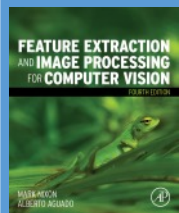


Lecture 3 Image Sampling

COMP3204 Computer Vision

How is an image sampled and what does it imply?



Book
pp
36-56

Department of
Electronics and
Computer Science

UNIVERSITY OF
Southampton
School of Electronics
and Computer Science

Content

1. How does the discrete Fourier transform work, and help?
2. What can go wrong with sampling?

1D Discrete Fourier transform

Discrete Fourier calculates frequency from data points

$$Fp_u = \frac{1}{N} \sum_{i=0}^{N-1} p_i e^{-j\frac{2\pi}{N}iu}$$

Comparison

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

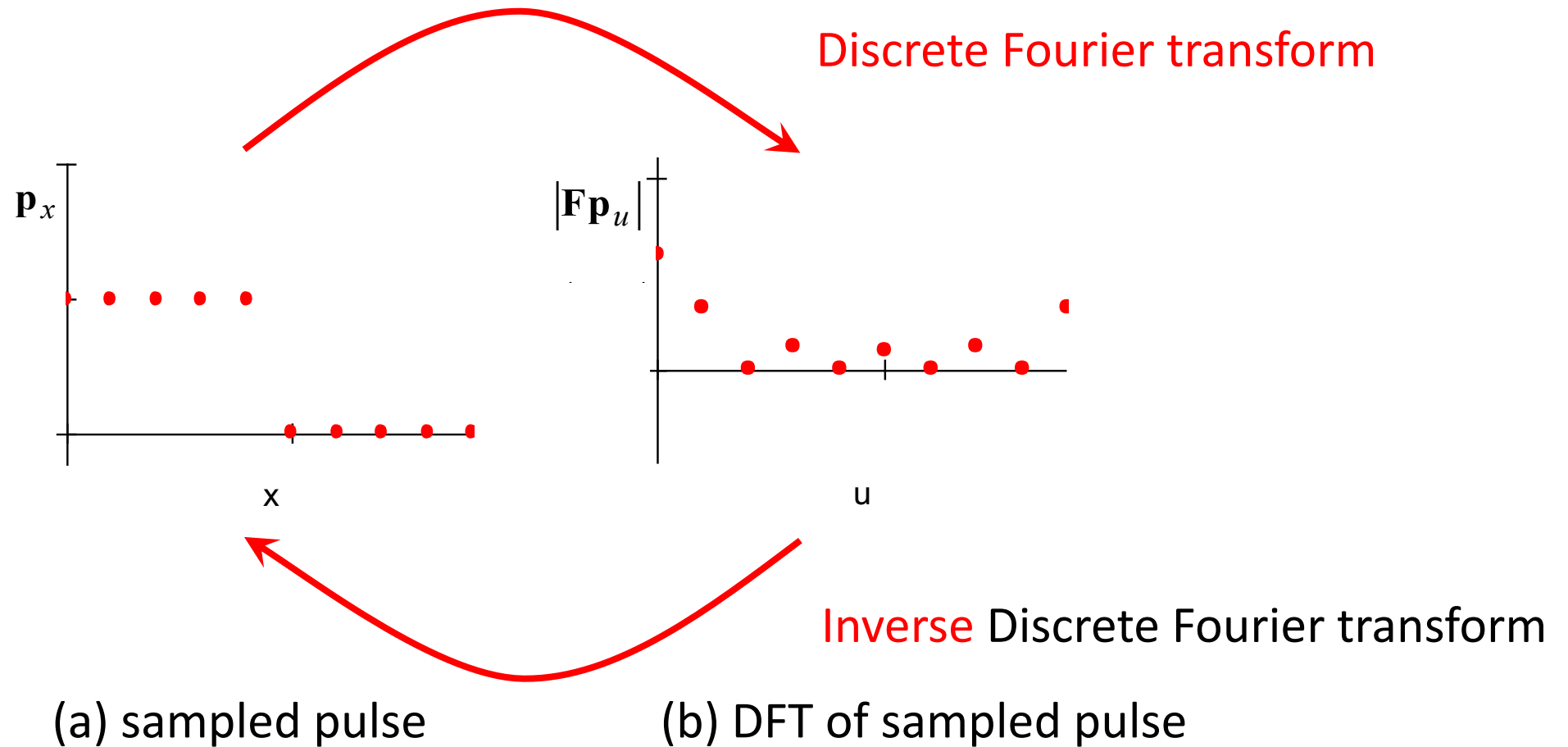
sampled frequency Fp_u

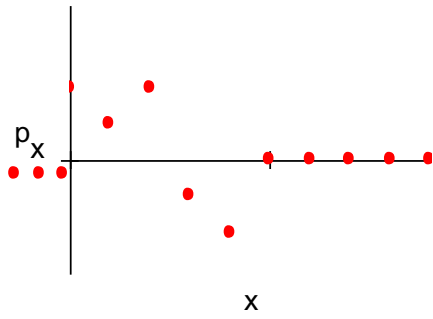
sampled points p_i

N points

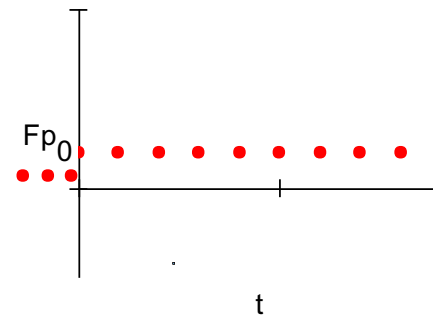
$$e^{-j\theta} = \cos \theta - j \sin \theta$$

Transform Pair for Sampled Pulse

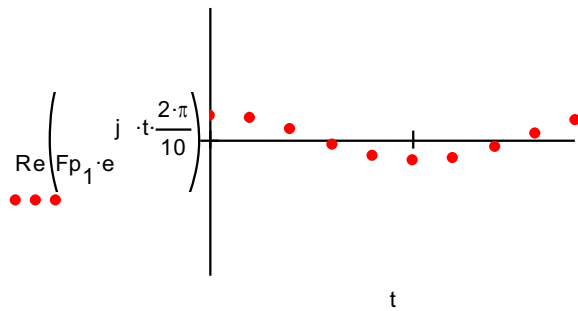




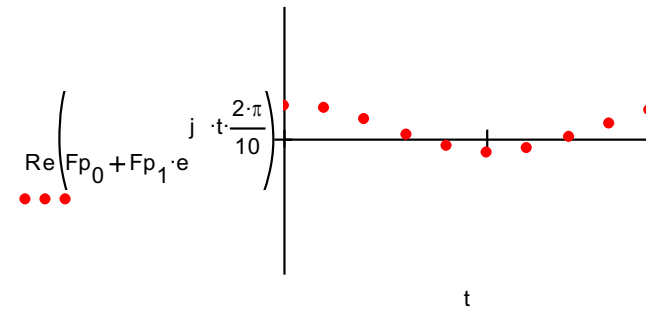
(a) original sampled signal



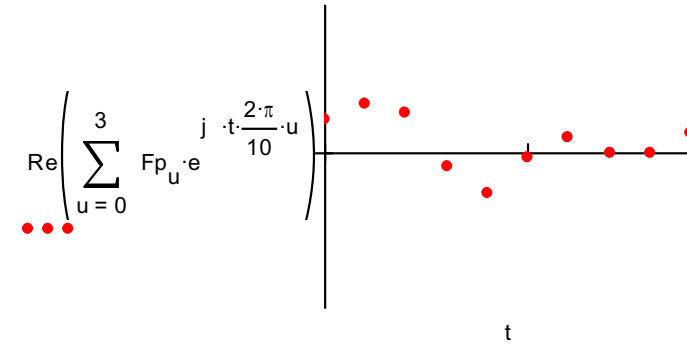
(b) first coefficient Fp_0



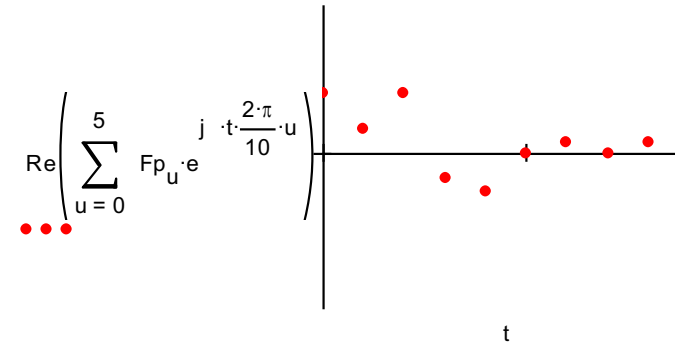
(c) second coefficient Fp_1



(d) adding Fp_1 and Fp_0



(e) adding Fp_0, Fp_1, Fp_2 and Fp_3



(f) adding all six frequency components



signal reconstruction from its transform components

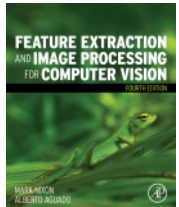
2D Fourier transform

Forward transform
$$\mathbf{FP}_{u,v} = \frac{1}{N^2} \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} \mathbf{P}_{x,y} e^{-j\left(\frac{2\pi}{N}\right)(ux+vy)}$$

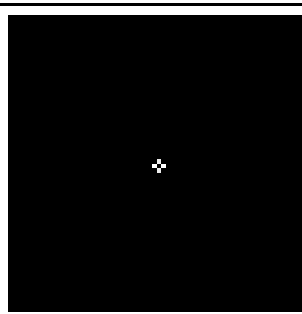
where two dimensions of space, x and y
two dimensions of frequency, u and v
image $N \times N$ pixels $\mathbf{P}_{x,y}$

Inverse transform
$$\mathbf{P}_{x,y} = \sum_{u=0}^{N-1} \sum_{v=0}^{N-1} \mathbf{FP}_{u,v} e^{j\left(\frac{2\pi}{N}\right)(ux+vy)}$$

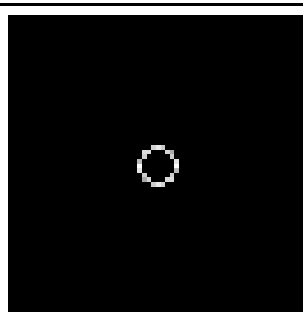
$\pi??$



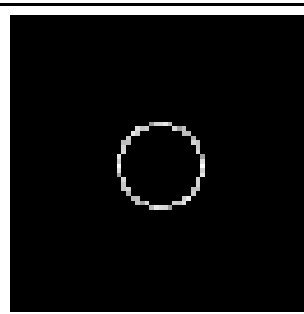
Reconstruction



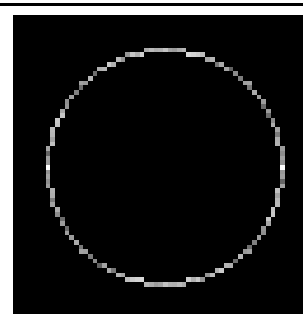
(a) transform
radius 1
components



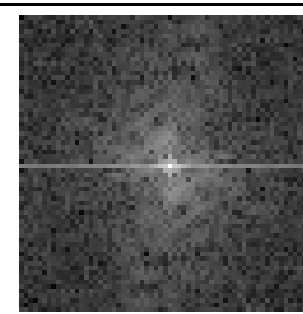
(b) transform
radius 4
components



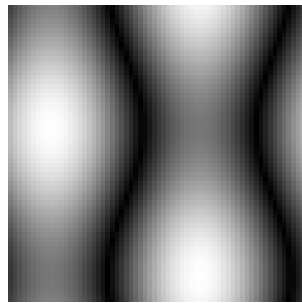
(c) transform
radius 9
components



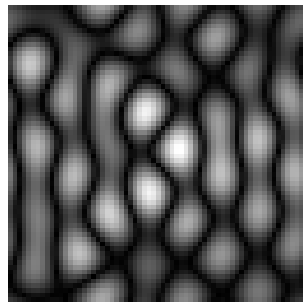
(d) transform
radius 25
components



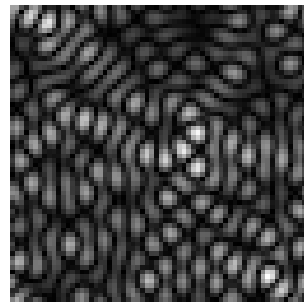
(e) complete
transform



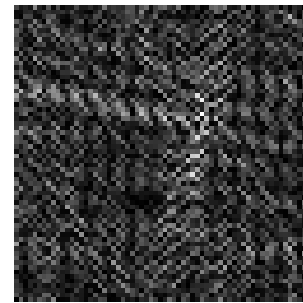
(f) image by radius
1 components



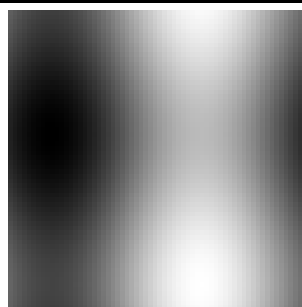
(g) image by
radius 4
components



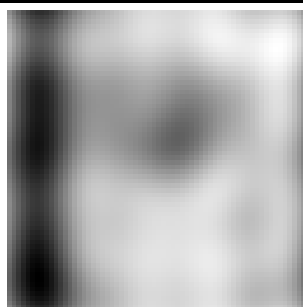
(h) image by
radius 9
components



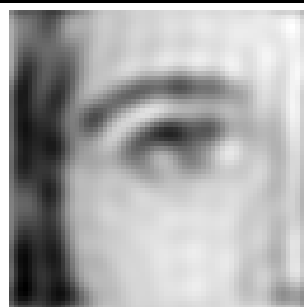
(i) image by radius
25 components



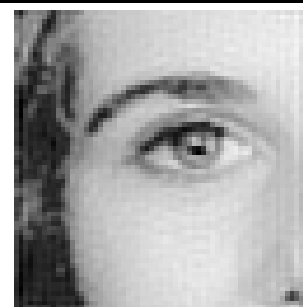
(j) reconstruction
up to 1st



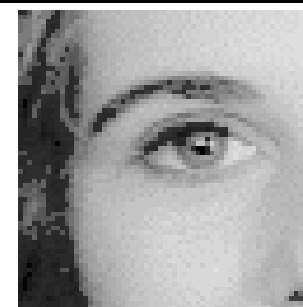
(k) reconstruction
up to 4th



(l) reconstruction
up to 9th



(m) reconstruction
up to 25th



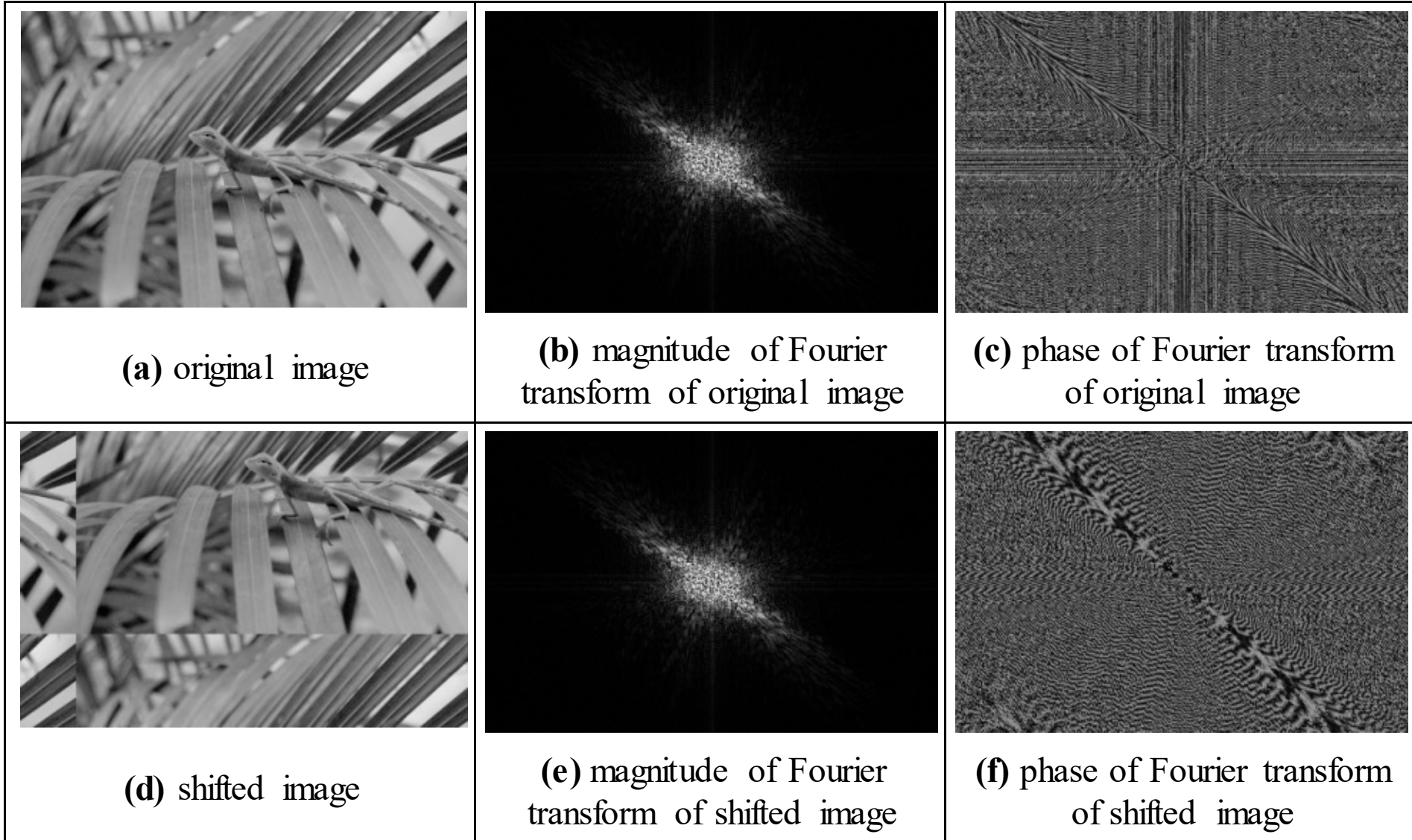
(n) reconstruction
with all

Implementation is via (Fast) FFT

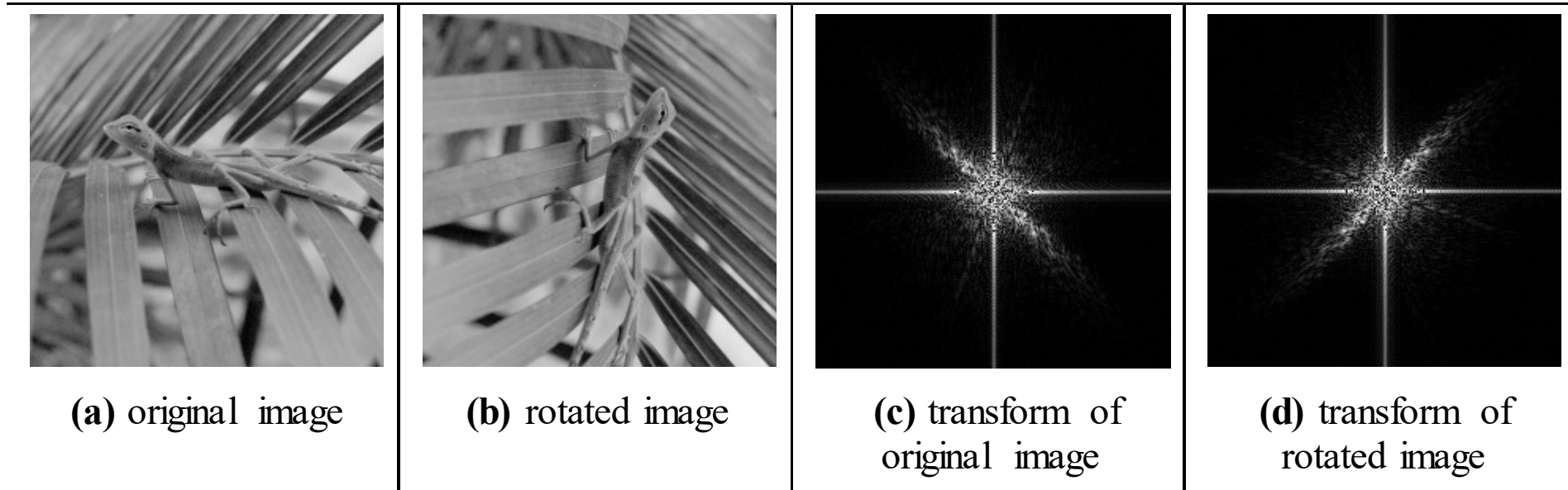
```
while L<cols %iterate until log2(cols)-1 levels have been performed
  for j=1:2*L:cols %do all the points in L/2 batches
    for i=1:L %now do L butterflies
      upp(((j+1)/2)+i-1)= Fp(j+i-1)+Fp(j+L+i-1)*exp(-1j*2*pi*(i-1)/(L*2));
      low(((j+1)/2)+i-1)= Fp(j+i-1)-Fp(j+L+i-1)*exp(-1j*2*pi*(i-1)/(L*2));
    end
  end
  for j=1:2*L:cols %copy the components across, to the right places
    for i=1:L
      Fp(j+i-1)=upp(((j+1)/2)+i-1);
      Fp(j+L+i-1)=low(((j+1)/2)+i-1);
    end
  end
  L=L*2; %and go and do the next level (up)
end
```

(this is a 1-D FFT)

Shift invariance



Rotation



$$\mathbf{FP}_{u,v} = \frac{1}{N} \sum_{y=0}^{N-1} \sum_{x=0}^{N-1} \mathbf{P}_{x,y} e^{-j\left(\frac{2\pi}{N}\right)(uy+vx)}$$



Filtering

Fourier gives access to
frequency components



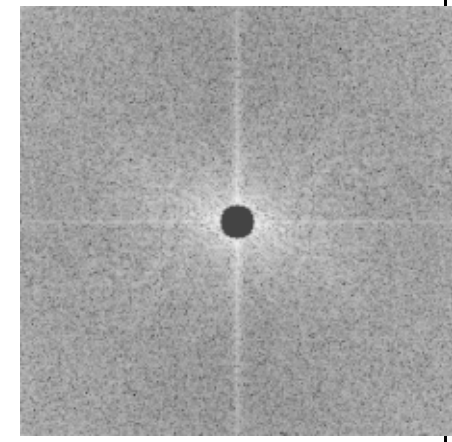
(a) low-pass filtered image



(b) low-pass filtered transform



(c) high-pass filtered image

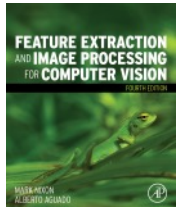


(d) high-pass filtered transform



Applications of 2D FT

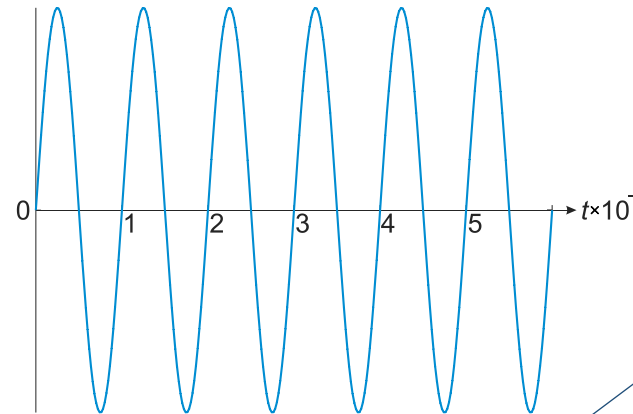
- Understanding and analysis
- Speeding up algorithms
- Representation (*invariance*)
- Coding
- Recognition/ understanding (e.g. texture)



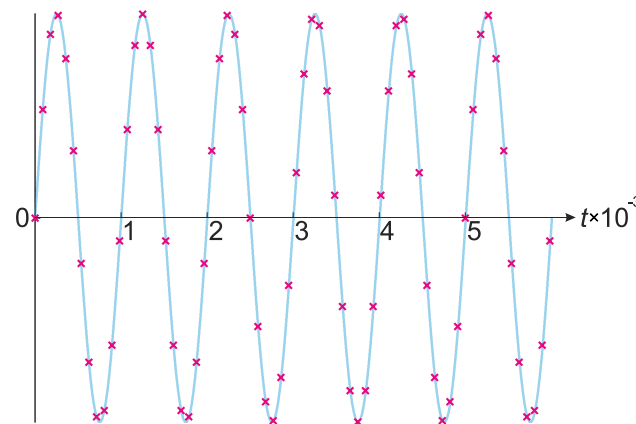
Sampling Signals



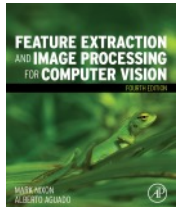
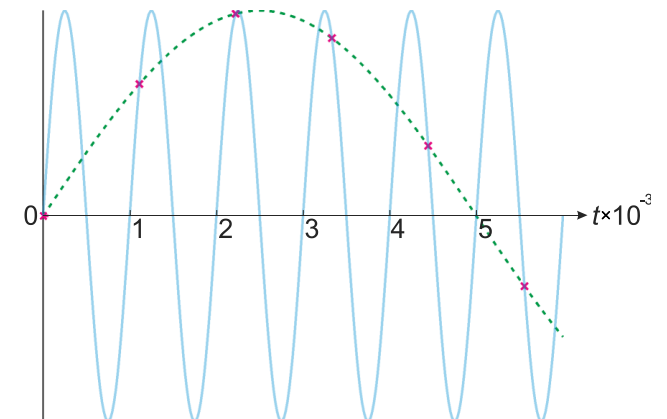
original continuous signal

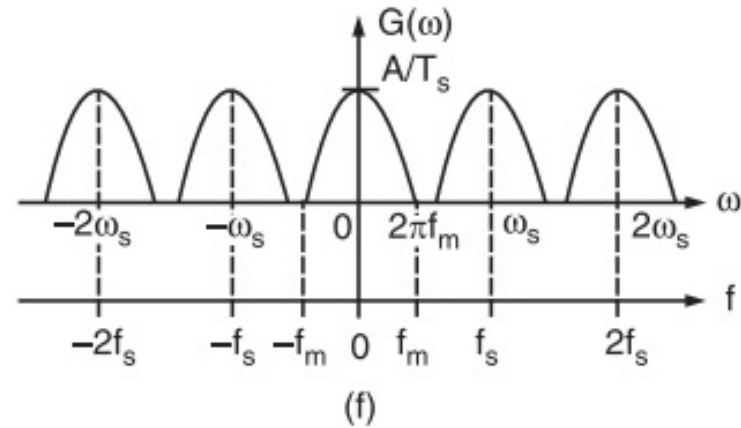
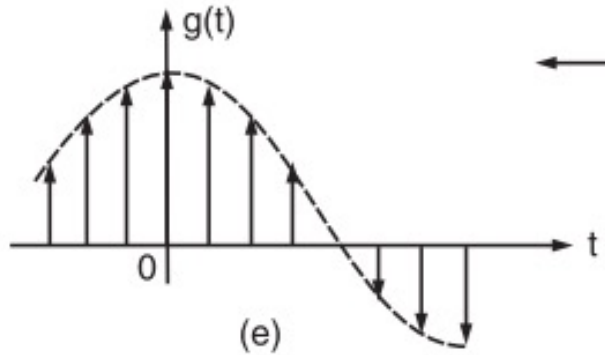
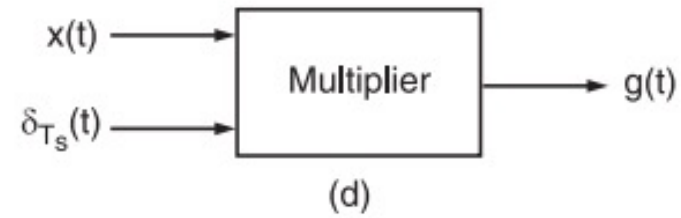
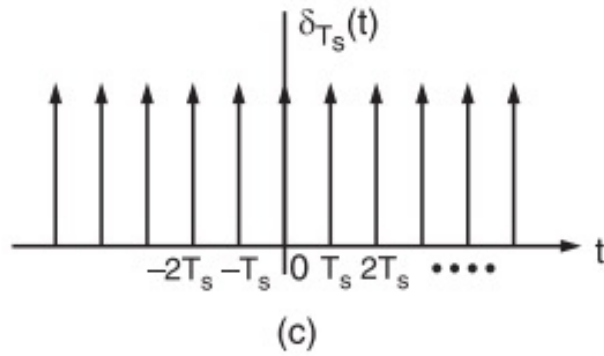
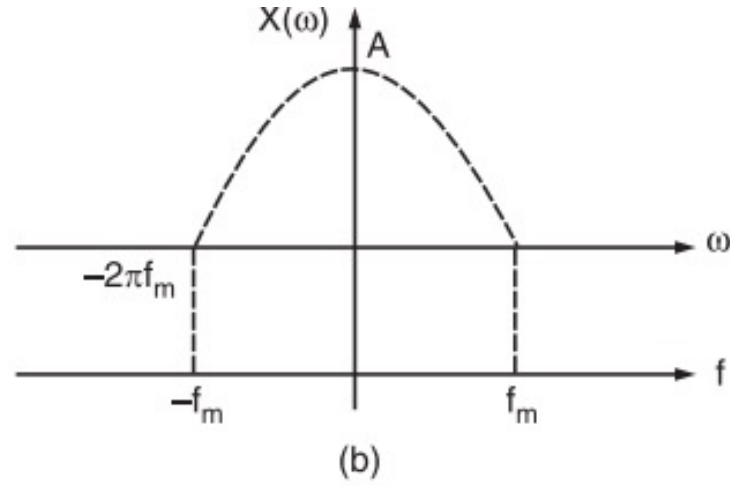
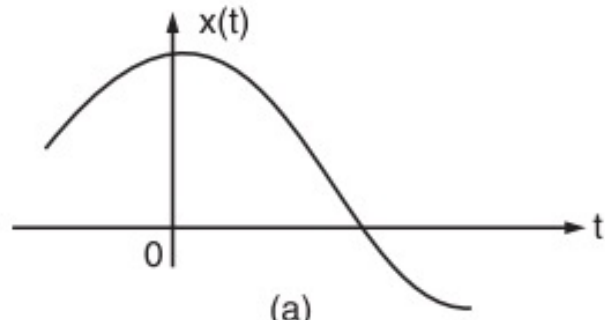


good
sampling

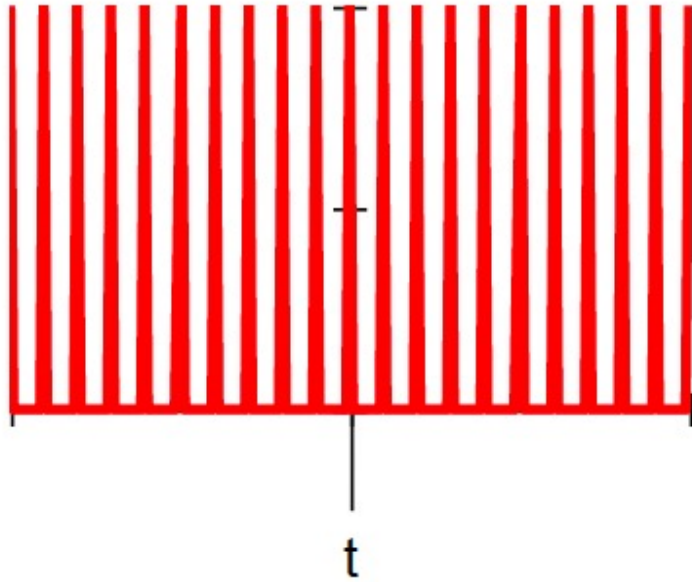


bad
sampling
(aliased)

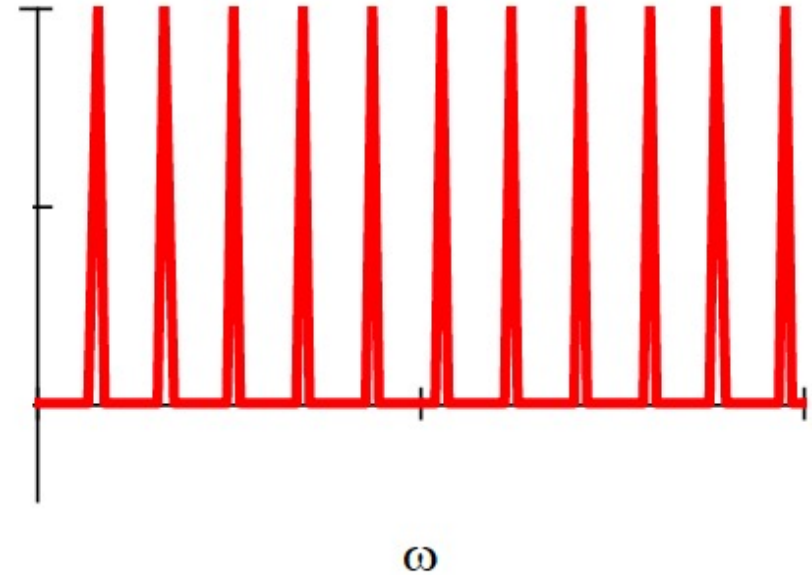




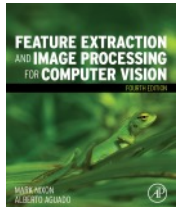
Sampling function



Sampling function in
time domain



Sampling function in
frequency domain - fft



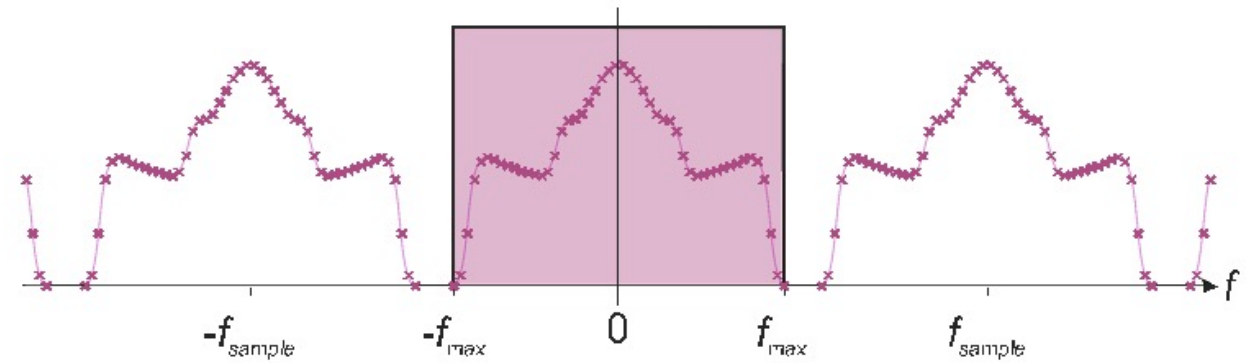
In the frequency domain

Spectra **repeat**

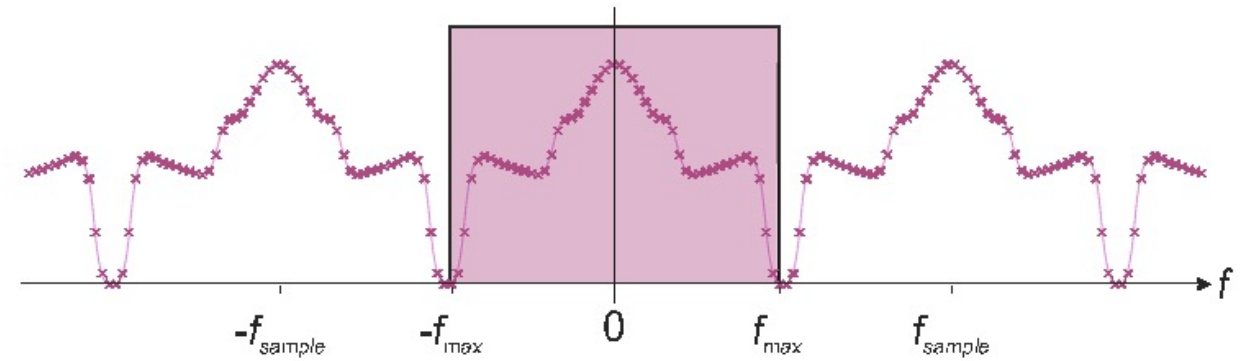
If **sampling** is just right, spectra just **touch**

Minimum sampling frequency = $2 \times \text{max}$

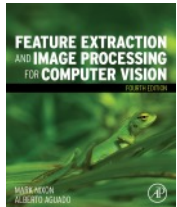
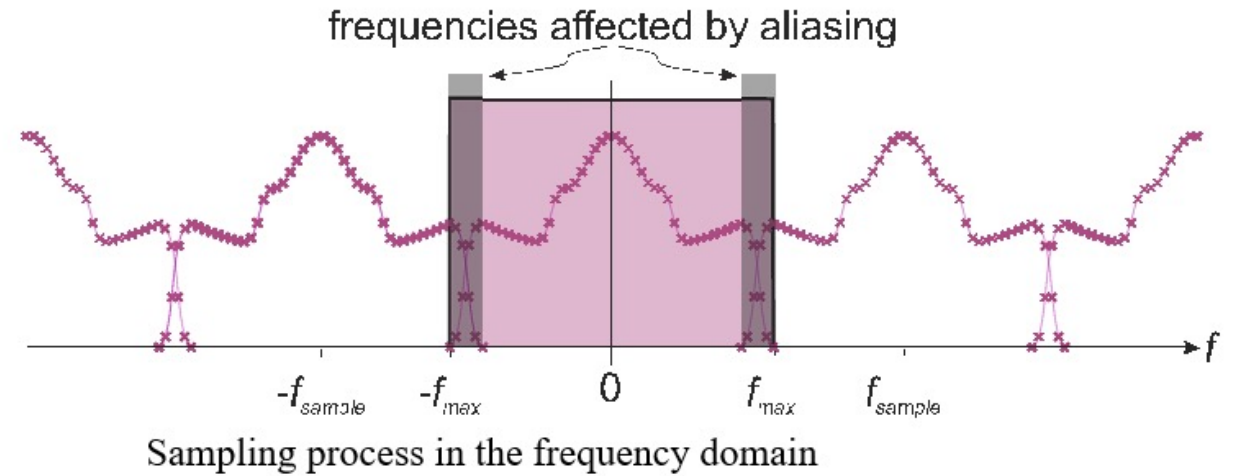
(a) Sampling at high frequency



(b) Sampling at the Nyquist frequency



(c) Sampling at low frequency, aliasing the data



Sampling theory

Nyquist's sampling theorem

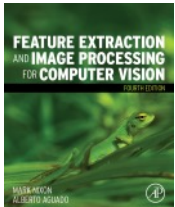
In order to be able to reconstruct a signal from its samples we must sample at minimum at twice the maximum frequency in the original signal

E.g. speech 6kHz, sample at 12 kHz

Video bandwidth (CCIR) is 5MHz

Sampling at 10MHz gave 576×576 images

Guideline: “two pixels for every pixel of interest”



Aliasing in Sampled Imagery



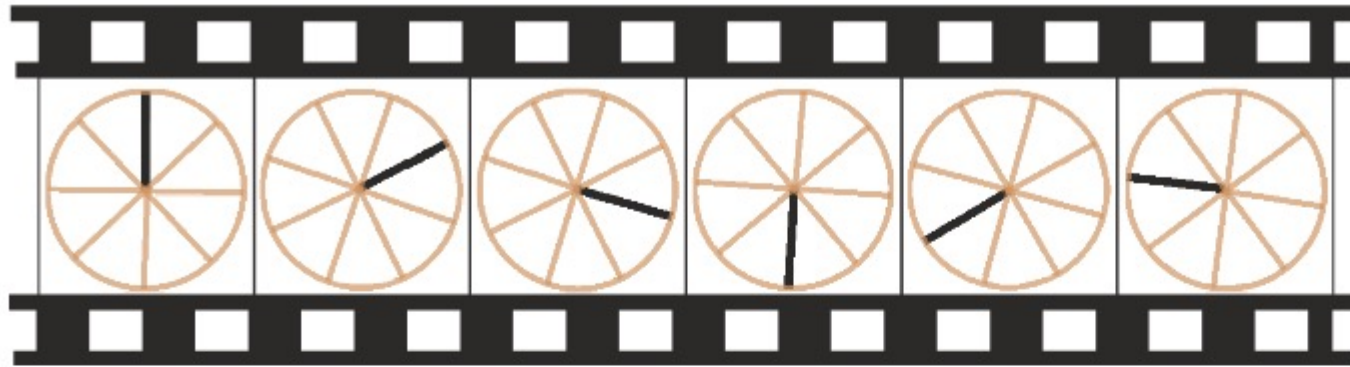
(a) high resolution



(c) low resolution – aliased



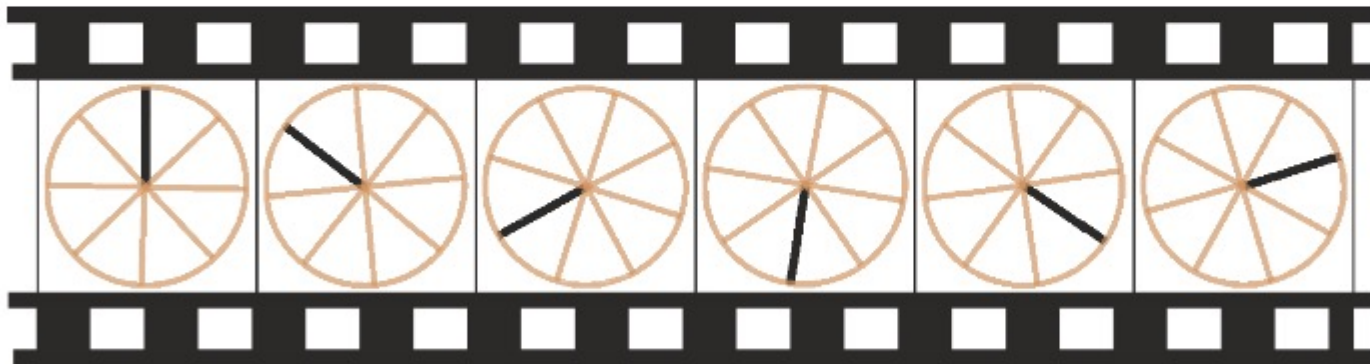
Correct and Incorrect Apparent Wheel Motion



(a) Oversampled rotating wheel



(b) Slow rotation



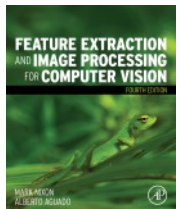
(c) Undersampled rotating wheel



(d) Fast rotation

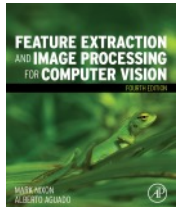
Figure 4.5 Correct and incorrect apparent wheel motion

<https://www.youtube.com/watch?v=e1EqXE06xr8>



Main points so far

- 1 – need to **sample** at a high enough frequency
 - 2 – **aliasing** corrupts image information
 - 3 – **discrete Fourier** allows analysis and understanding
 - 4 – Fourier has many **properties** and advantages
- but it's complex. So we'll move on to processing images



More sampling theories

Compressed sensing

Many signals are
sparse...

Regularisation