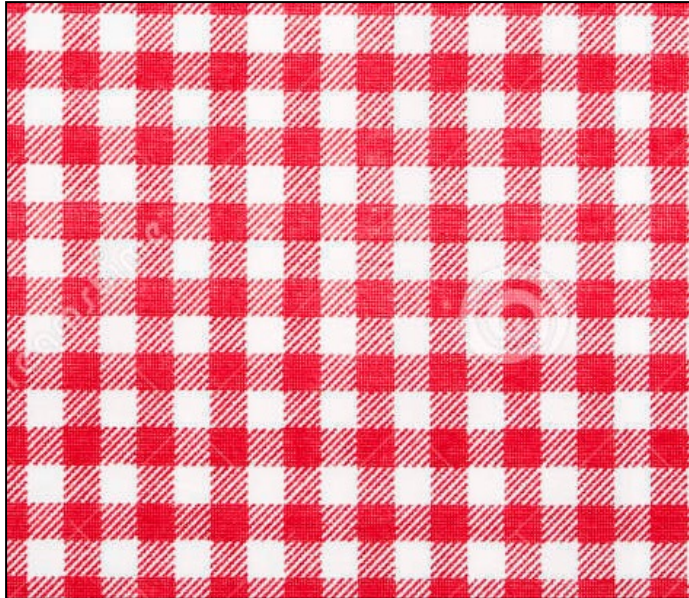
# What are 2D waves?

2D waves are along x and y axes simultaneously



and in terms of frequency

- N.b. colour immaterial (just for visuals)



# Step up Fourier...

$$Fp(f) = \mathcal{F}(p(t)) = \int_{-\infty}^{\infty} p(t)e^{-jft} dt$$

Whoa! Where from....

First, we have that the FT is a function  $a(\ )$  of a time-variant signal  $p(t)$

The Fourier transform is then

$$Fp = a(p(t))$$

The transform is a function of frequency so

$$Fp(f) = a(p(t))$$

$\mathcal{F}$  stands for the Fourier transform so

$$Fp(f) = \mathcal{F}(p(t)) = a(p(t))$$

The function  $a(\ )$  is actually an integral

$$Fp(f) = \mathcal{F}(p(t)) = \int_{-\infty}^{\infty} p(t)cas(t)dt$$

$cas(t)$  describes cosine and sine waves, so

$$Fp(f) = \mathcal{F}(p(t)) = \int_{-\infty}^{\infty} p(t)e^{-jft} dt$$

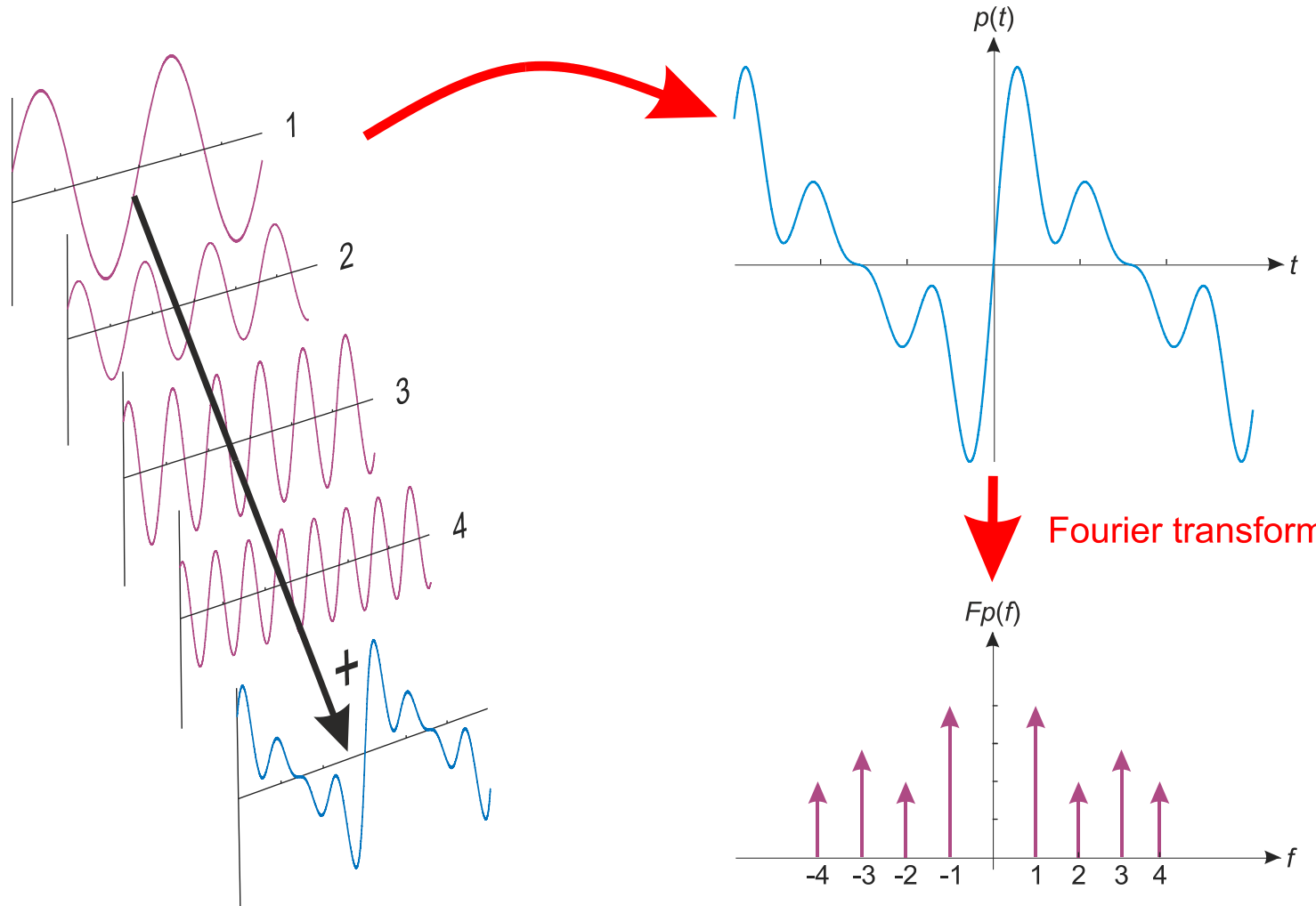
where  $cas = \text{'cos and sin'}$  since  $e^{-jft} = \cos(ft) - j \sin(ft)$ ,

where  $j$  is the complex number  $j = \sqrt{-1}$





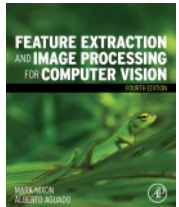
# What does the Fourier transform do?



Addition in the time domain

Fourier transform

Separation in the frequency domain

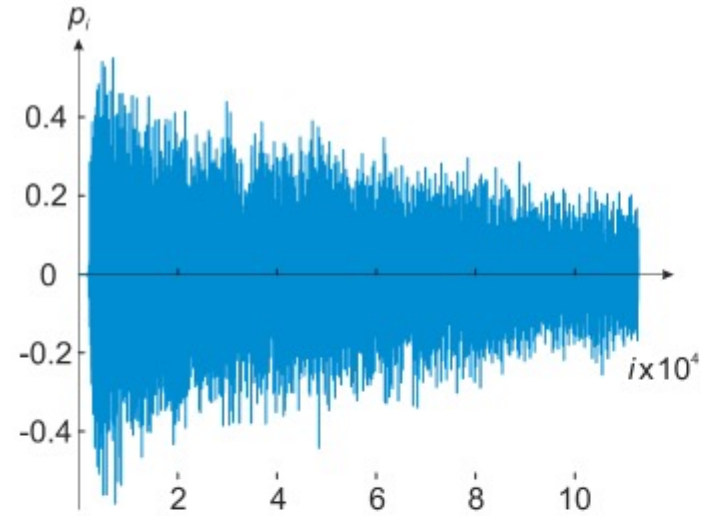


# Hard day?

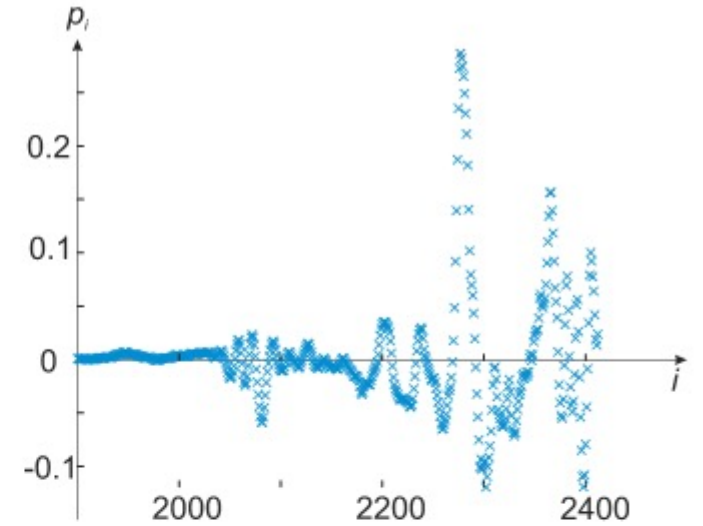
Let's see the Fourier transform of the Hard Day's night chord



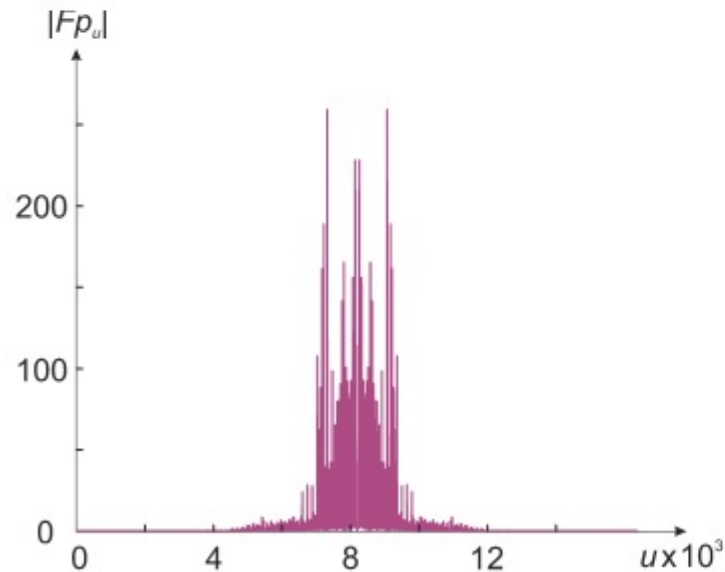
Apparently, the **piano** is more dominant than expected



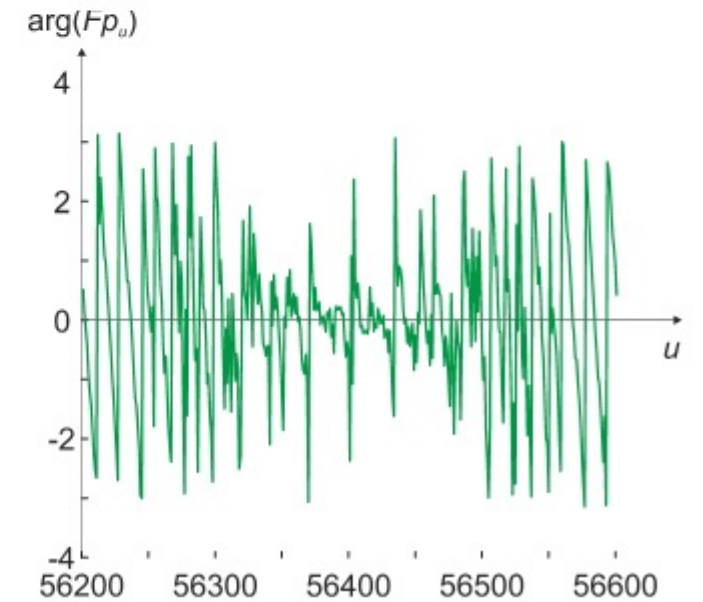
(a) Recorded data



(b) A closer look at the start of (a)

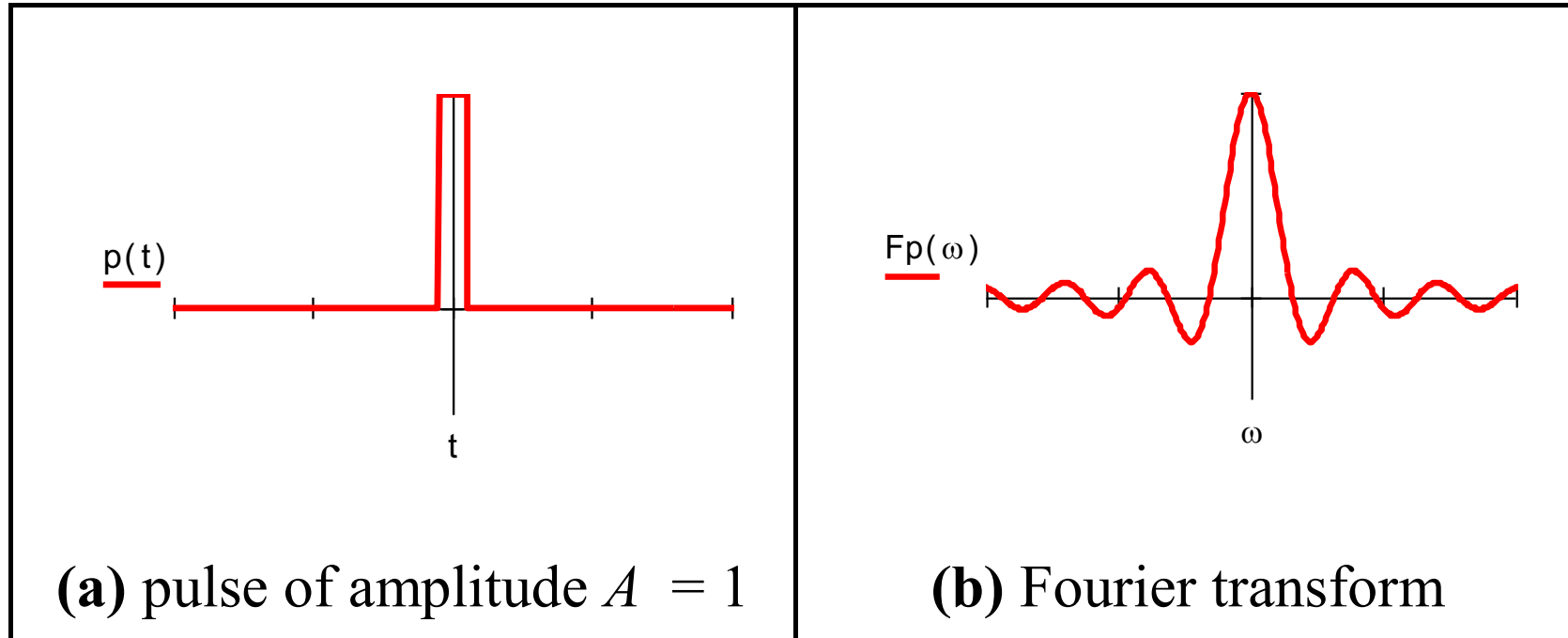


(c) Fourier Spectrum



(d) Phase of central portion

# A rectangular pulse and its Fourier transform

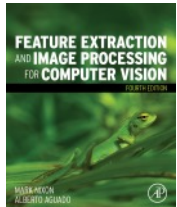


- Pulse  $p(t) = \begin{cases} A & \text{if } -T/2 \leq t \leq T/2 \\ 0 & \text{otherwise} \end{cases}$

- Use Fourier  $Fp(\omega) = \int_{-T/2}^{T/2} Ae^{-j\omega t} dt$

- Evaluate integral  $Fp(\omega) = -\frac{Ae^{-j\omega T/2} - Ae^{j\omega T/2}}{j\omega}$

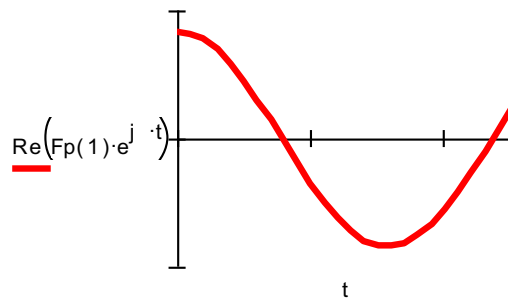
- And get result  $Fp(\omega) = \begin{cases} \frac{2A}{\omega} \sin\left(\frac{\omega T}{2}\right) & \text{if } \omega \neq 0 \\ AT & \text{if } \omega = 0 \end{cases}$



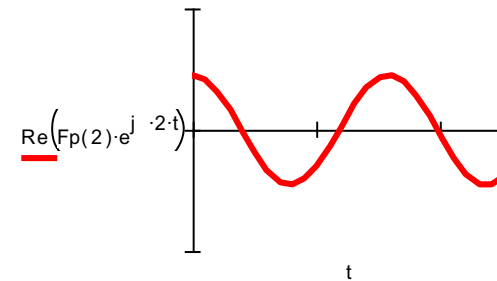


# Reconstructing a signal from its Fourier transform

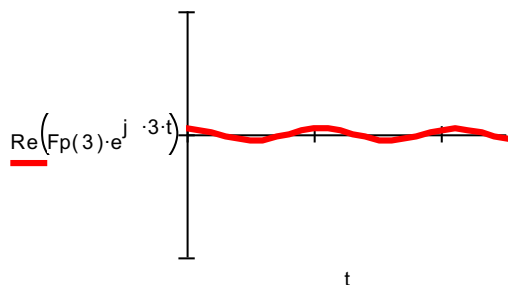
This is the **inverse Fourier transform**



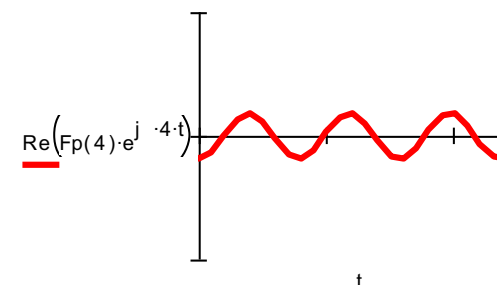
(a) contribution for  $\omega = 1$



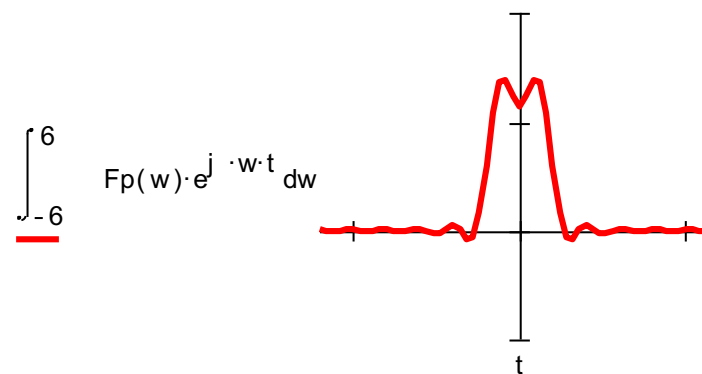
(b) contribution for  $\omega = 2$



(c) contribution for  $\omega = 3$

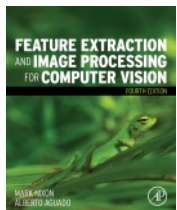


(d) contribution for  $\omega = 4$

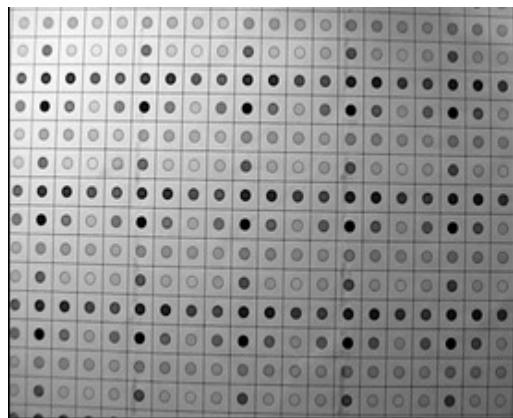
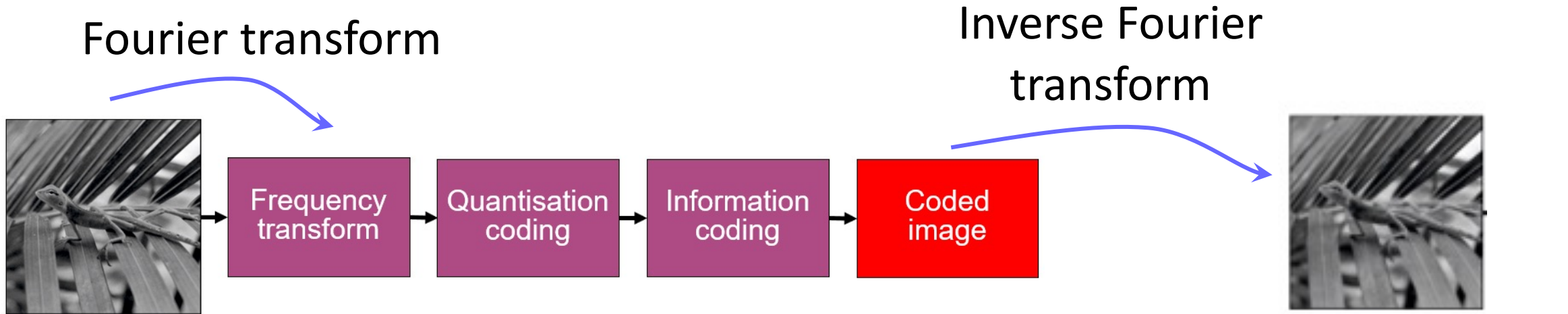


(e) reconstruction by integration

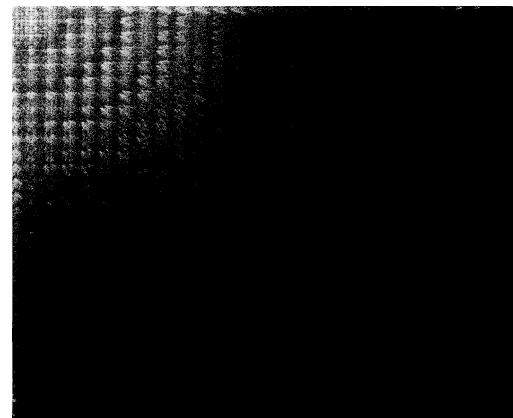
**Reconstructing a Signal from its Transform**



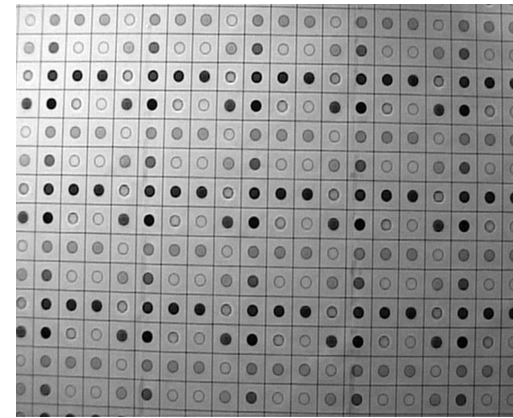
# Inverse Fourier transform is used for reconstruction



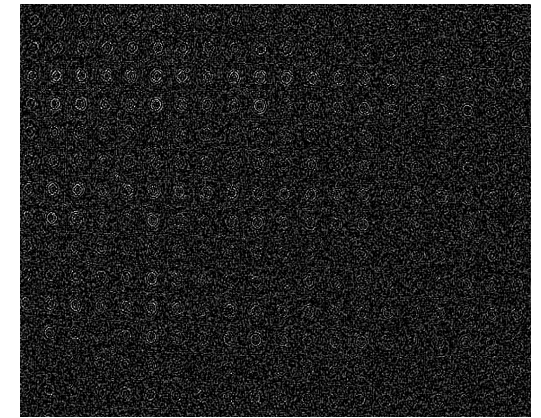
image



5% of transform components



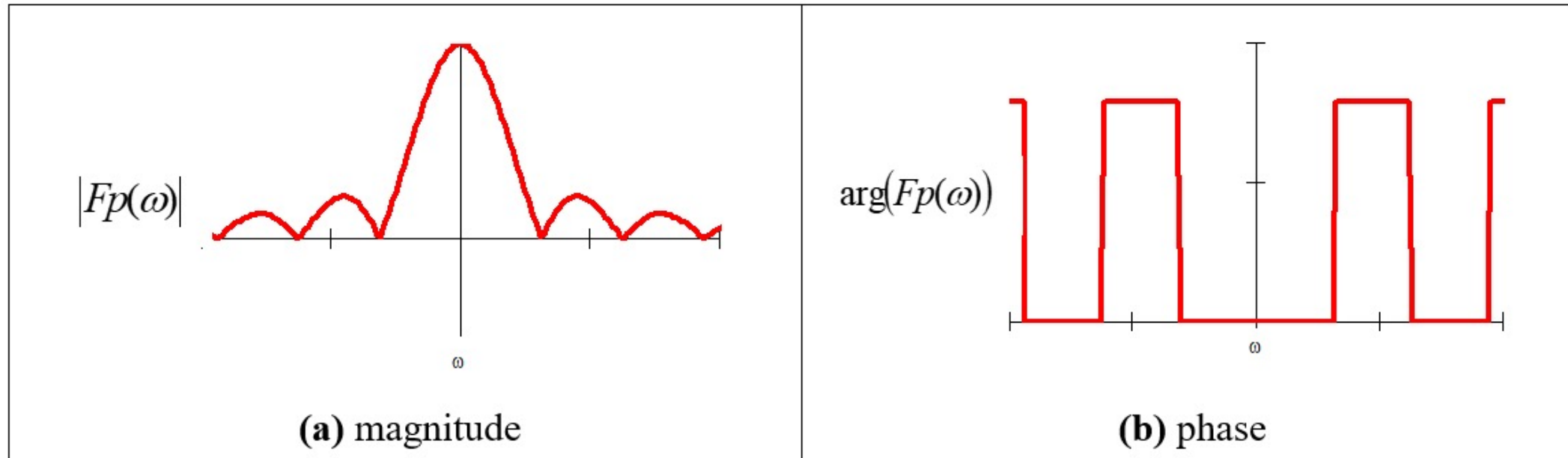
reconstructed image



error

# Magnitude and phase of Fourier transform of a pulse

$$Fp(\omega) = \int_{-\infty}^{\infty} p(t)e^{-j\omega t} dt = \text{Re}(Fp(\omega)) + j \text{Im}(Fp(\omega))$$



$$|Fp(\omega)| = \sqrt{\text{Re}(Fp(\omega))^2 + \text{Im}(Fp(\omega))^2}$$

$$\arg(Fp(\omega)) = \tan^{-1}\left(\frac{\text{Im}(Fp(\omega))}{\text{Re}(Fp(\omega))}\right)$$



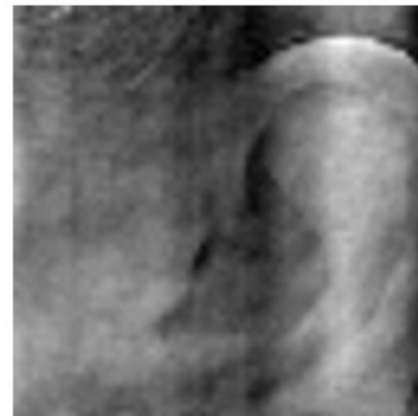
# Illustrating the importance of phase



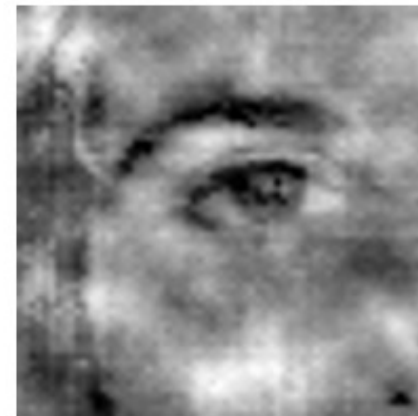
**(a)** eye image



**(b)** ear image



**(c)** reconstruction  
from magnitude(eye)  
and phase(ear)



**(d)** reconstruction from  
magnitude(ear) and  
phase(eye)



# Main points so far

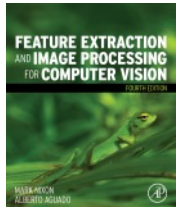
1 – **sampling** data is not as simple as it appears

2 – sampling affects **space** and **brightness**

3 – Fourier allows us to understand **frequency**

4 – Fourier allows for **coding** and more

Next, Fourier will allow us to understand  
sampling



# Other transforms

- Discrete Cosine (Sine) Transform
- Discrete Hartley Transform

## Wavelets

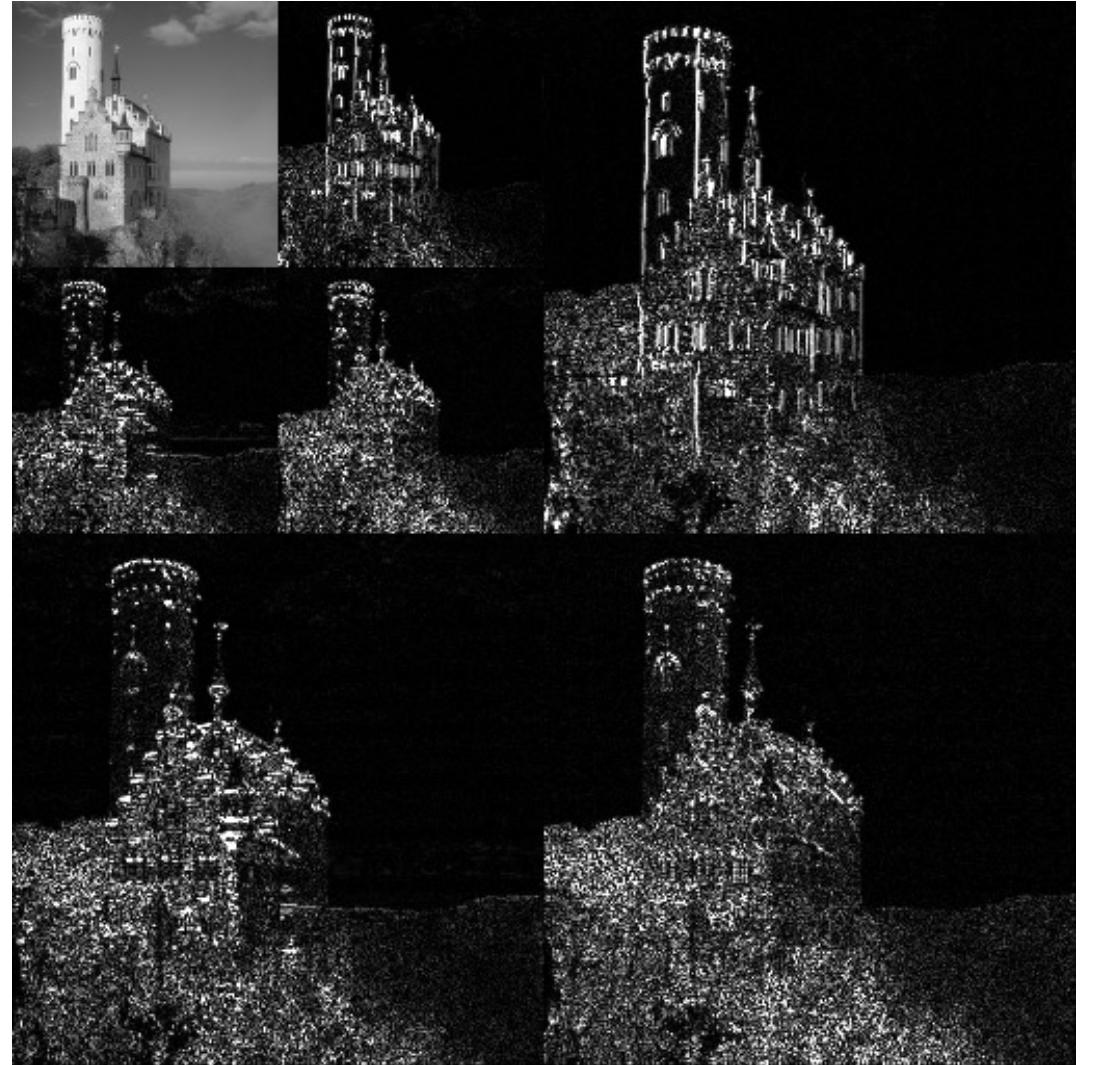
Continuous wavelet  
Discrete wavelet  
Complex wavelet  
Stationary wavelet  
Dual wavelet  
Haar wavelet  
Daubechies wavelet  
Morlet wavelet  
Gabor wavelet  
.....

## Curvelets

## Shearlets

Bandelet  
Contourlet  
Fresnelet  
Chirplet  
Noiselet  
.....

# Wavelet transform



An example of the 2D discrete wavelet transform that is used in JPEG2000 [Credit: Wikipedia]