

Lecture 10

Multiscale Image Representation

ECEN 5283 Computer Vision

Dr. Guoliang Fan
School of Electrical and Computer Engineering
Oklahoma State University

Goals

- ▶ To introduce the concepts of **multiscale image analysis**
- ▶ To create the **Gaussian pyramid** that supports efficient multiscale image processing
- ▶ To discuss the **Laplacian pyramid** that plays a complementary role to the Gaussian pyramid

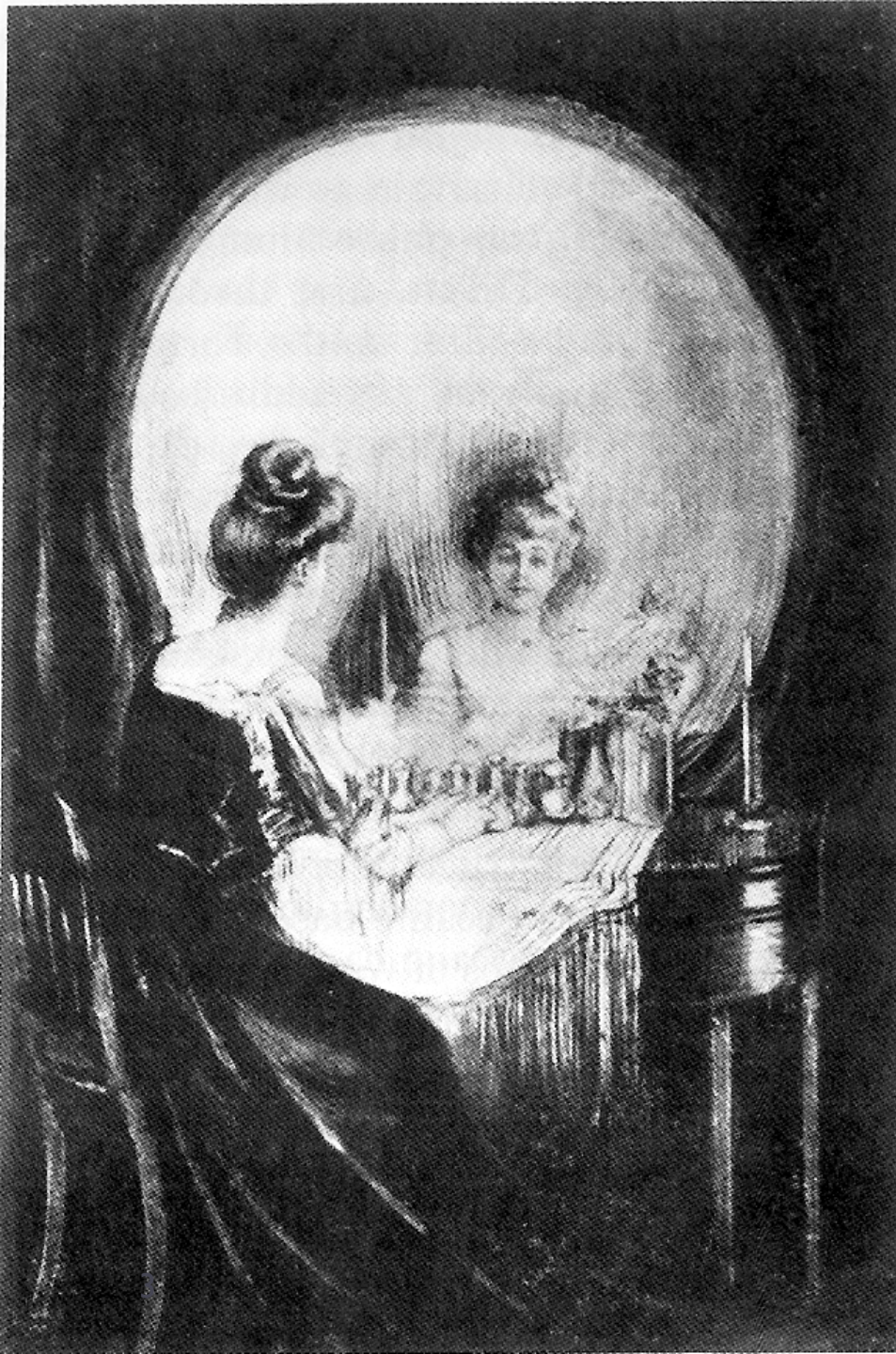


Image Scales

- ▶ Images look quite different at different scales.
 - ▶ **Different scales** are coded in terms of the response of a set of filters that operate at different number of pixels.
 - ▶ **A small scale filter** is used to find details while **a large scale filter** can find the major structures.

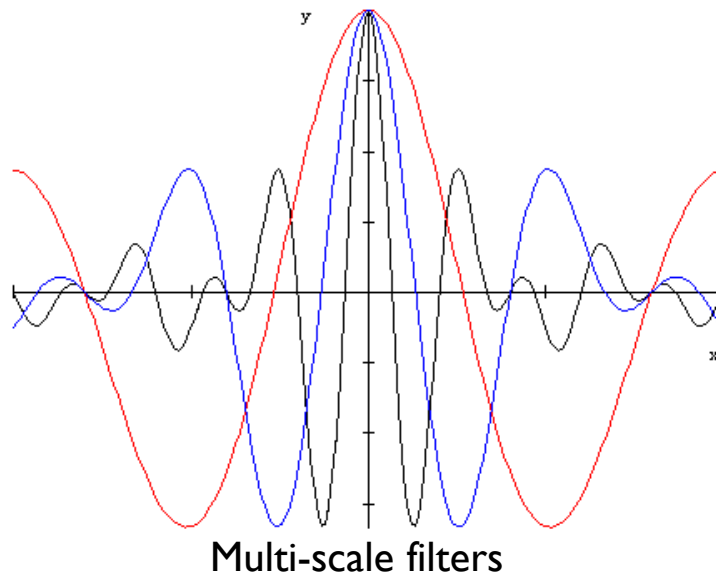
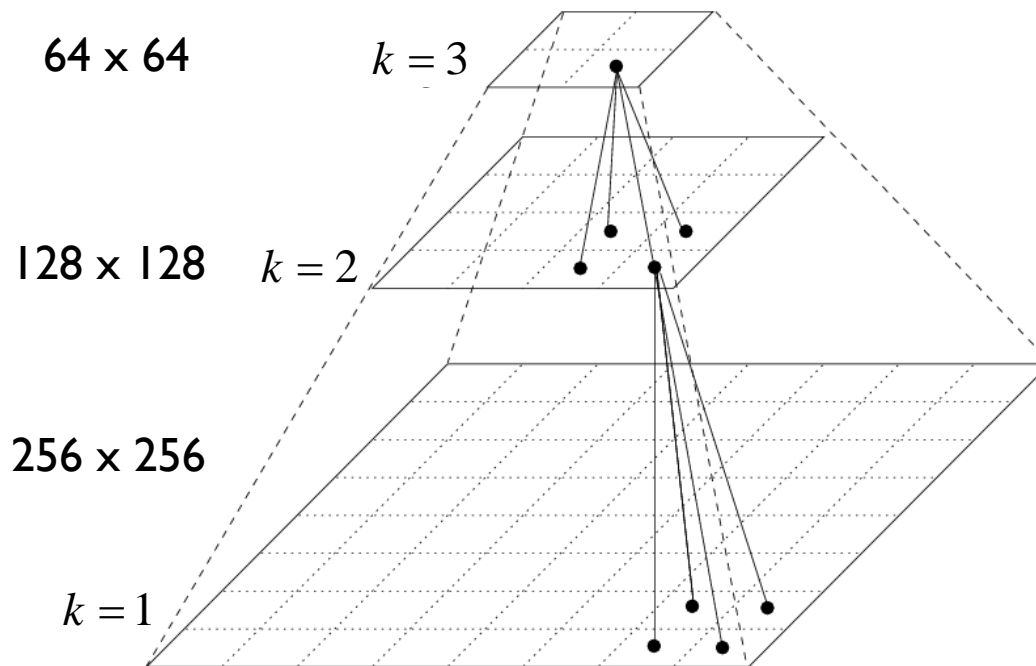


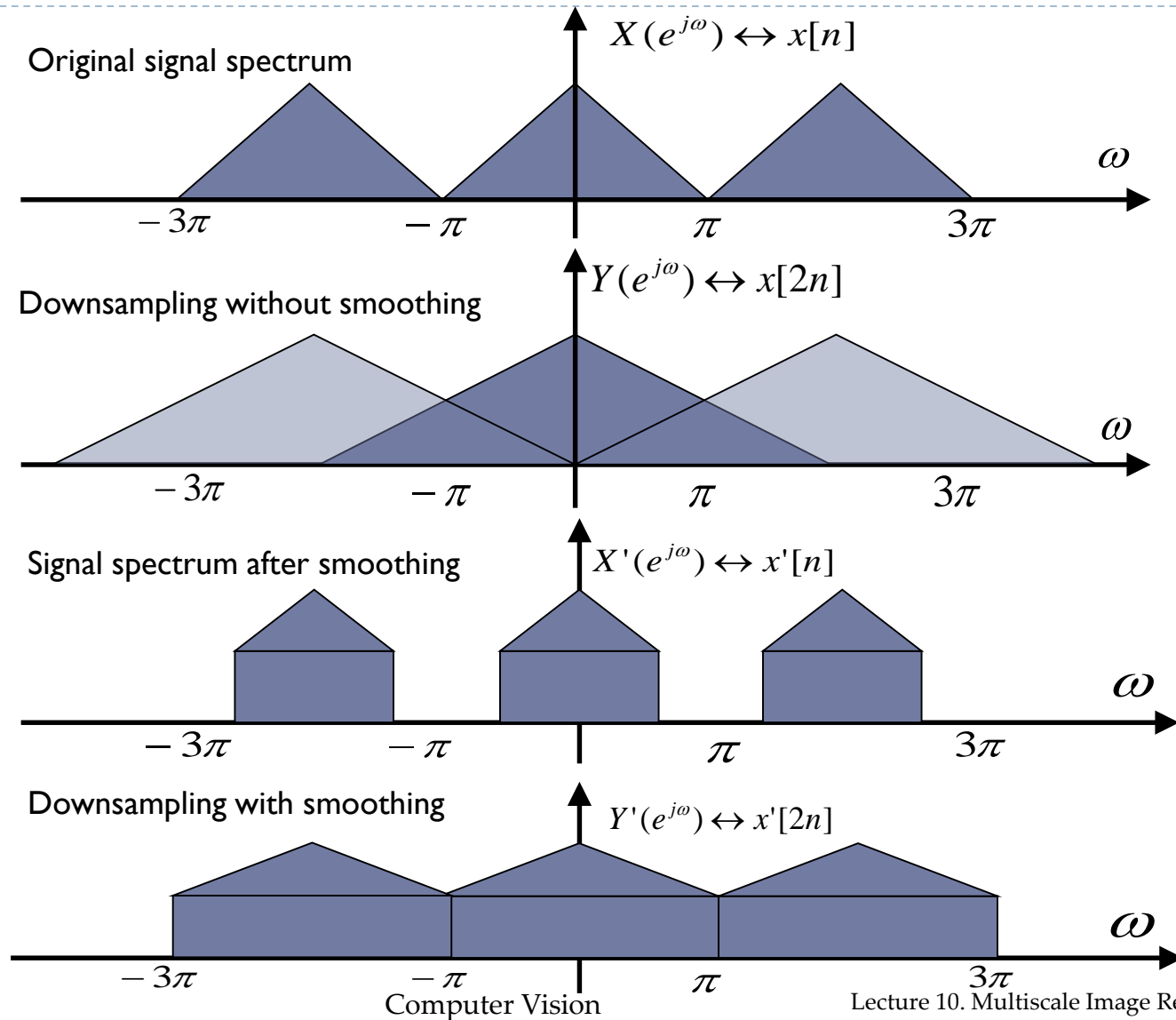
Image Pyramid: Why?

- ▶ Instead of using a large scale filter, we can use a single small scale filter to a smoothed and re-sampled version of an image.

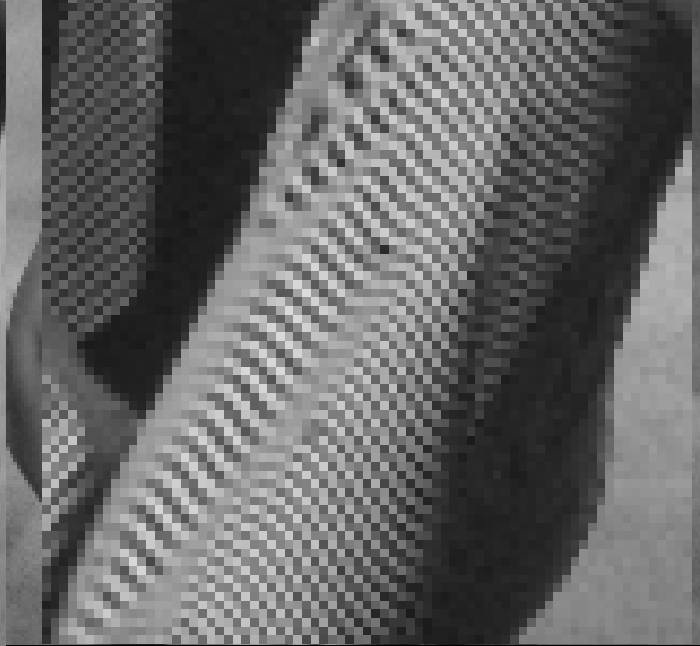


$$I_{k+1} = S \downarrow (G ** I_k) = S \downarrow I_k^G \rightarrow I_{k+1}(i, j) = I_k^G(2i, 2j)$$

Why smoothing before resampling?



Why smoothing before resampling?



Original Image

Subsampled image

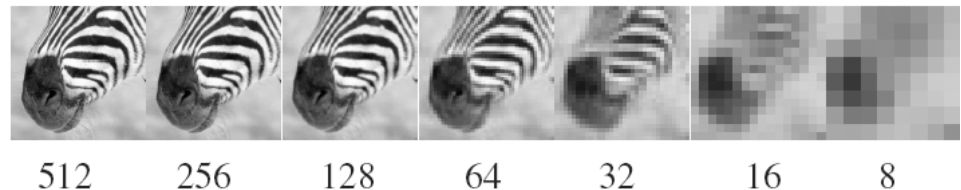
Sampling with anti-aliasing

Gaussian Pyramid

- ▶ **Gaussian pyramid** is a collection of representation of an image at different scales or resolutions.

$$P_{\text{Gaussian}}(I)_1 = I \quad (\text{finest scale})$$

$$P_{\text{Gaussian}}(I)_{n+1} = S \downarrow (G_{\delta} ** P_{\text{Gaussian}}(I)_n)$$



- ▶ Applications
 - ▶ Coarse-to-fine search
 - ▶ Multiscale image segmentation



Gaussian Pyramid Applications: Hierarchical Pattern Matching

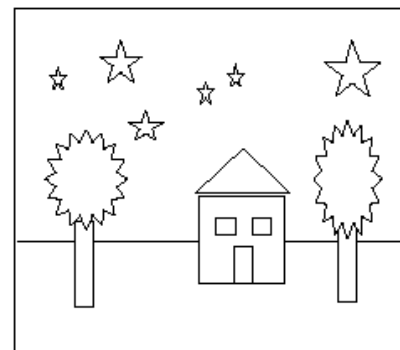
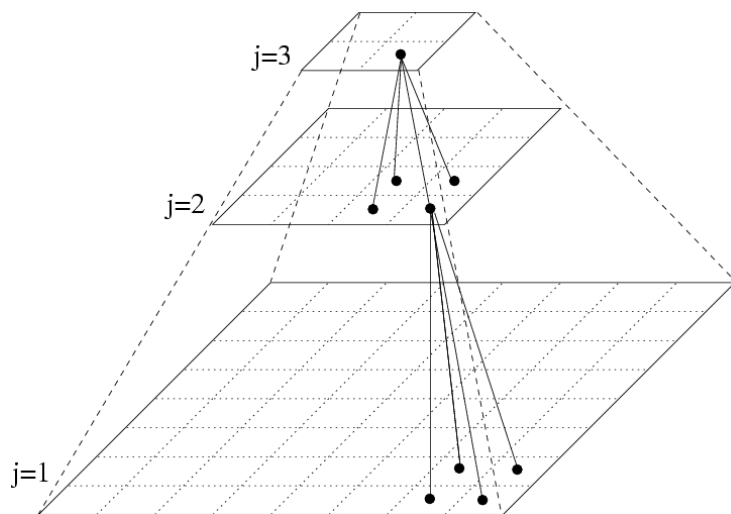


► Memory

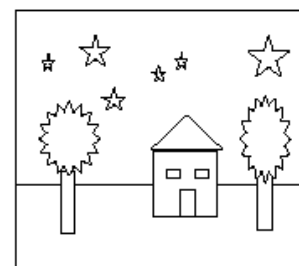
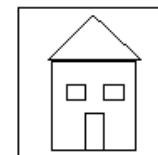
$$M = 2^N \times 2^N (1 + 1/4 + 1/16 + \dots)$$

$$= 2^N \times 2^N \times 4/3$$

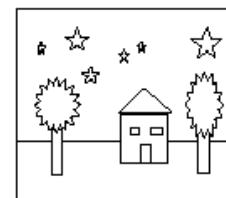
► Coarse-to-fine search



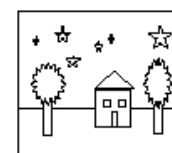
search



search



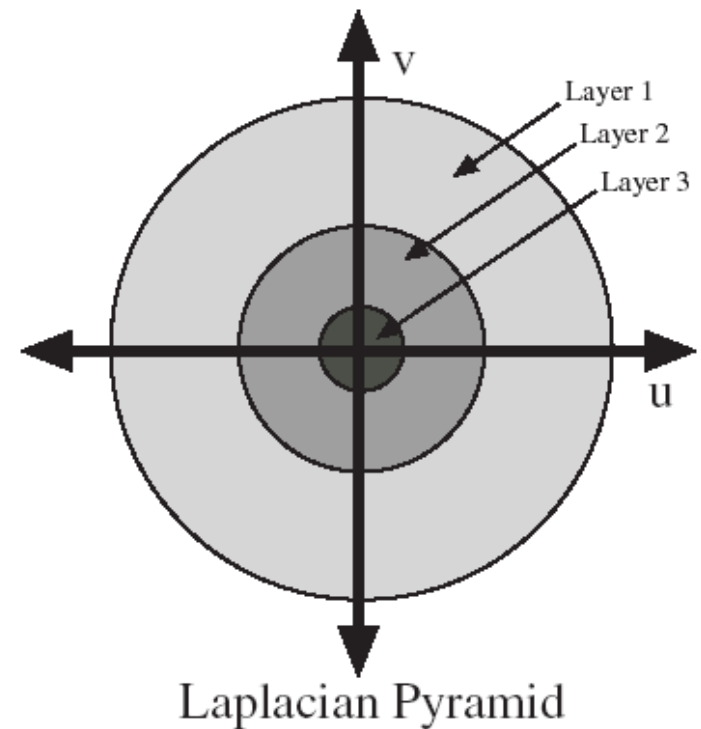
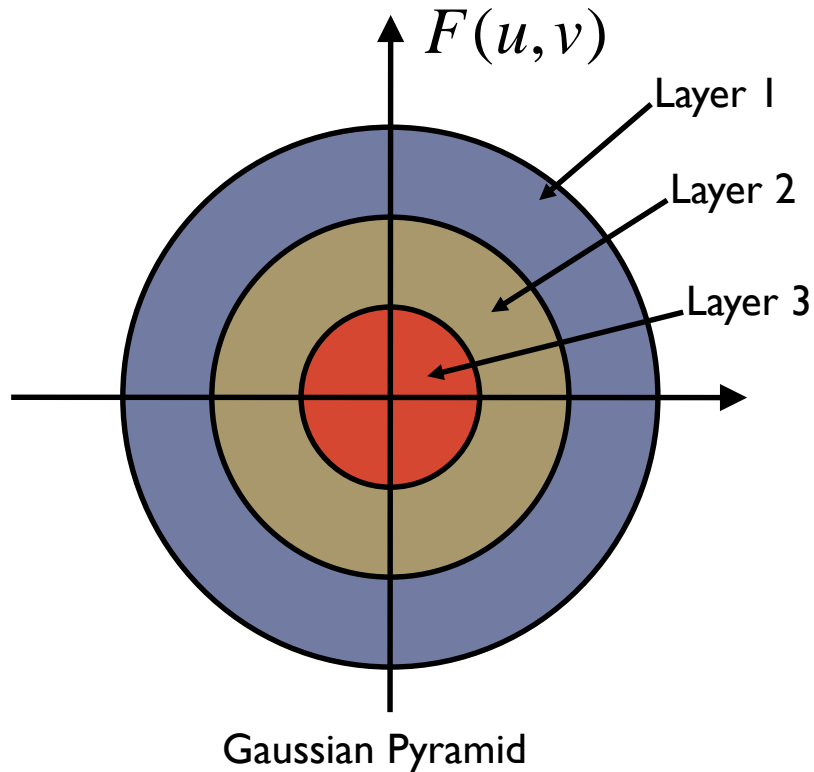
search



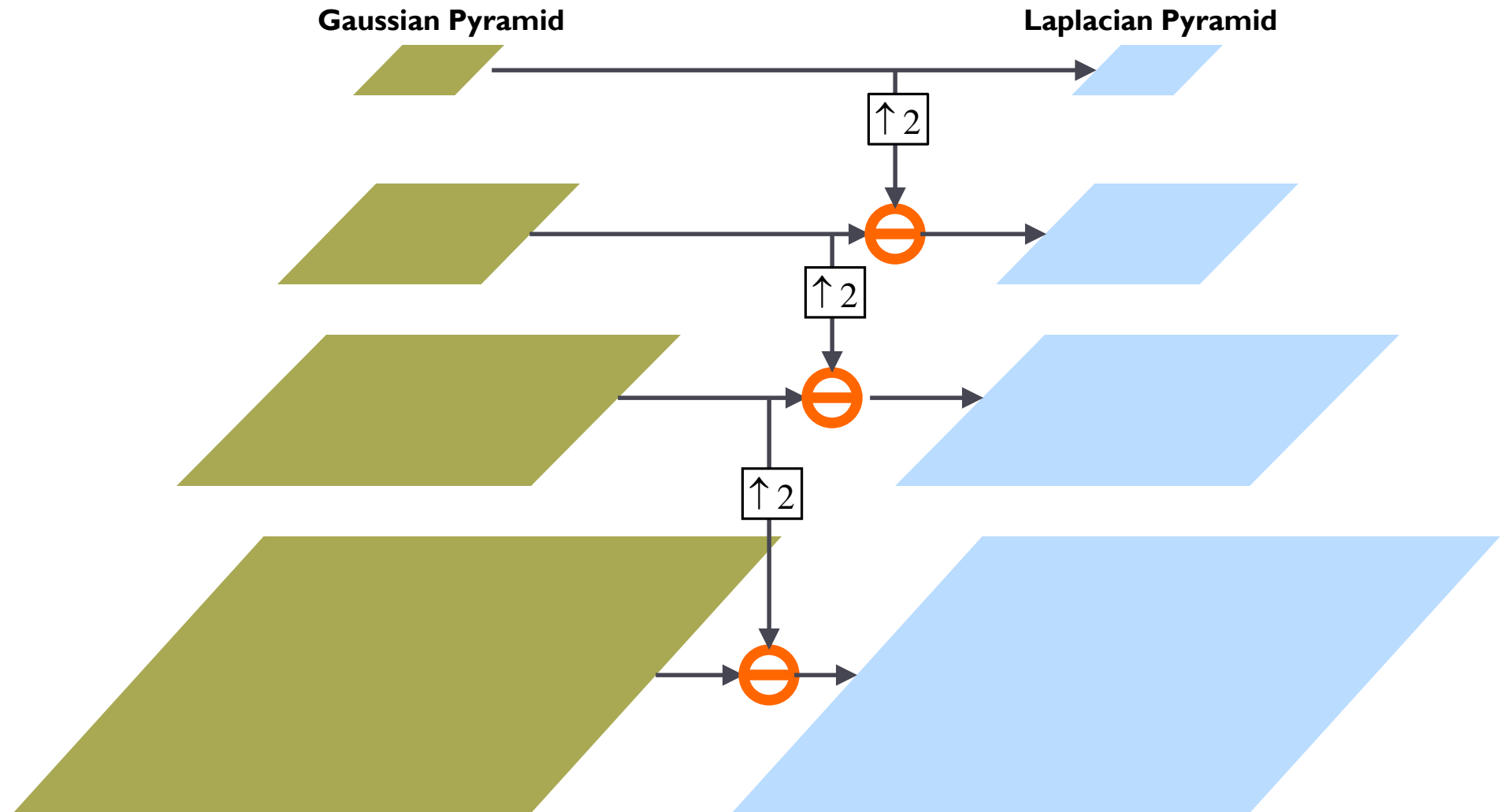
search



Frequency-domain Interpretation

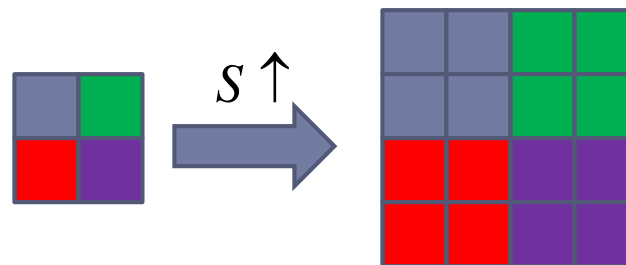


Laplacian and Gaussian Pyramids

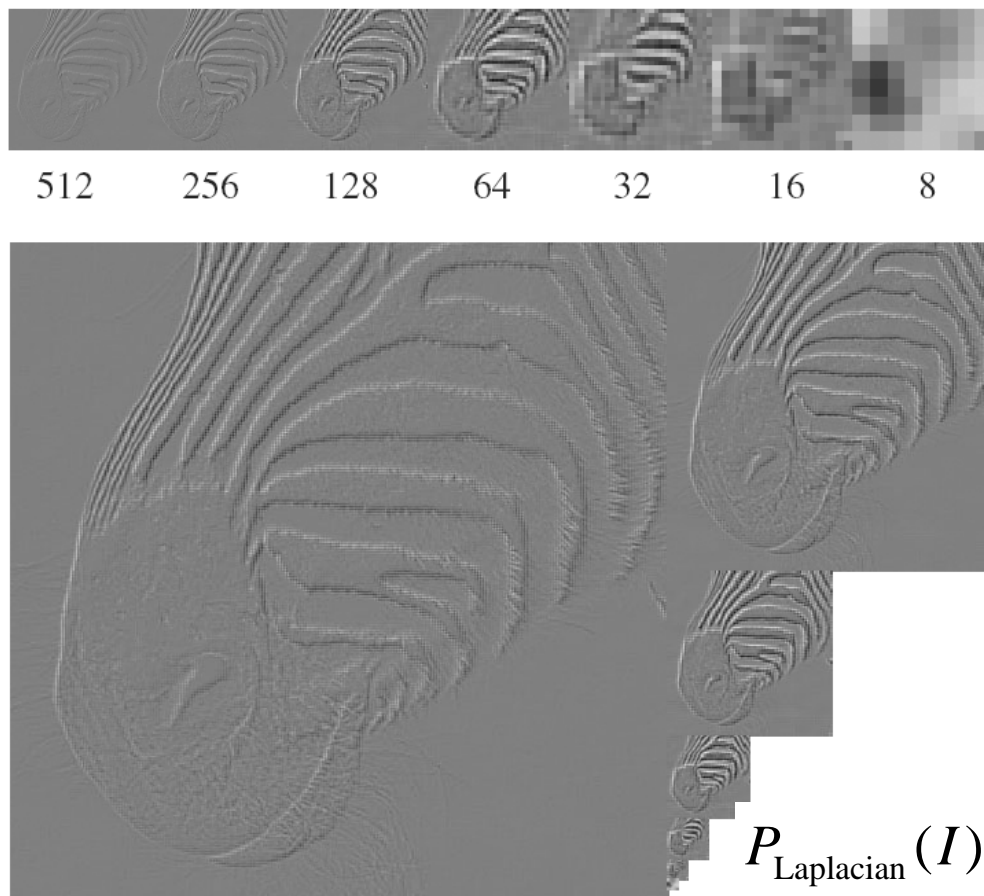


Laplacian Pyramid

- ▶ Laplacian pyramid makes use of the fact that a coarser layer of Gaussian pyramid can predict the next finer scale.
- ▶ We can **expand a coarser scale image** by replicating pixels which involves an up-sampling operator $S \uparrow$.
- ▶ Compared to Gaussian pyramid, Laplacian pyramid is less redundant and has **rich high-frequency information** (why?)
- ▶ We need only **store the difference** between this prediction and the next finer layer itself (high-frequency).



Laplacian Pyramid

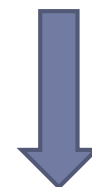


$$P_{\text{Gaussian}}(I)_m \quad (k = 1, \dots, m)$$



$$P_{\text{Laplacian}}(I)_m = P_{\text{Gaussian}}(I)_m$$

(where m is the coarsest scale.)

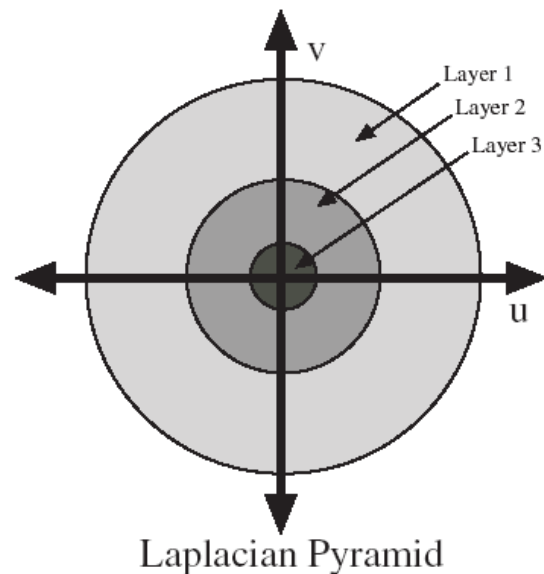


$$P_{\text{Laplacian}}(I)_k = P_{\text{Gaussian}}(I)_k - S \uparrow (P_{\text{Gaussian}}(I)_{k+1})$$

Figure 11.7. A Laplacian pyramid of images, running from 512x512 to 8x8. A zero response is coded with a mid-grey; positive values are lighter and negative values are darker. Notice that the stripes give stronger responses at particular scales, because each layer corresponds (roughly) to the output of a band-pass filter.

Laplacian Pyramid Application

- ▶ Image compression
- ▶ Texture analysis
- ▶ Edge detection



Gaussian pyramid

Laplacian pyramid