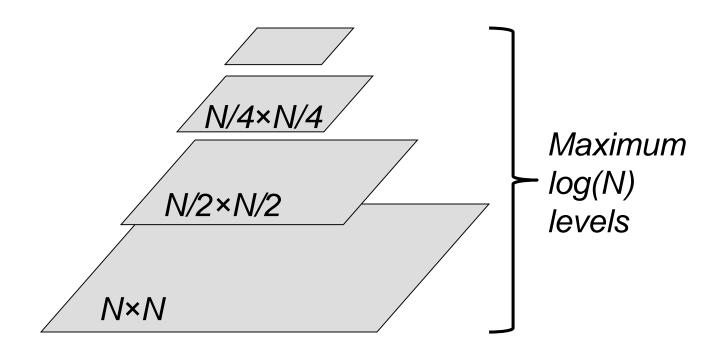
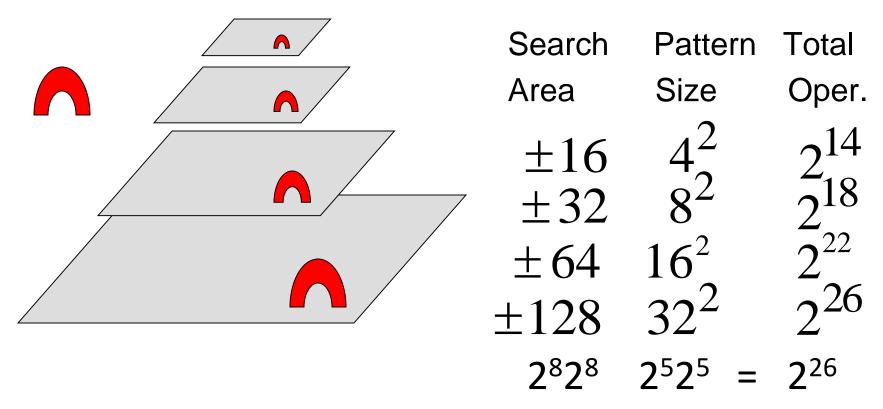
## Image Pyramids



$$N^2 + \frac{1}{4}N^2 + \frac{1}{16}N^2 + \dots = 1\frac{1}{3}N^2$$

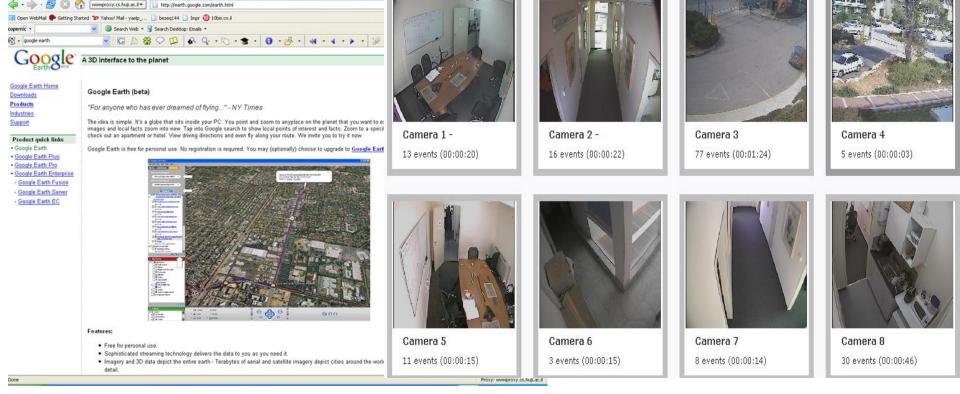
#### Efficient Visual Search



- Pyramids: Start the search in a small image
- •search area is small in larger levels (e.g. ±1) using the estimate from the smaller level

## **Applications for Pyramids**

- Detection and Search (Esp. huge images)
- Browsing in Image Databases
- Motion Computation



## Image Resizing

#### Reduce:

$$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

#### 1. Blur

-E.g. Convolve with a 3×3 filter  $(\frac{1}{4}, \frac{1}{2}, \frac{1}{4}) \times (\frac{1}{4}, \frac{1}{2}, \frac{1}{4})^T$ , or a 5×5 filter  $\frac{1}{16}(1, 4, 6, 4, 1) \times ...$  or larger

#### 2. Sub-sample

- Select only every 2<sup>nd</sup> pixel in every 2<sup>nd</sup> raw

#### **Expand:**

- 1. Zero Padding  $(a_1, 0, a_2, 0, a_3, 0, ...)$
- 2. Blur
  - Note: Expand blur needs different normalizations!
  - Is zero padding followed by blur with  $(\frac{1}{2}, 1, \frac{1}{2})$  OK?

#### Blur Kernels

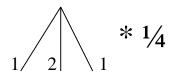
#### Commonly Used - Binomial Coefficients

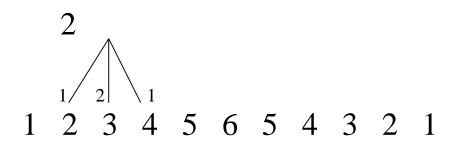
- Odd number of coefficients (have a center)
- Sum of coefficients normalized to 1
- Fast to compute (using shift and add)
- Asymptotically similar to a Gaussian

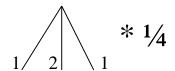
$$(1\ 1) *...*(1\ 1) / 2^{2k}$$

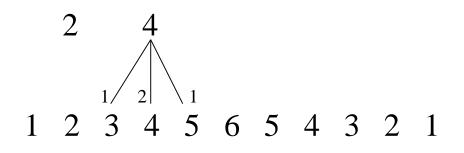
# Blur & Sub-sample (Reduce)

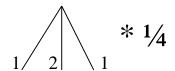


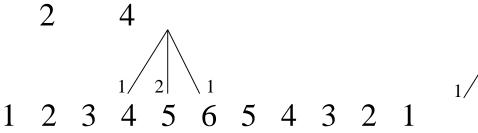




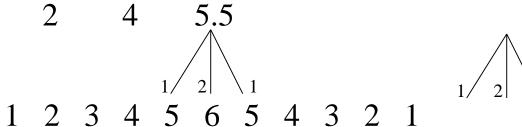




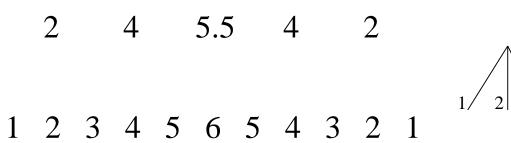




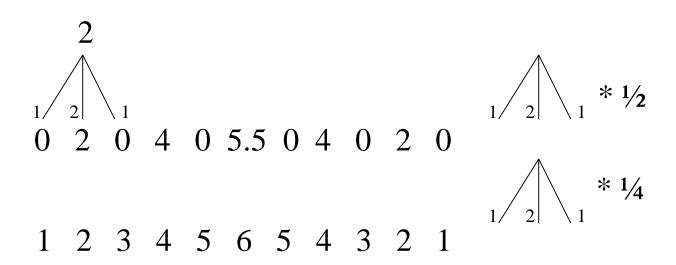


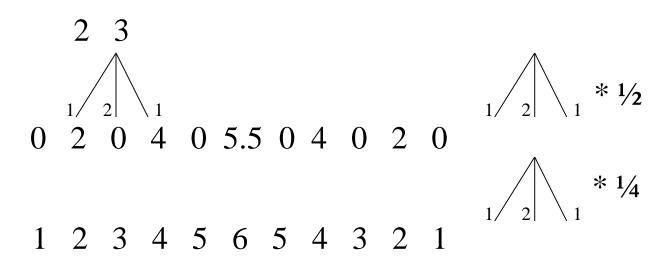


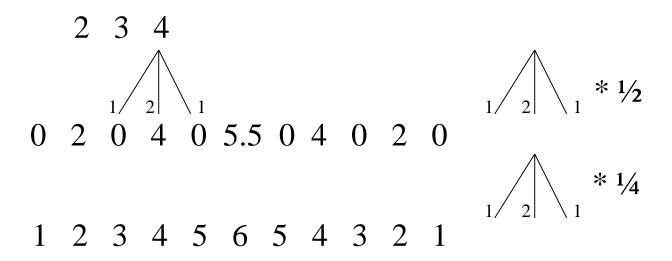


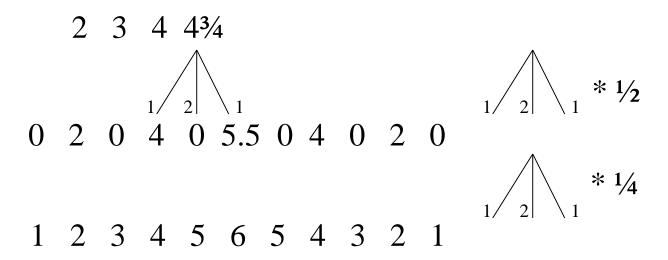


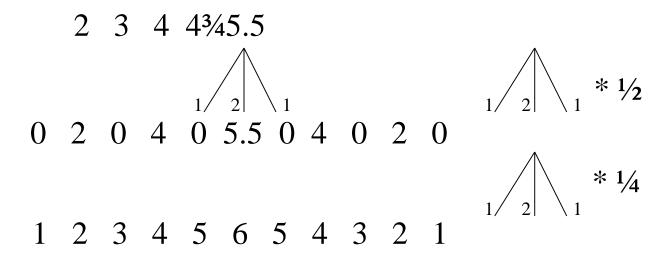
# Zero-Pad & Blur (Expand)

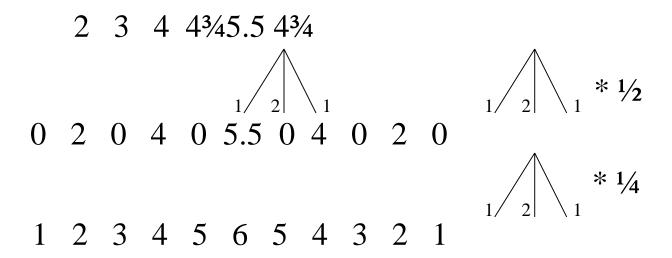


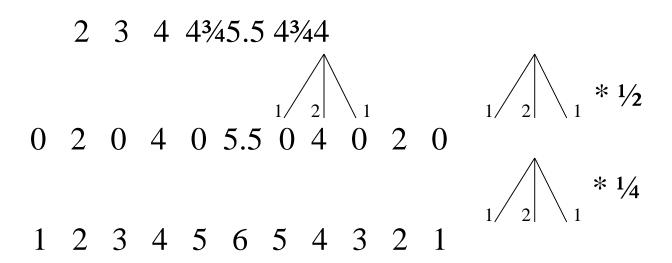


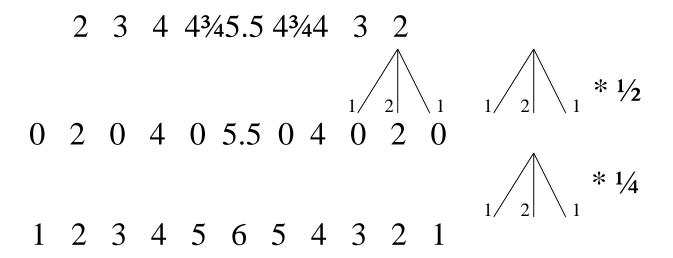




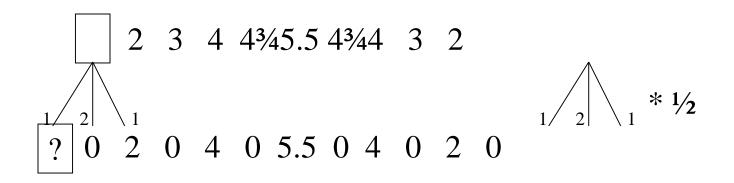








# Handling Image Boundaries Never Cyclic...



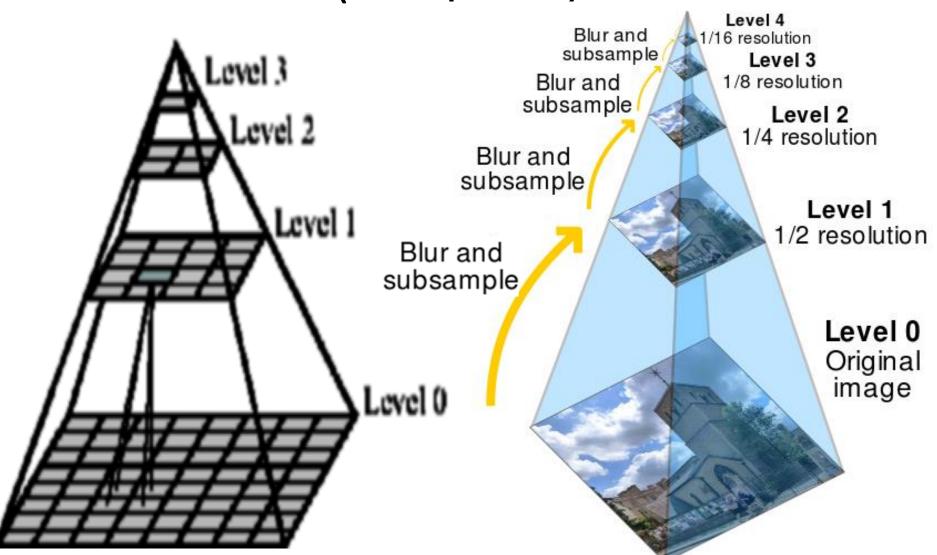
Mirror on last pixel. ?=2 Mirror after last pixel. ?=0 Duplicate last pixel. ?=0

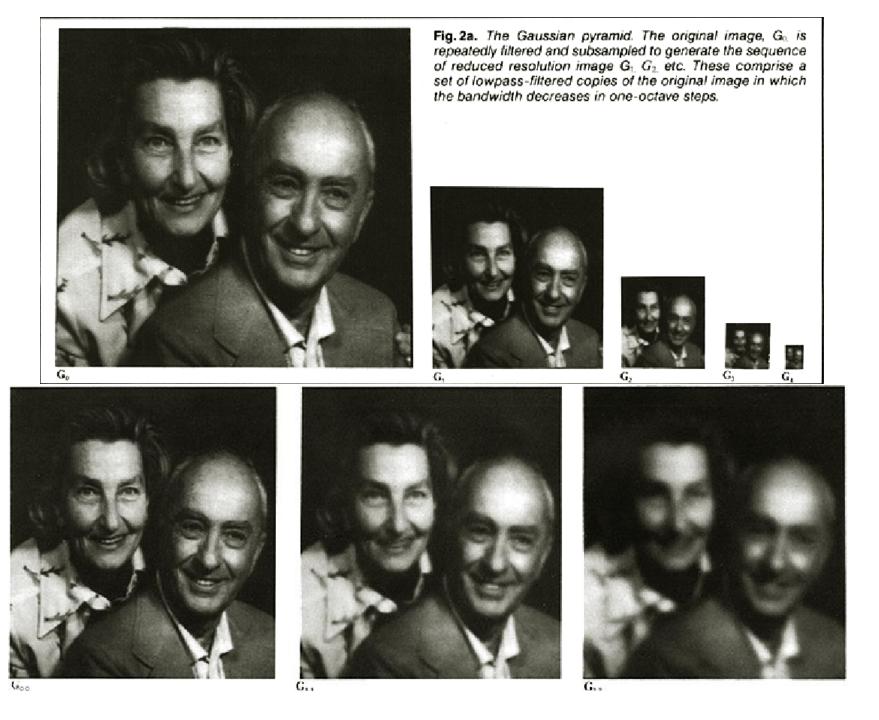
## Gaussian Pyramid

```
G_n - Reduce\{G_{n-1}\} G_3 — Reduce\{G_2\} G_2 — Reduce\{G_1\} G_1 — Reduce\{G_0\} G_0 — Original Image
```

- 2D Picture
  - Reduce Rows, Reduce Columns

# 5-Level Gaussian Pyramid (Wikipedia)





## Laplacian Pyramid

$$L_n + L_{n-1} = Expand\{L_n\} + L_{n-1} =$$
  
=  $Expand\{G_n\} + (G_{n-1} - Expand\{G_n\}) = G_{n-1}$ 

$$\sum_{i=k}^{n} (\mathbf{L}_i) = \mathbf{G}_k \qquad \sum_{i=0}^{n} (\mathbf{L}_i) = \mathbf{G}_0$$





 $G_1$ 

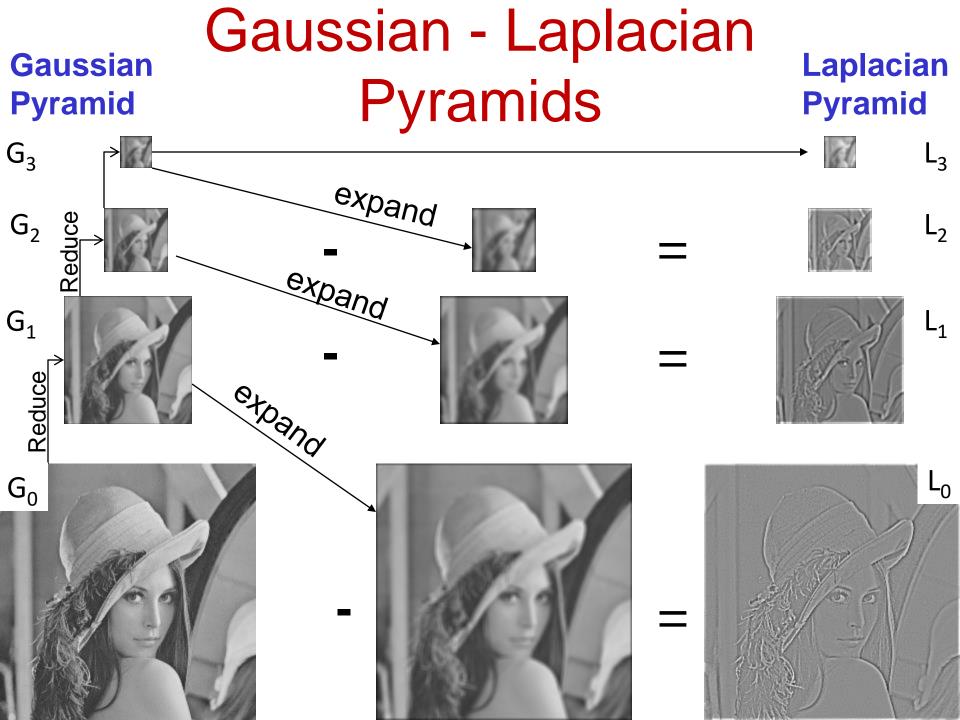


 $G_0$ 



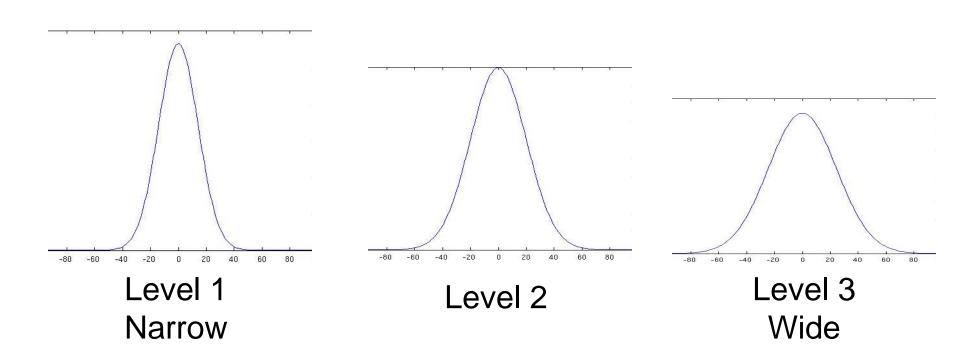
 $Expand\{G_1\}$ 

 $L_0 = G_0 - Expand\{G_1\}$ 



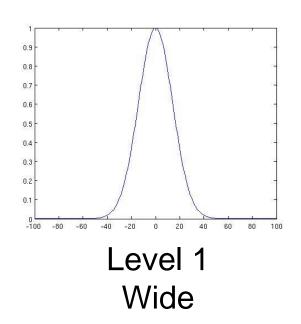
## Laplacian Pyramid as a Band-Pass Filter

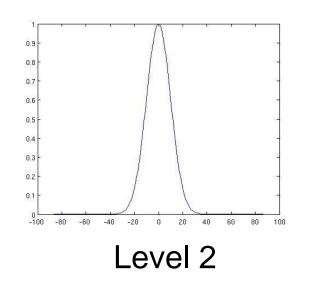
## Gaussian Pyramid – Convolution with a Gaussian filter

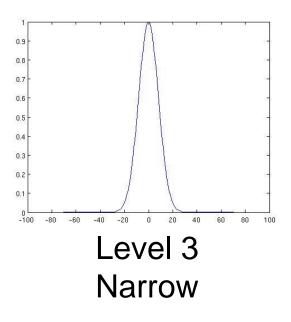


## Laplacian Pyramid as a Band-Pass Filter

Gaussian Pyramid (<u>in the Fourier domain</u>) – Multiplication with a Gaussian kernel

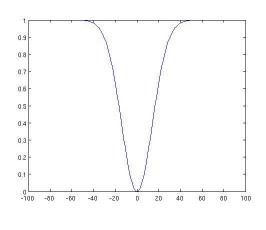


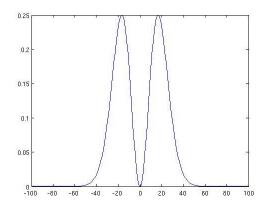


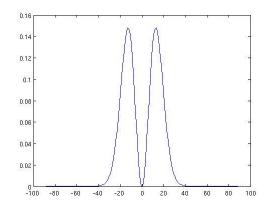


## Laplacian Pyramid as a Band-Pass Filter

 In the Fourier domain, the Laplacian is the difference between two powers of Gaussian kernels:







Level 1

Level 2

Level 3

## Pyramid Compression

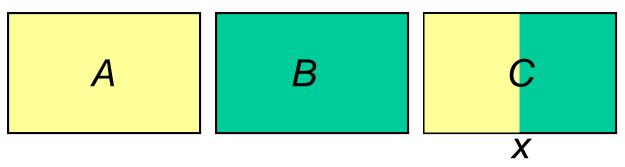
- Build a Laplacian Pyramid
- Quantize pyramid values to 3-5 values
  - Optimal Quantization
- Compress using Entropy Compression
  - (Huffman, Lempel-Ziv)
- Reconstruct normally
- Next Generation: Wavelet Compression

## **Pyramid Compression**



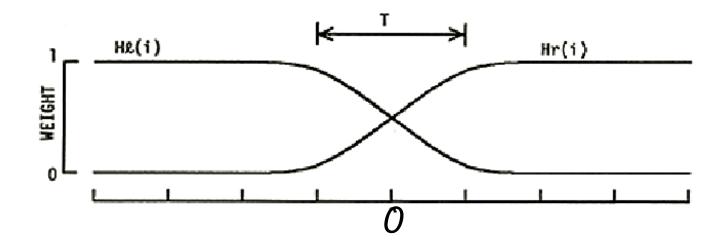
Fig. 5. Pyramid data compression. The original image represented at 8 bits perpixel is shown in (a). The node values of the Laplacian pyramid representation of this image were quantitized to obtain effective data rates of 1 b/p and 1/2 b/p. Reconstructed images (b) and (c) show relatively little degradation.

## Picture Merging with Spline



For every Row:

$$C(i) = H_l(i-x)A(i) + H_r(i-x)B(i)$$

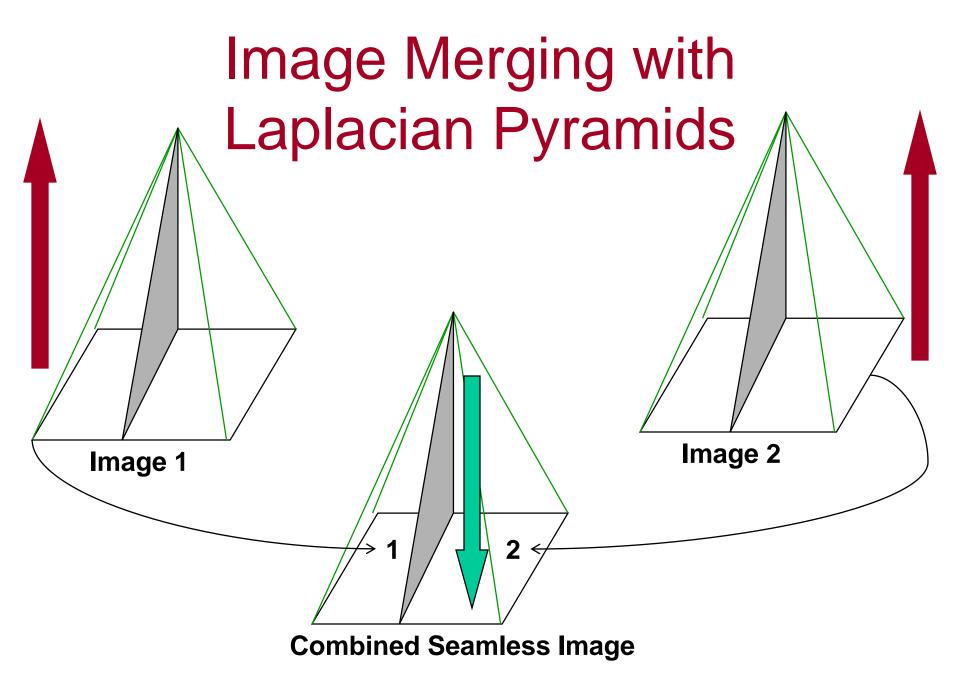


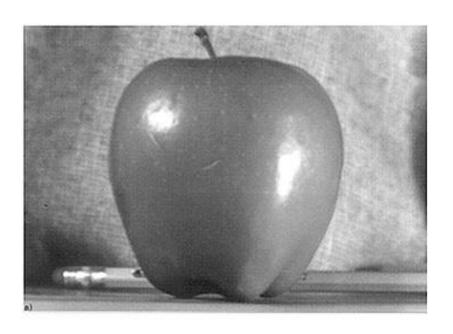
## Multiresolution Spline

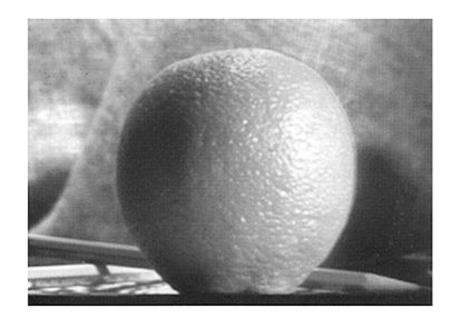
- Given two images A and B to be splined in middle
- Construct Laplacian Pyramid  $L_a$  and  $L_b$
- Create a third Laplacian Pyramid L<sub>c</sub> where for each level k

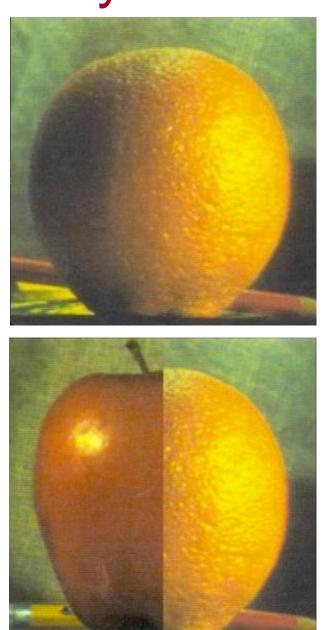
$$L_{c}(i,j) = \begin{cases} L_{a}(i,j) & \text{if } i < width/2 \\ (L_{a}(i,j) + L_{b}(i,j))/2 & \text{if } i = width/2 \\ L_{b}(i,j) & \text{if } i > width/2 \end{cases}$$

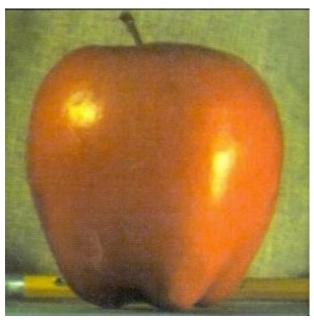
Sum all levels in L<sub>c</sub> to get the blended image

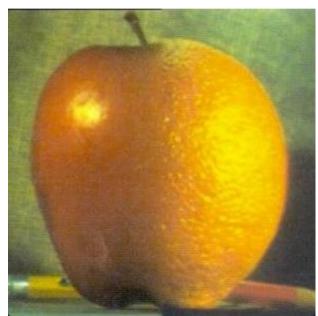












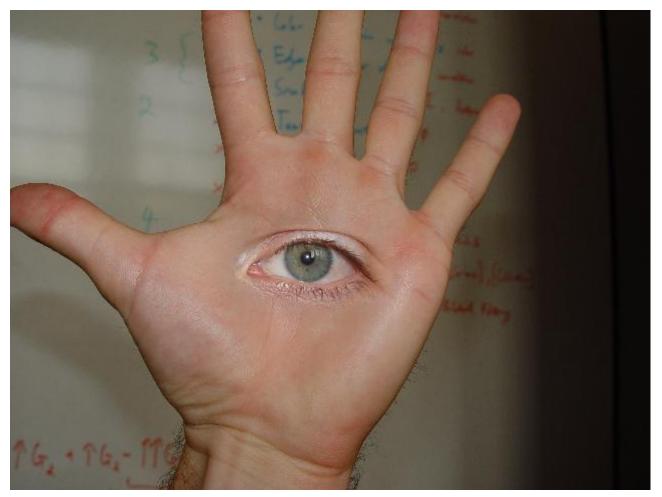
## Pyramid Blending Arbitrary Shape

- Given two images A and B, and a binary mask M
- Construct Laplacian Pyramids L<sub>a</sub> and L<sub>b</sub>
- Construct a Gaussian Pyramid G<sub>m</sub>
- Create a third Laplacian Pyramid L<sub>c</sub> where for each level k

$$L_c(i,j) = G_m(i,j)L_a(i,j) + (1 - G_m(i,j))L_b(i,j)$$

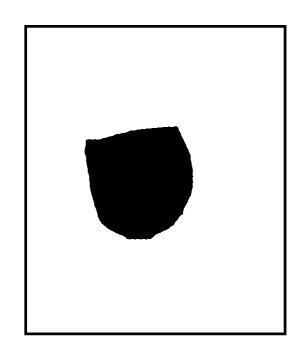
• Sum all levels  $L_c$  in to get the blended image



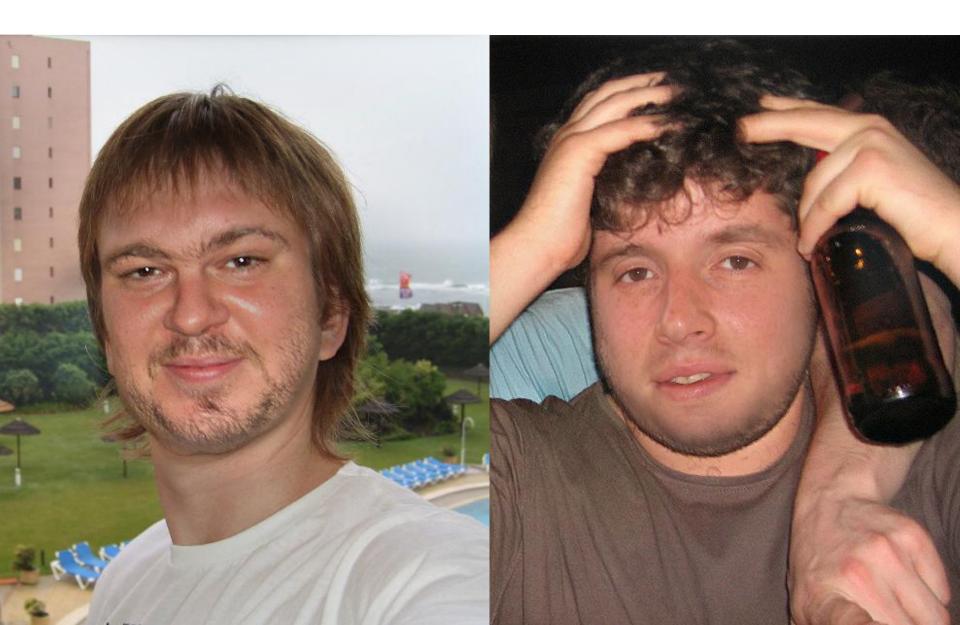


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#### Pyramids (not Laplacian ) & sea (talu)







