



#### Thinking in Frequency

Computer Vision
Brown

James Hays

#### Recap of Wednesday

linear filtering convolution differential filters filter types boundary conditions.

#### Review: questions

1. Write down a 3x3 filter that returns a positive value if the average value of the 4-adjacent neighbors is less than the center and a negative value otherwise

2. Write down a filter that will compute the gradient in the x-direction:

```
gradx(y,x) = im(y,x+1)-im(y,x) for each x, y
```

Slide: Hoiem

## Review: questions

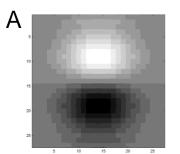
3. Fill in the blanks:

a) 
$$_{-}$$
 = D  $*$  E

$$C) F = D *$$

$$d) = D * D$$

Filtering Operator





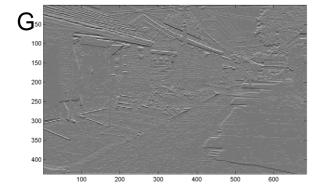


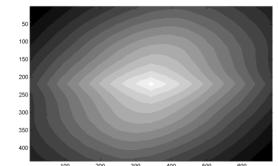
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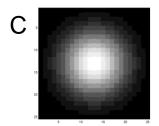
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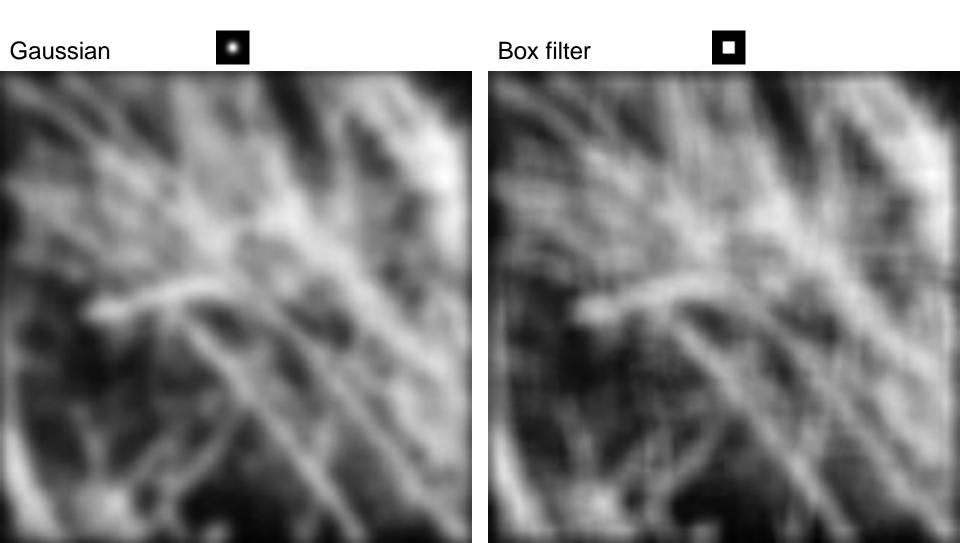
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## Today's Class

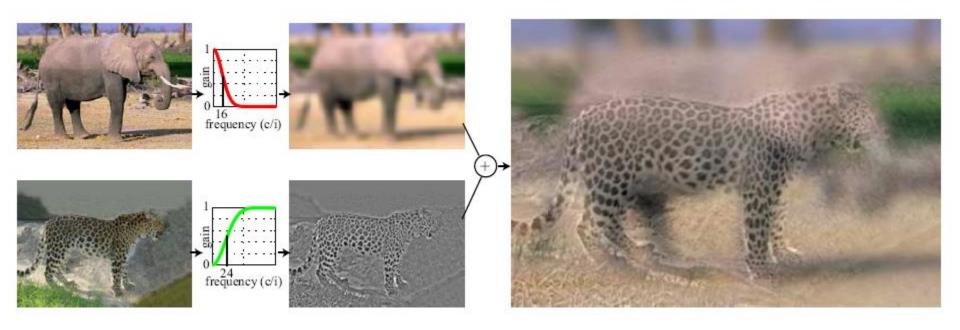
- Fourier transform and frequency domain
  - Frequency view of filtering
  - Hybrid images
  - Sampling
- Reminder: Read your textbook
  - Today's lecture covers material in 3.4

Slide: Hoiem

# Why does the Gaussian give a nice smooth image, but the square filter give edgy artifacts?

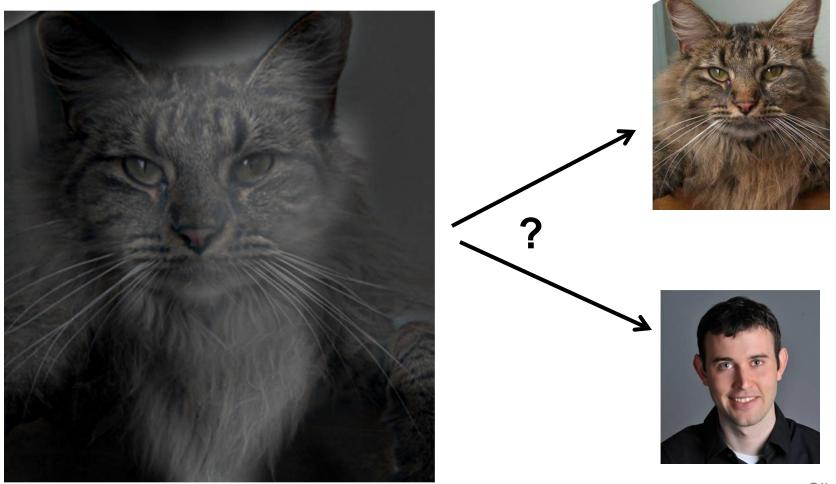


#### Hybrid Images



 A. Oliva, A. Torralba, P.G. Schyns, <u>"Hybrid Images,"</u> SIGGRAPH 2006

# Why do we get different, distance-dependent interpretations of hybrid images?



Slide: Hoiem

# Why does a lower resolution image still make sense to us? What do we lose?



Image: <a href="http://www.flickr.com/photos/igorms/136916757/">http://www.flickr.com/photos/igorms/136916757/</a>

Slide: Hoiem

#### Thinking in terms of frequency

#### Jean Baptiste Joseph Fourier (1768-1830)

#### had crazy idea (1807):

**Any** univariate function can rewritten as a weighted sum sines and cosines of different frequencies.

• Don't believe it?

- Neither did Lagrange,
   Laplace, Poisson and
   other big wigs
- Not translated into English until 1878!
- But it's (mostly) true!
  - called Fourier Series
  - there are some subtle restrictions

...the manner in which the author arrives at these equations is not exempt of difficulties and...his analysis to integrate them still leaves something to be desired on the score of generality and even rigour.

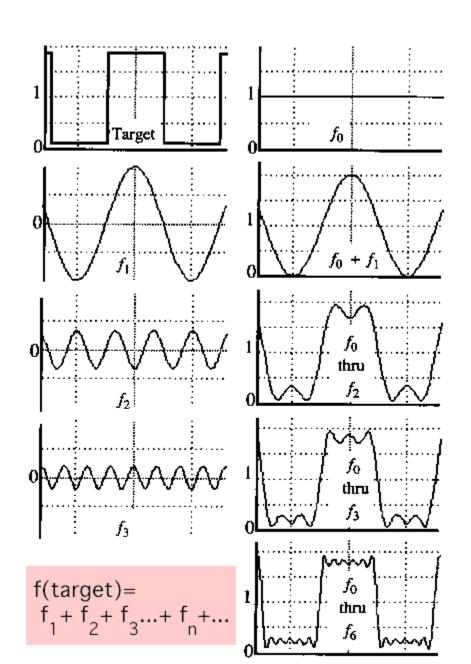


#### A sum of sines

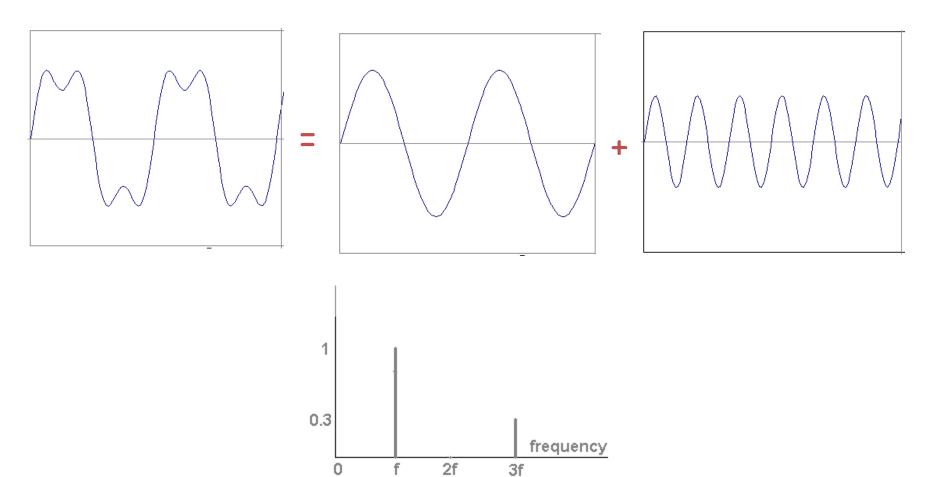
Our building block:

$$A\sin(\omega x + \phi)$$

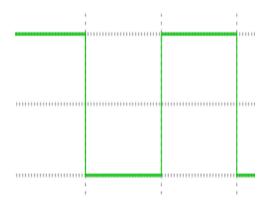
Add enough of them to get any signal g(x) you want!

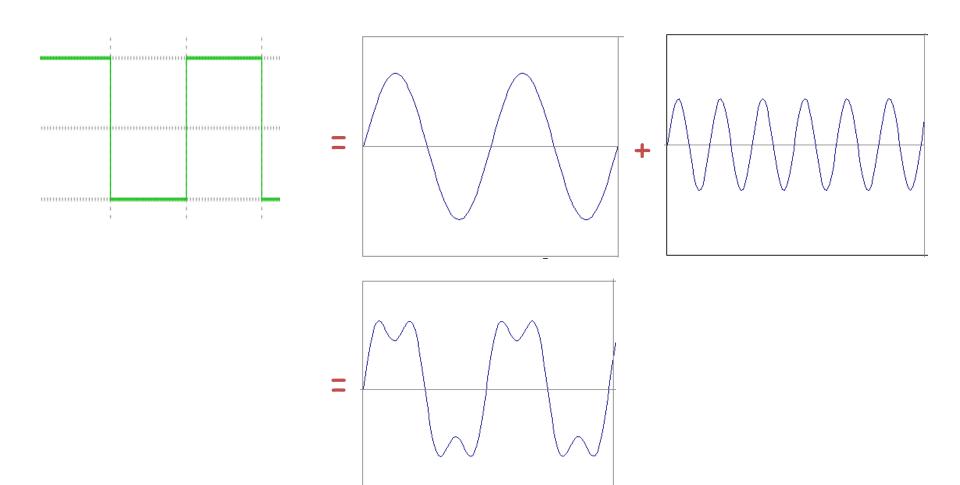


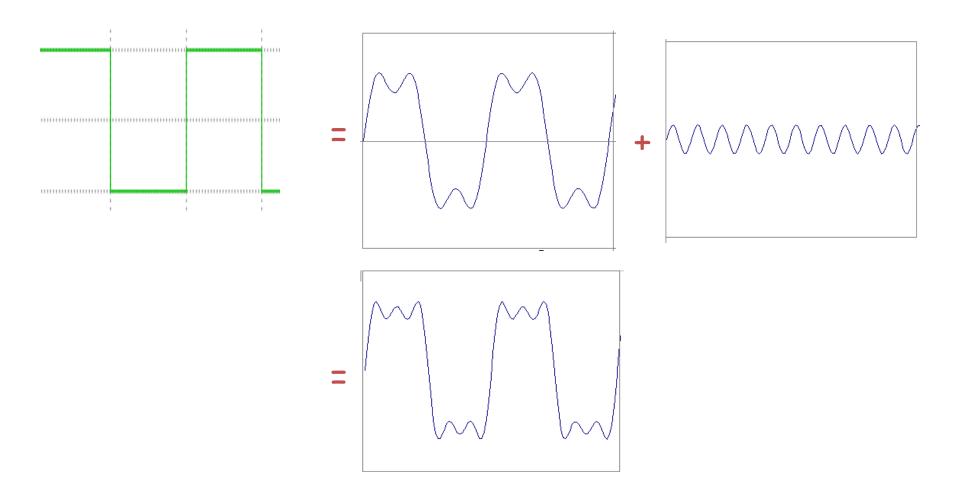
• example :  $g(t) = \sin(2\pi f t) + (1/3)\sin(2\pi(3f) t)$ 

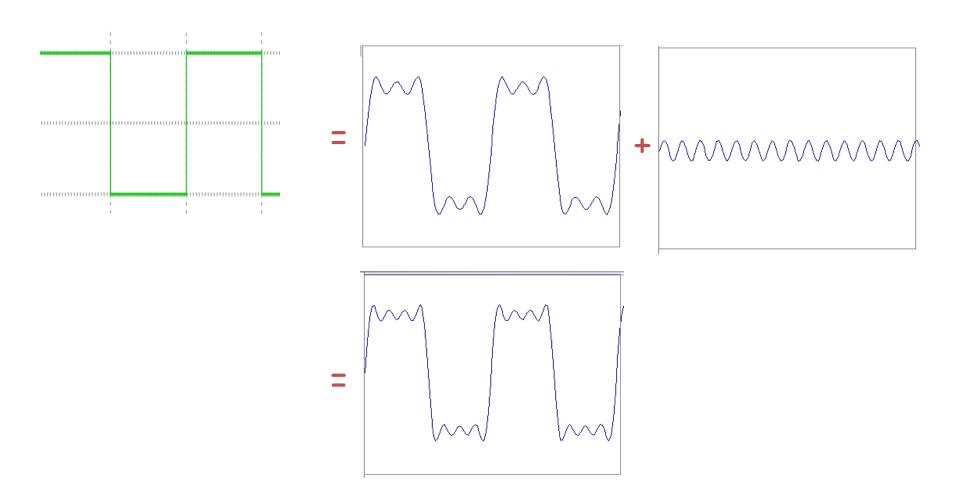


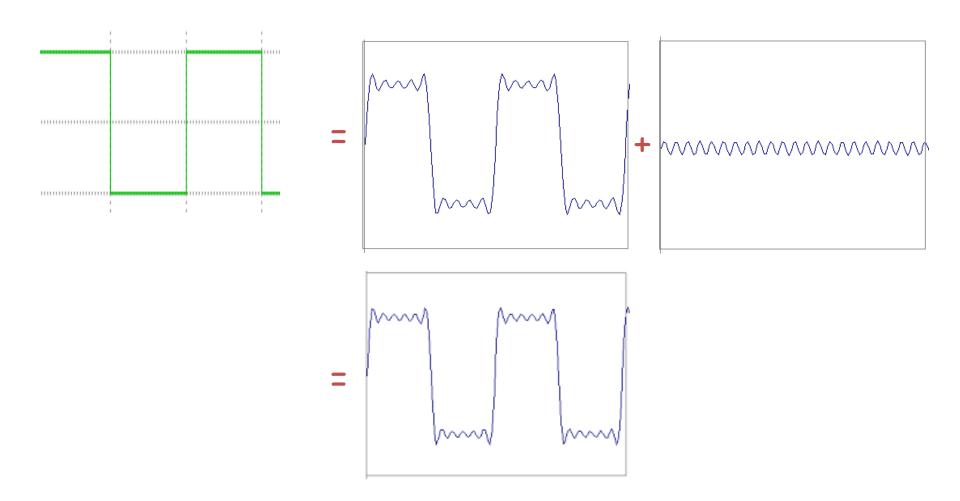
Slides: Efros

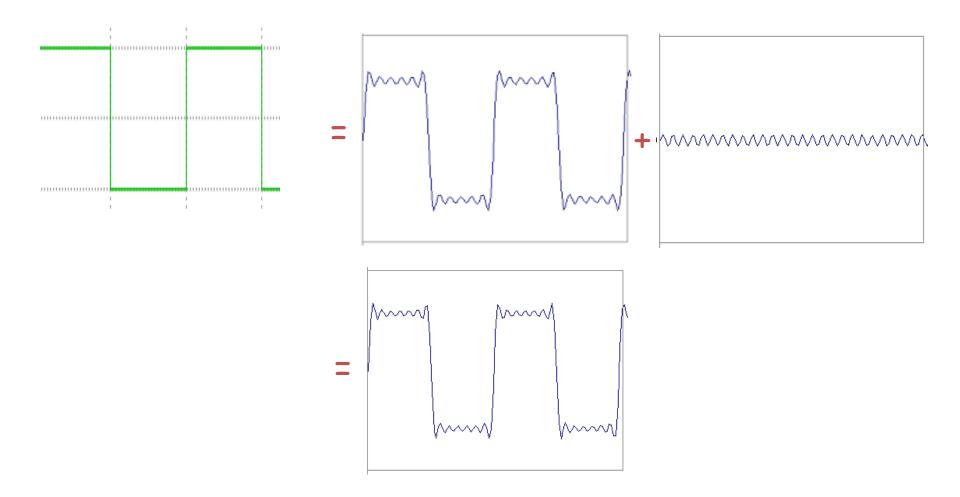


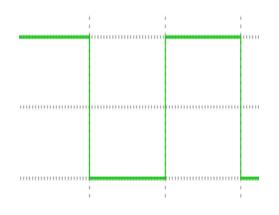




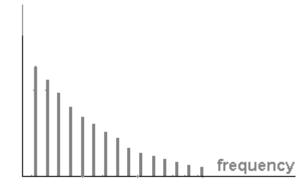






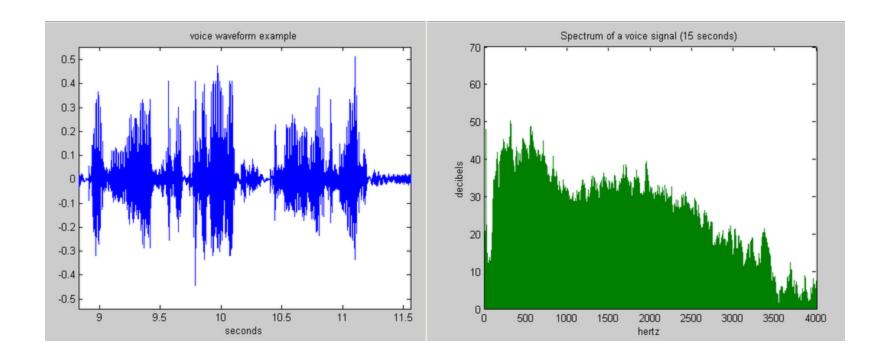


$$A\sum_{k=1}^{\infty} \frac{1}{k} \sin(2\pi kt)$$



## Example: Music

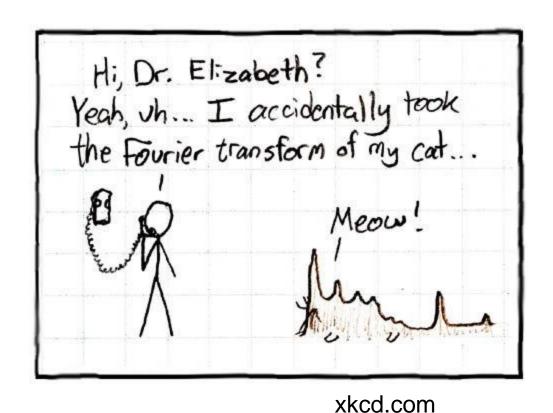
 We think of music in terms of frequencies at different magnitudes



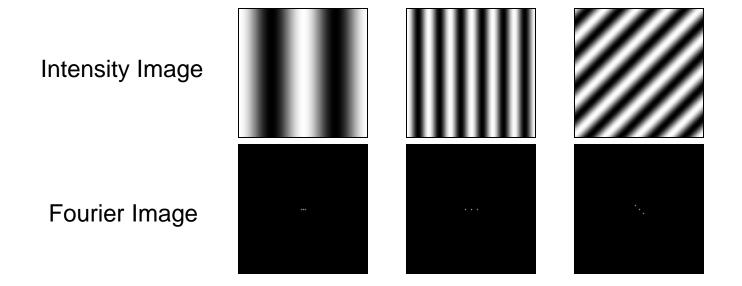
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#### Other signals

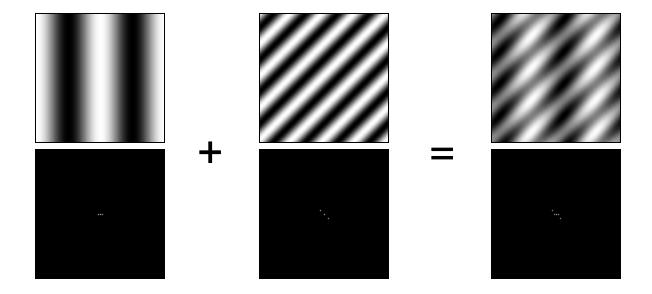
 We can also think of all kinds of other signals the same way



# Fourier analysis in images



#### Signals can be composed



http://sharp.bu.edu/~slehar/fourier/fourier.html#filtering More: http://www.cs.unm.edu/~brayer/vision/fourier.html

#### **Fourier Transform**

- Fourier transform stores the magnitude and phase at each frequency
  - Magnitude encodes how much signal there is at a particular frequency
  - Phase encodes spatial information (indirectly)
  - For mathematical convenience, this is often notated in terms of real and complex numbers

Amplitude: 
$$A = \pm \sqrt{R(\omega)^2 + I(\omega)^2}$$
 Phase:  $\phi = \tan^{-1} \frac{I(\omega)}{R(\omega)}$ 

#### The Convolution Theorem

 The Fourier transform of the convolution of two functions is the product of their Fourier transforms

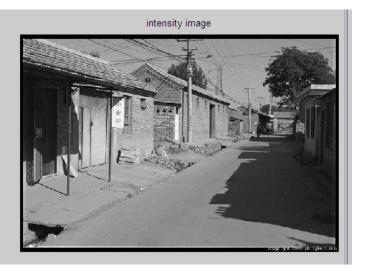
$$F[g * h] = F[g]F[h]$$

 Convolution in spatial domain is equivalent to multiplication in frequency domain!

$$g * h = F^{-1}[F[g]F[h]]$$

# Filtering in spatial domain

1	0	-1
2	0	-2
1	0	-1







# Filtering in frequency domain **FFT** log fft magnitude FFT Inverse FFT Slide: Hoiem

#### Fourier Matlab demo

#### FFT in Matlab

#### Filtering with fft

```
im = double(imread('...'))/255;
im = rgb2gray(im); % "im" should be a gray-scale floating point image
[imh, imw] = size(im);
hs = 50; % filter half-size
fil = fspecial('gaussian', hs*2+1, 10);
fftsize = 1024; % should be order of 2 (for speed) and include padding
im fft = fft2(im, fftsize, fftsize);
                                                           % 1) fft im with padding
fil fft = fft2(fil, fftsize, fftsize);
                                                           % 2) fft fil, pad to same size as
image
im fil fft = im fft .* fil fft;
                                                           % 3) multiply fft images
im fil = ifft2(im fil fft);
                                                          % 4) inverse fft2
im fil = im fil(1+hs:size(im,1)+hs, 1+hs:size(im, 2)+hs); % 5) remove padding
```

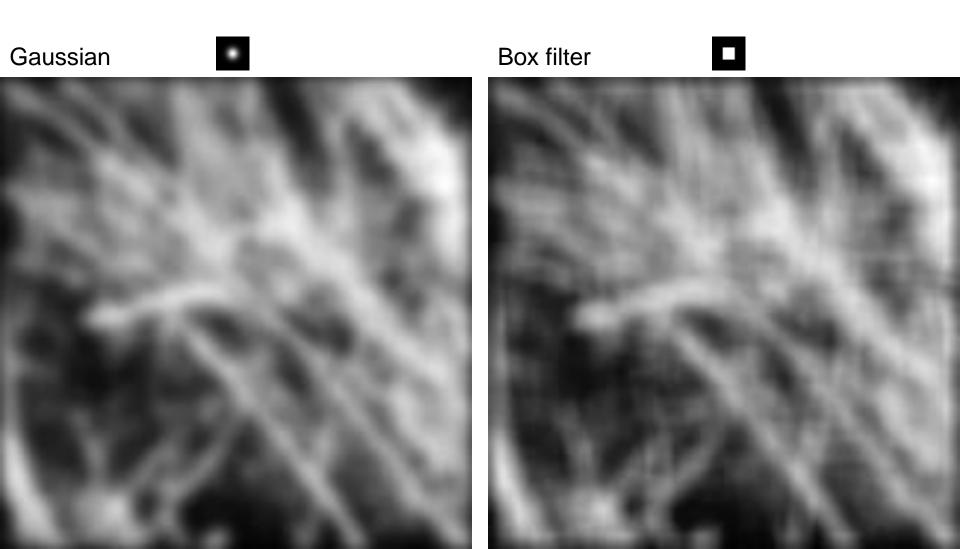
#### Displaying with fft

```
figure(1), imagesc(log(abs(fftshift(im_fft)))), axis image, colormap jet
```

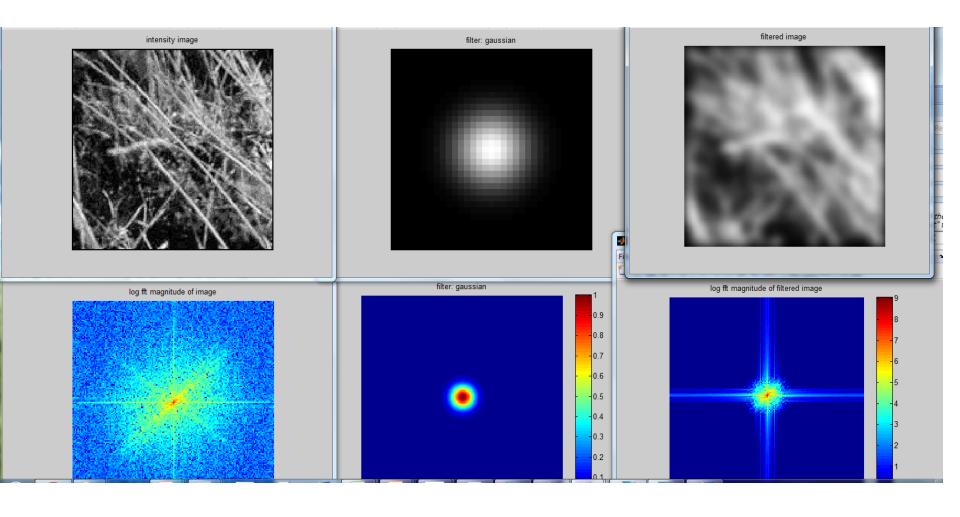
Slide: Hoiem

Filtering

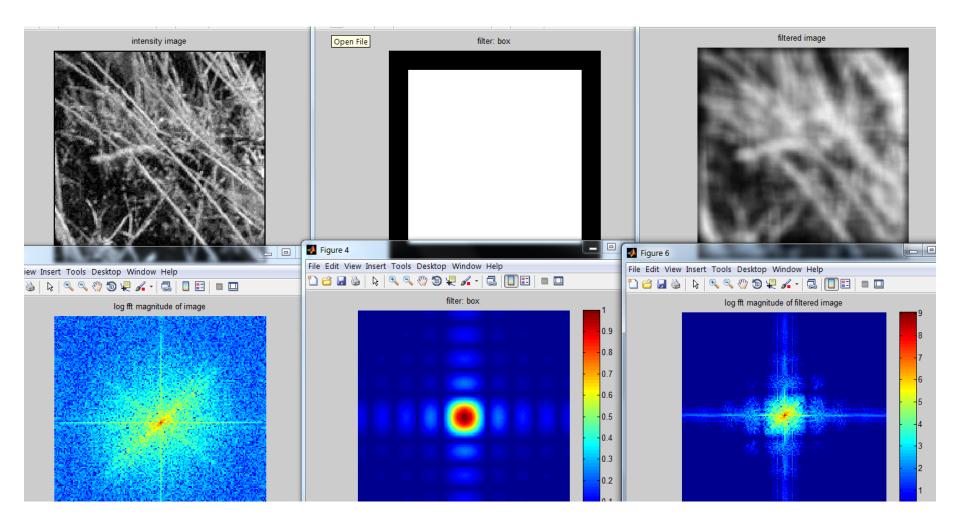
Why does the Gaussian give a nice smooth image, but the square filter give edgy artifacts?



#### Gaussian



#### **Box Filter**

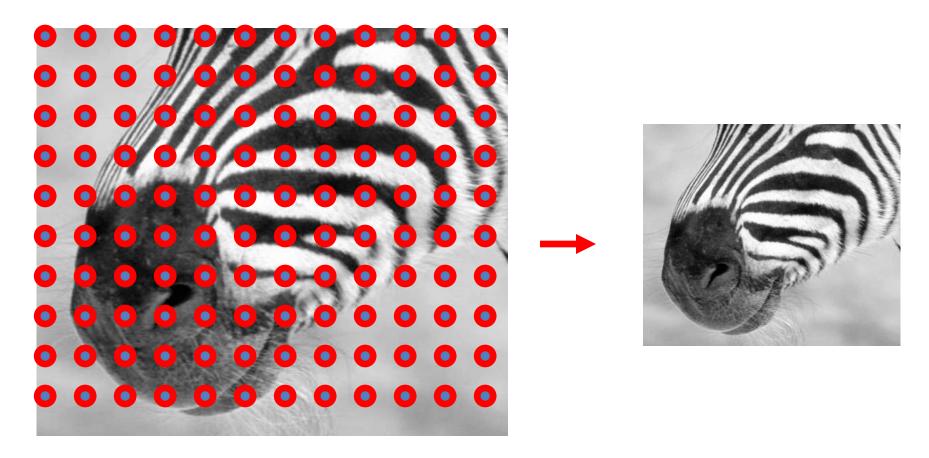


#### Sampling

# Why does a lower resolution image still make sense to us? What do we lose?



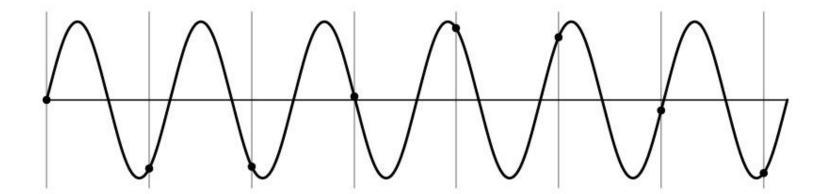
# Subsampling by a factor of 2



Throw away every other row and column to create a 1/2 size image

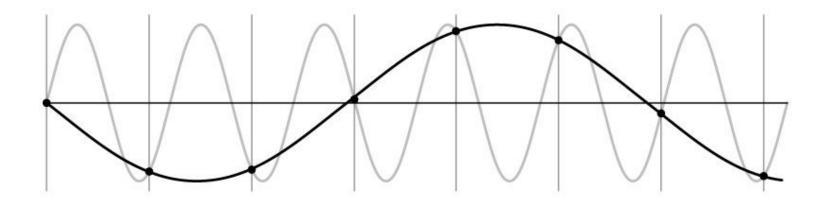
# Aliasing problem

1D example (sinewave):



## Aliasing problem

• 1D example (sinewave):



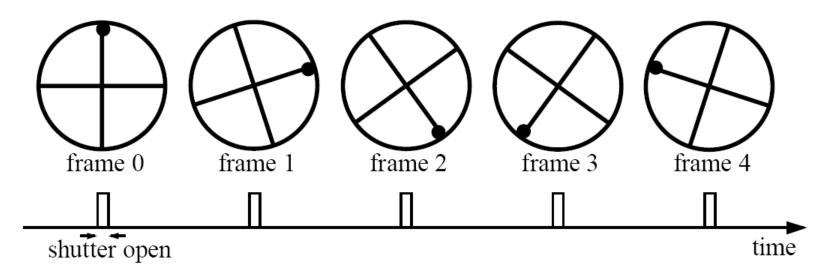
## Aliasing problem

- Sub-sampling may be dangerous....
- Characteristic errors may appear:
  - "Wagon wheels rolling the wrong way in movies"
  - "Checkerboards disintegrate in ray tracing"
  - "Striped shirts look funny on color television"

### Aliasing in video

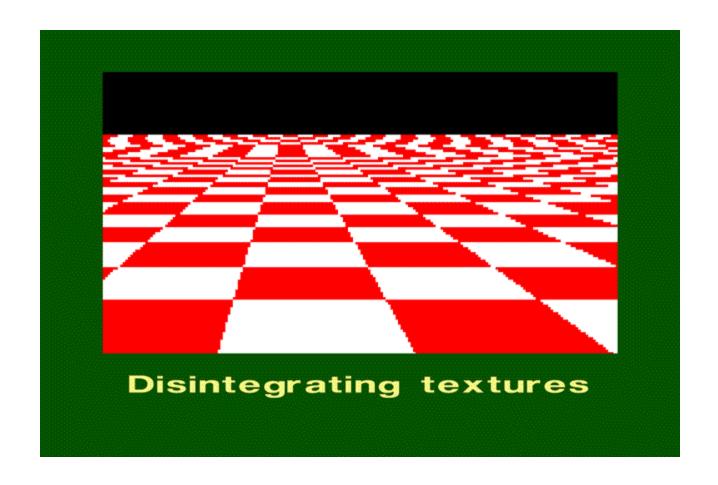
Imagine a spoked wheel moving to the right (rotating clockwise). Mark wheel with dot so we can see what's happening.

If camera shutter is only open for a fraction of a frame time (frame time = 1/30 sec. for video, 1/24 sec. for film):

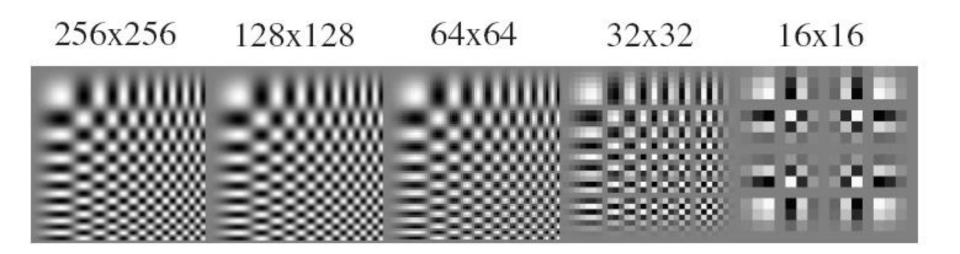


Without dot, wheel appears to be rotating slowly backwards! (counterclockwise)

### Aliasing in graphics

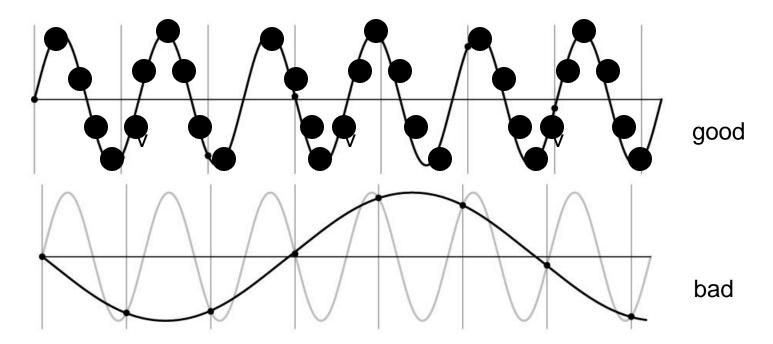


# Sampling and aliasing



## Nyquist-Shannon Sampling Theorem

- When sampling a signal at discrete intervals, the sampling frequency must be  $\geq 2 \times f_{max}$
- f<sub>max</sub> = max frequency of the input signal
- This will allows to reconstruct the original perfectly from the sampled version



## Anti-aliasing

### Solutions:

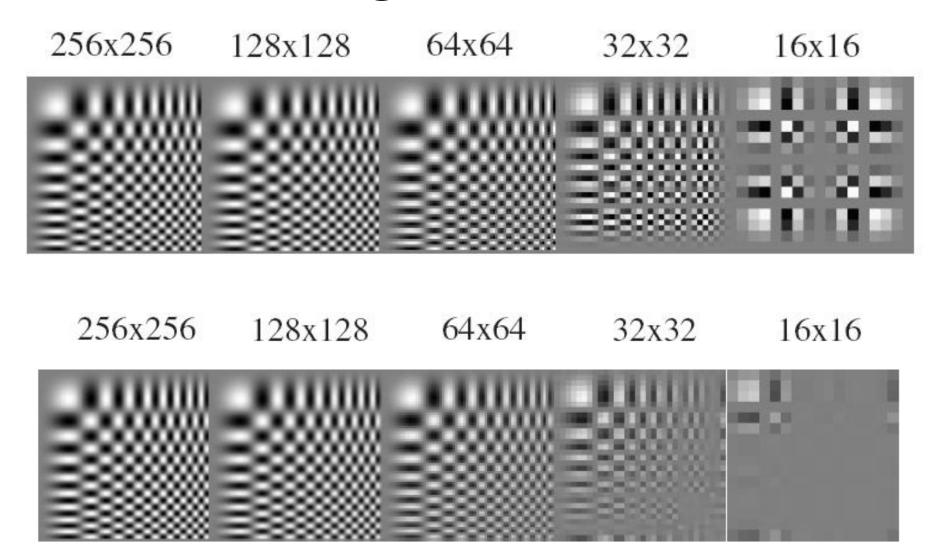
Sample more often

- Get rid of all frequencies that are greater than half the new sampling frequency
  - Will lose information
  - But it's better than aliasing
  - Apply a smoothing filter

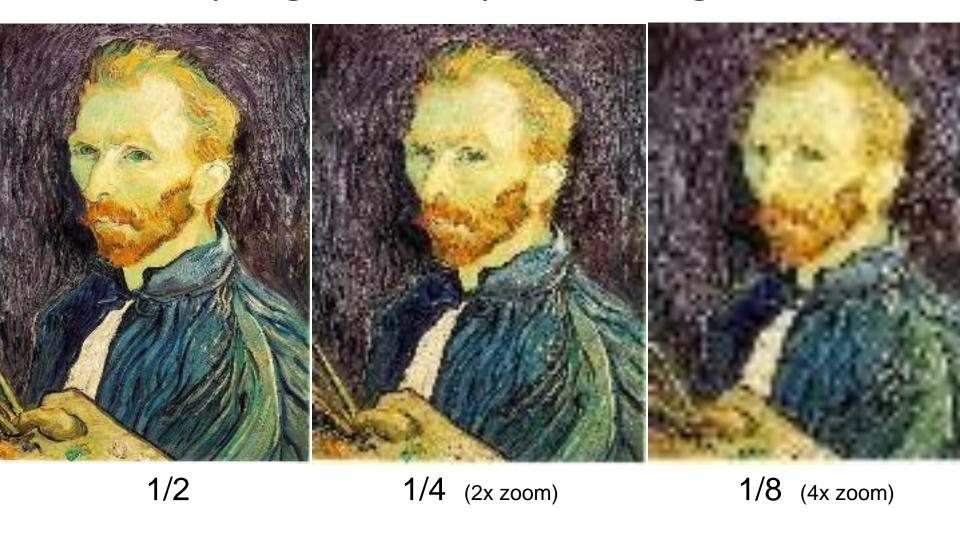
### Algorithm for downsampling by factor of 2

- 1. Start with image(h, w)
- 2. Apply low-pass filter
  im\_blur = imfilter(image, fspecial('gaussian', 7, 1))
- 3. Sample every other pixel
   im\_small = im\_blur(1:2:end, 1:2:end);

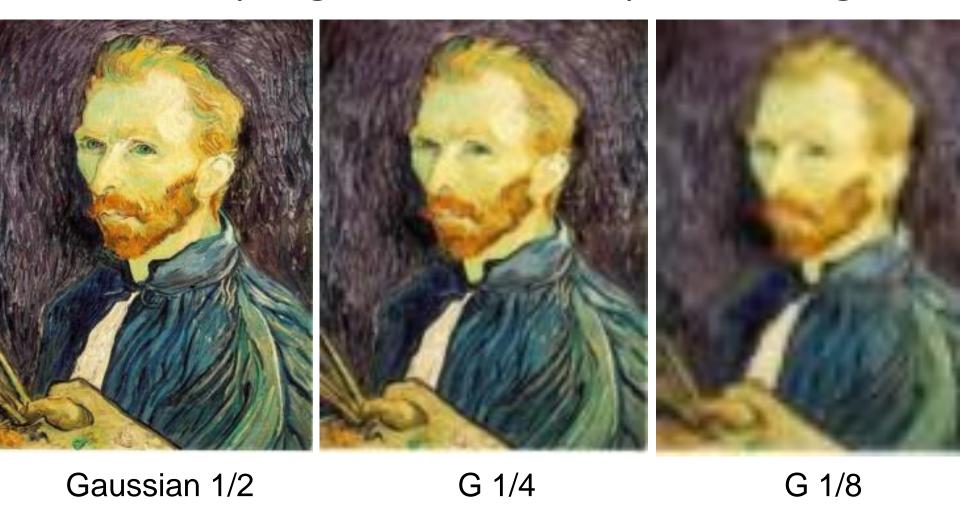
# Anti-aliasing



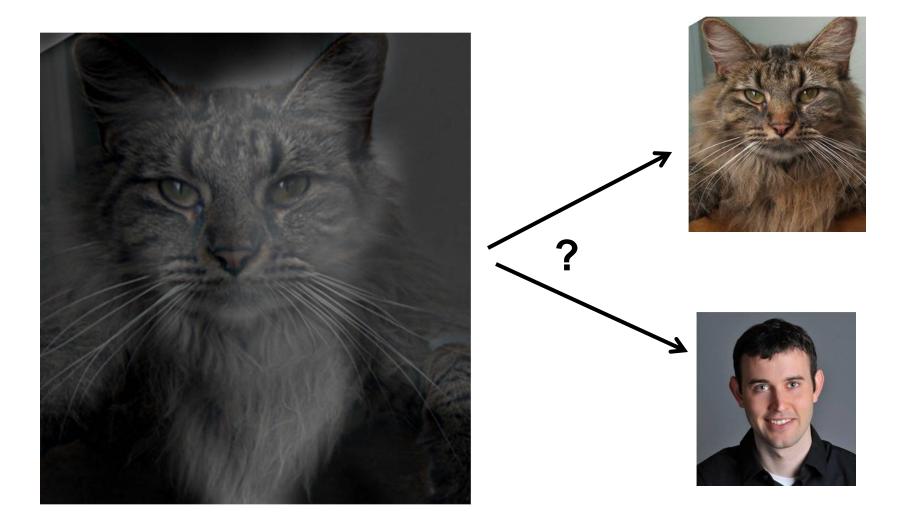
### Subsampling without pre-filtering

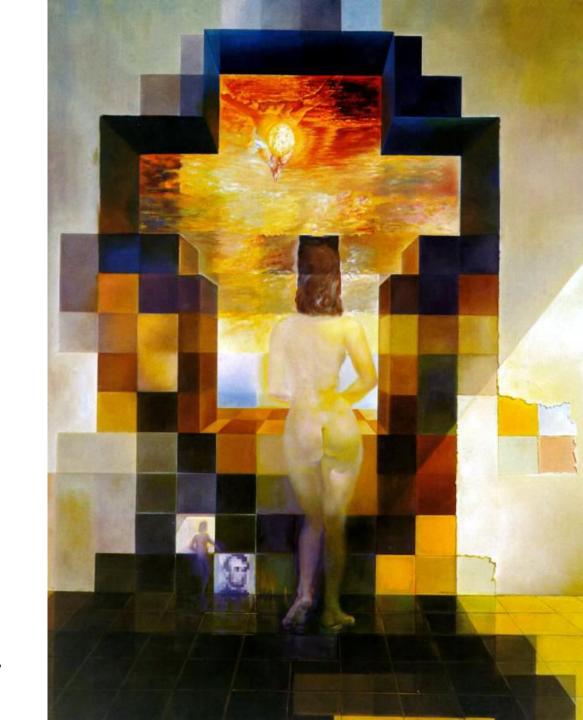


## Subsampling with Gaussian pre-filtering



# Why do we get different, distance-dependent interpretations of hybrid images?

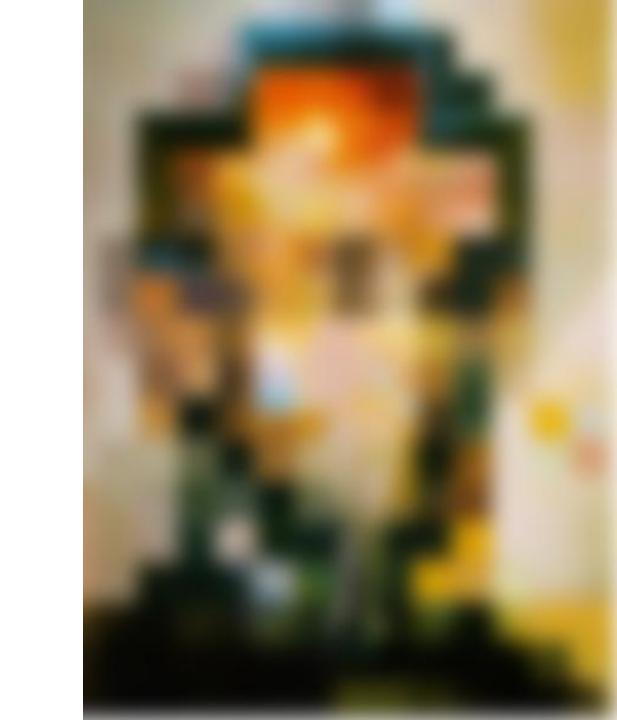


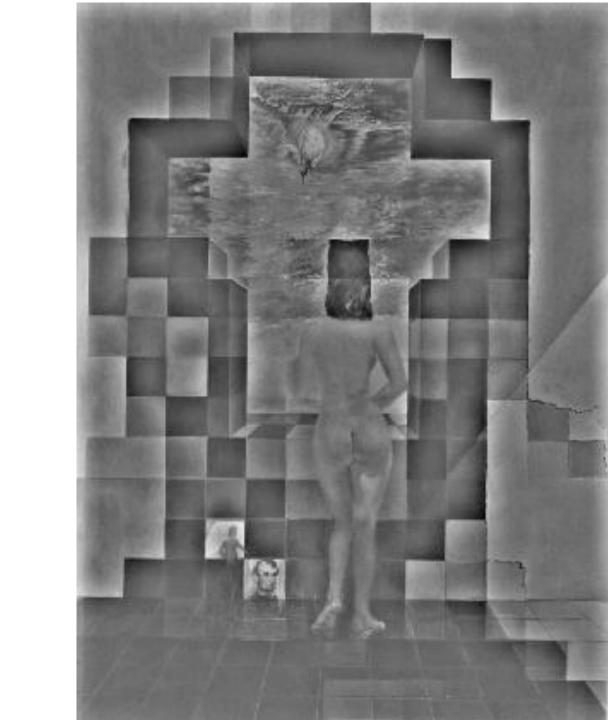


#### Salvador Dali invented Hybrid Images?

#### Salvador Dali

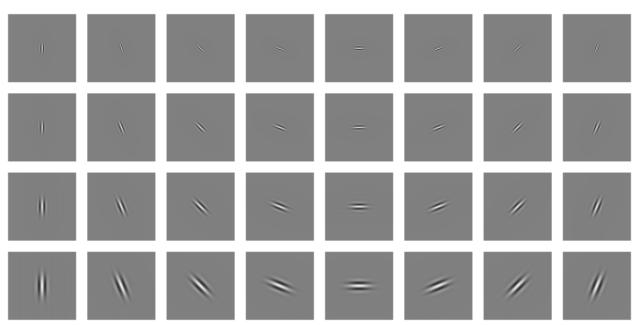
"Gala Contemplating the Mediterranean Sea, which at 30 meters becomes the portrait of Abraham Lincoln", 1976





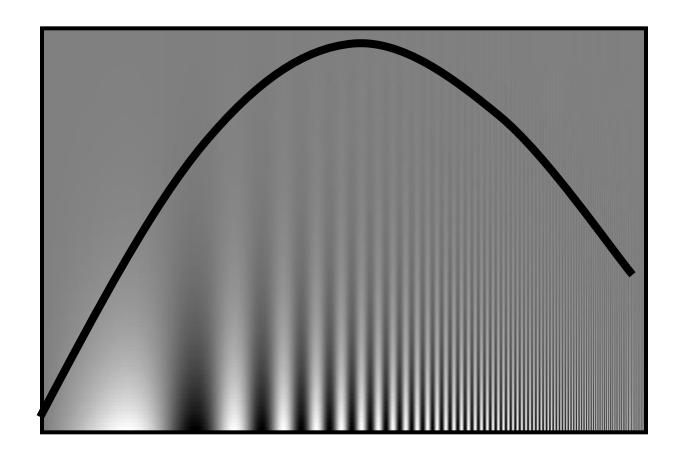
### Clues from Human Perception

- Early processing in humans filters for various orientations and scales of frequency
- Perceptual cues in the mid-high frequencies dominate perception
- When we see an image from far away, we are effectively subsampling it

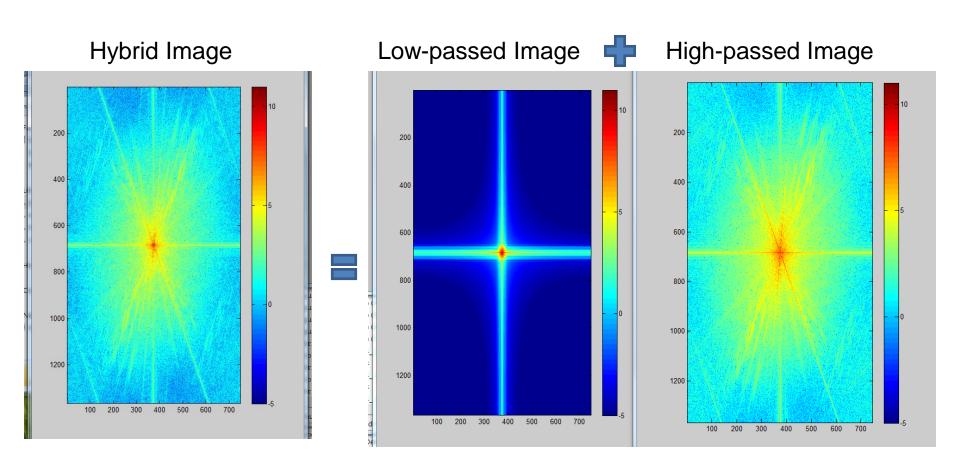


Early Visual Processing: Multi-scale edge and blob filters

### Campbell-Robson contrast sensitivity curve

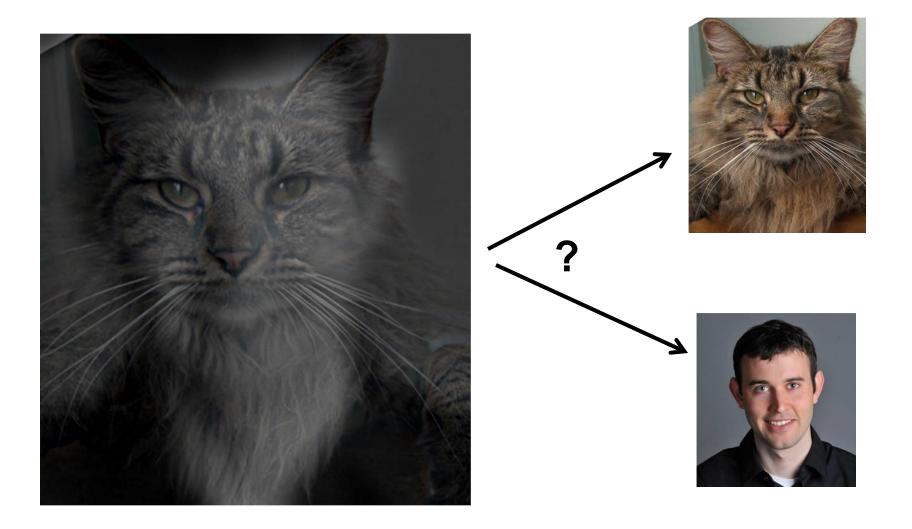


# Hybrid Image in FFT



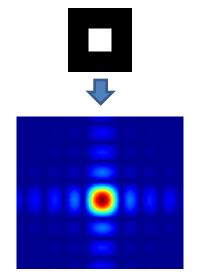
### Perception

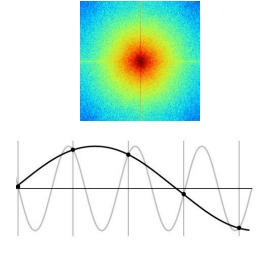
# Why do we get different, distance-dependent interpretations of hybrid images?



## Things to Remember

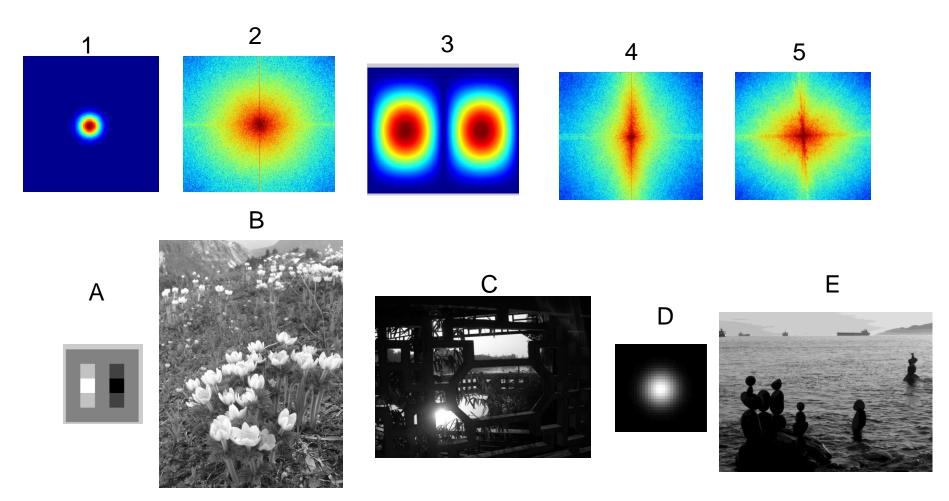
- Sometimes it makes sense to think of images and filtering in the frequency domain
  - Fourier analysis
- Can be faster to filter using FFT for large images (N logN vs. N<sup>2</sup> for autocorrelation)
- Images are mostly smooth
  - Basis for compression
- Remember to low-pass before sampling





### Practice question

1. Match the spatial domain image to the Fourier magnitude image



### Next class

Template matching

Image Pyramids

Filter banks and texture

# Questions