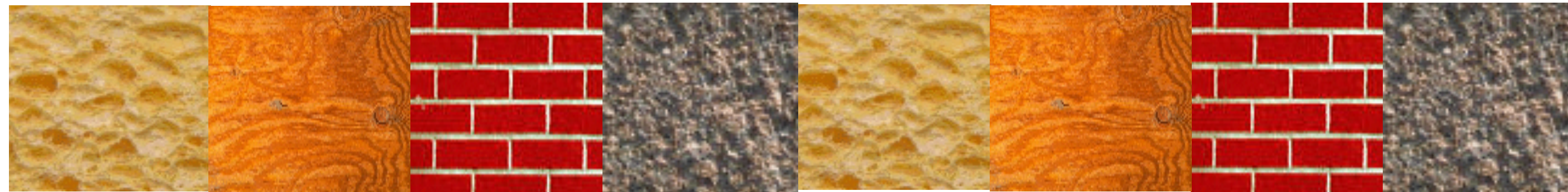




# CPSC 425: Computer Vision



## Lecture 15: Texture, Intro to Color

( unless otherwise stated slides are taken or adopted from **Bob Woodham, Jim Little** and **Fred Tung** )

# Menu for Today (October 10, 2018)

## Topics:

- Texture Analysis
- Laplacian and Oriented Pyramids
- iClicker Quiz
- Introduction to Color

## Readings:

- **Today's** Lecture: Forsyth & Ponce (2nd ed.) 3.1-3.3
- **Next** Lecture: N/A

## Reminders:

- **Assignment 2:** Face Detection in a Scaled Representation is **due today**
- **Assignment 3:** Texture Synthesis will be **out today**
- **Practice questions** are available now, Additional Office hours Fri, Mon

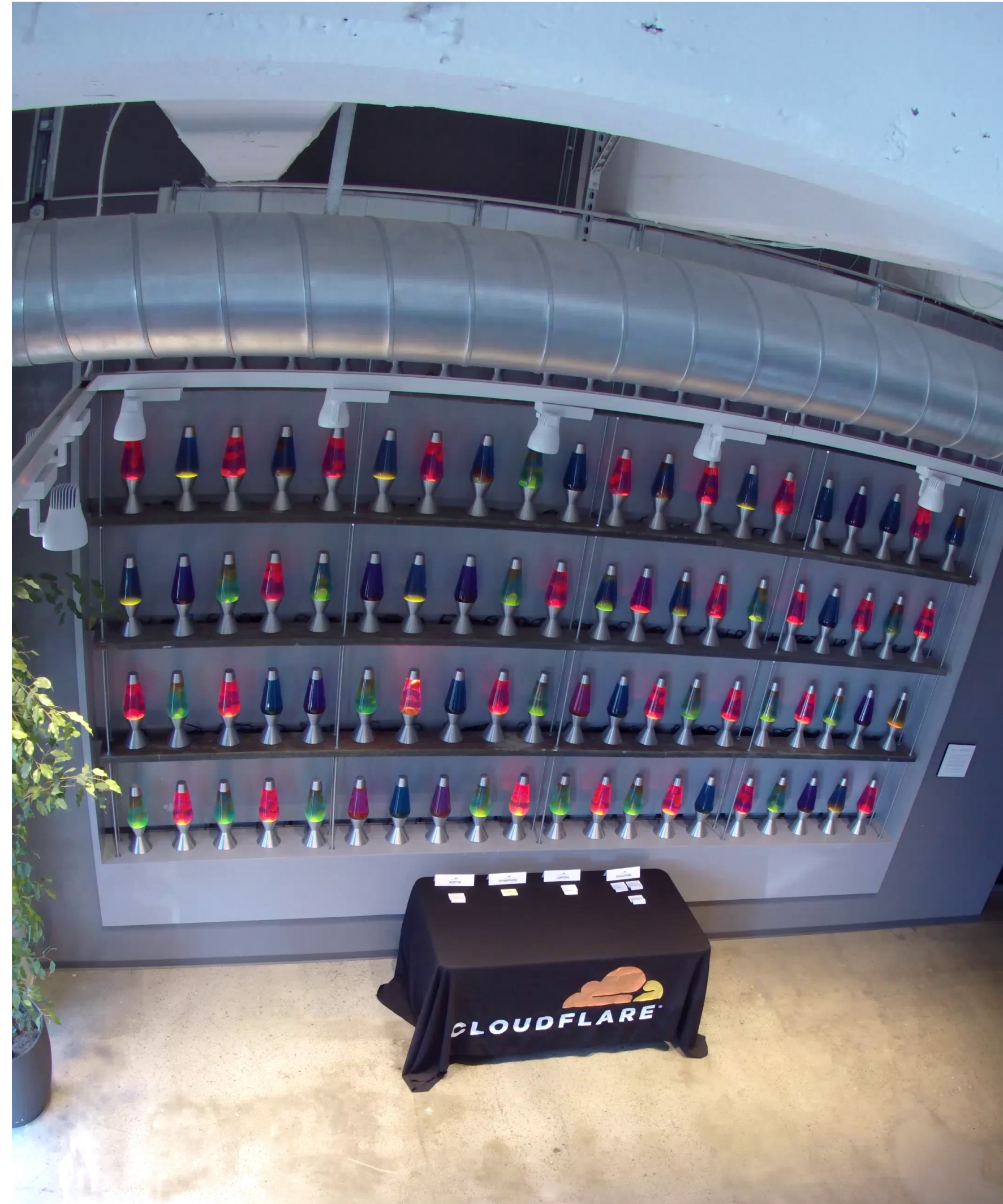
# Today's "fun" Example: NCIS



# Today's "fun" Example: LavaRAND



# Today's "fun" Example: LavaRAND at Cloudflare



# Lecture 14: Re-cap

**Texture** representation is hard

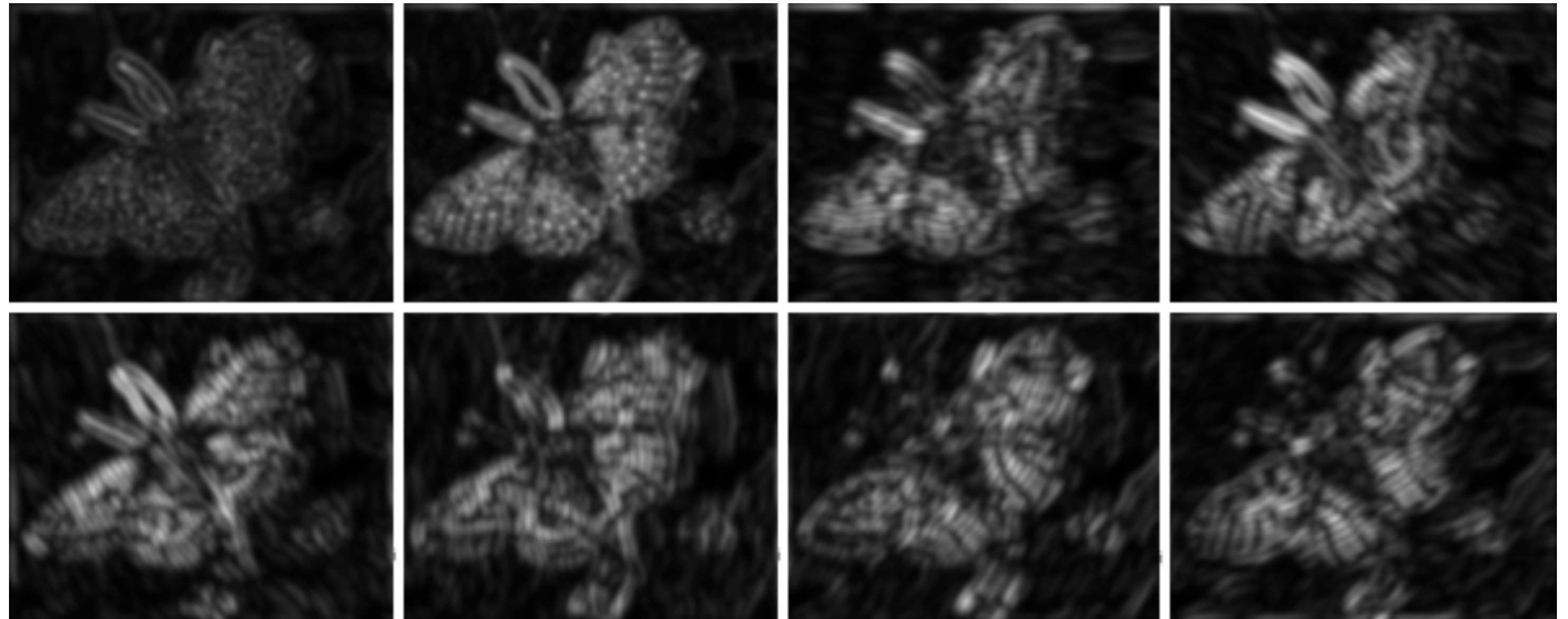
- difficult to define, to analyze
- texture synthesis appears more tractable

Objective of texture **synthesis** is to generate new examples of a texture

- Efros and Leung: Draw samples directly from the texture to generate one pixel at a time. A “data-driven” approach.

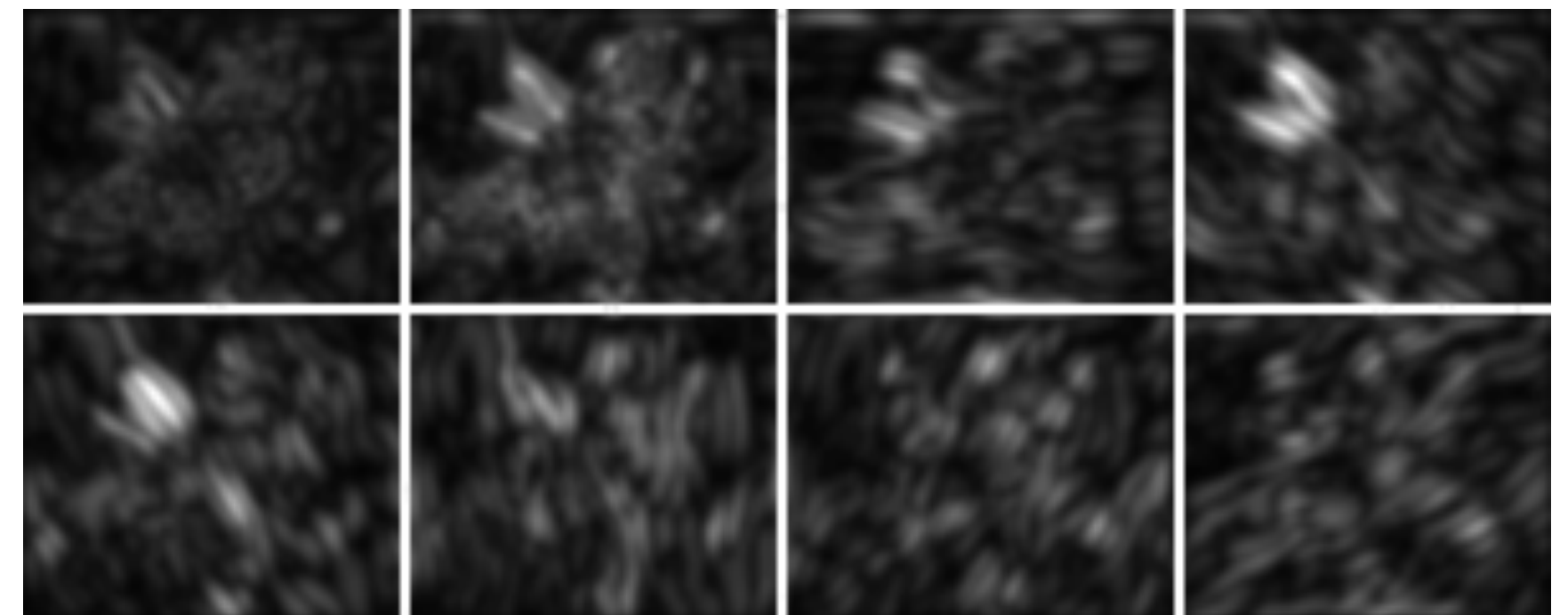
Approaches to texture embed assumptions related to human perception

# Spots and Bars (Fine Scale)



Forsyth & Ponce (1st ed.) Figures 9.3–9.4

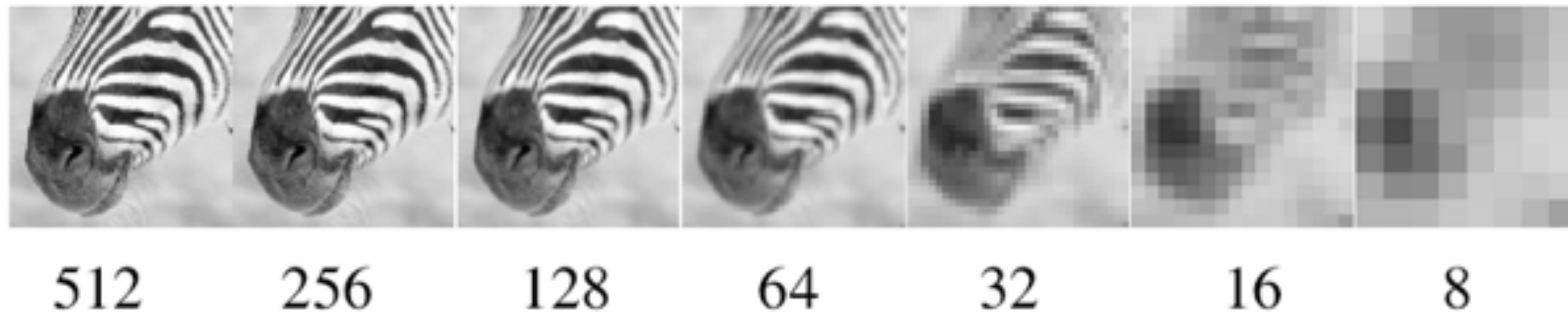
# Spots and Bars (Coarse Scale)



Forsyth & Ponce (1st ed.) Figures 9.3 and 9.5



# Gaussian Pyramid



What happens to the details?

- They get smoothed out as we move to higher levels

What is preserved at the higher levels?

- Mostly large uniform regions in the original image

How would you reconstruct the original image from the image at the upper level?

- That's not possible

Forsyth & Ponce (2nd ed.) Figure 4.17

# Laplacian Pyramid

Building a **Laplacian** pyramid:

- Create a Gaussian pyramid
- Take the difference between one Gaussian pyramid level and the next (before subsampling)

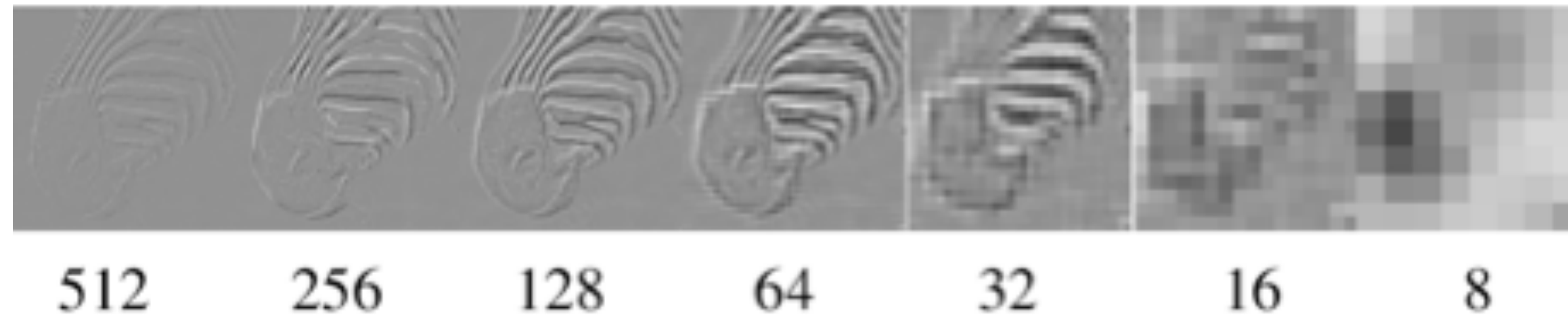
## Properties

- Also known as the difference-of-Gaussian (DOG) function, a close approximation to the Laplacian
- It is a band pass filter – each level represents a different band of spatial frequencies

**Reconstructing** the original image:

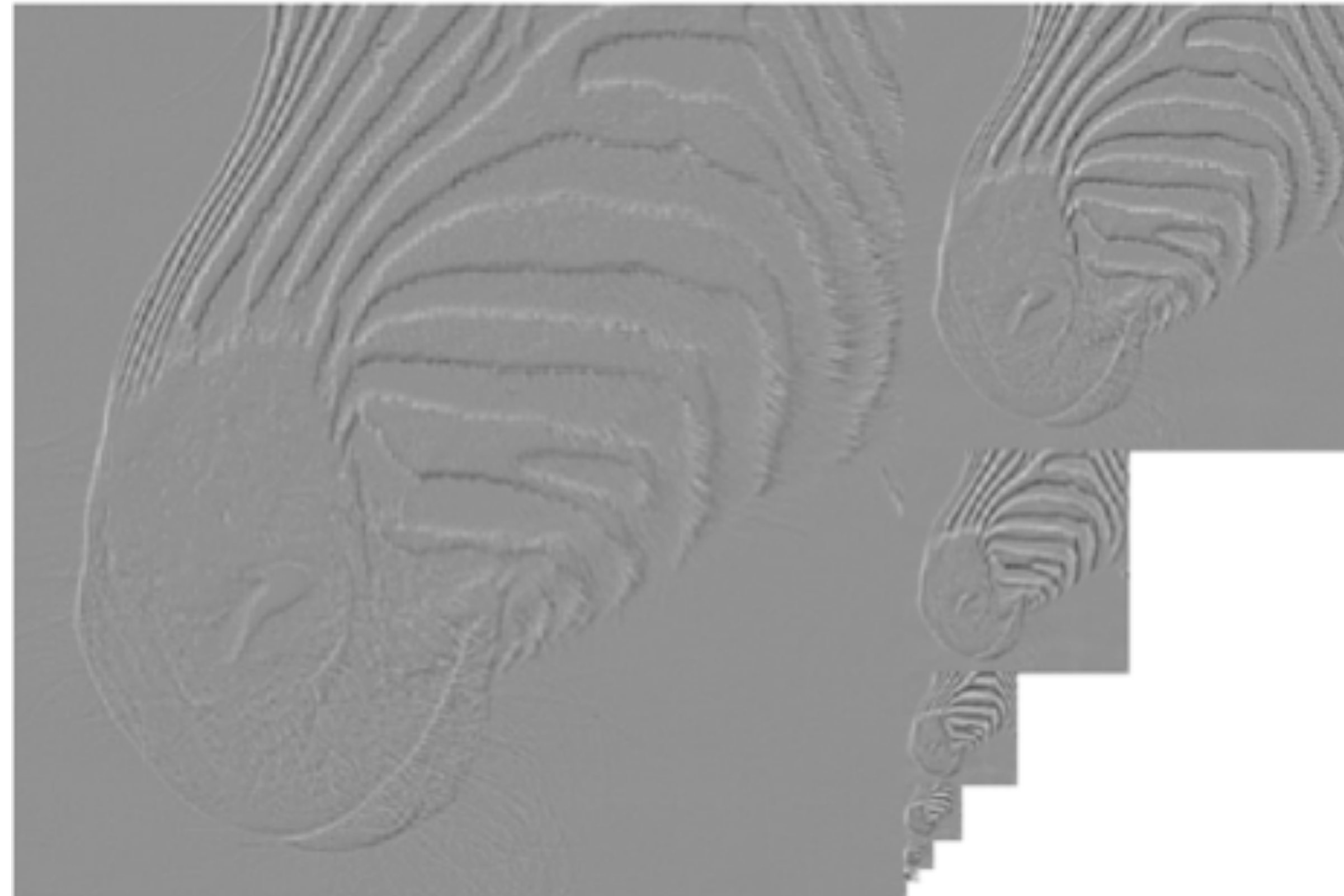
- Reconstruct the Gaussian pyramid starting at top

# Laplacian Pyramid

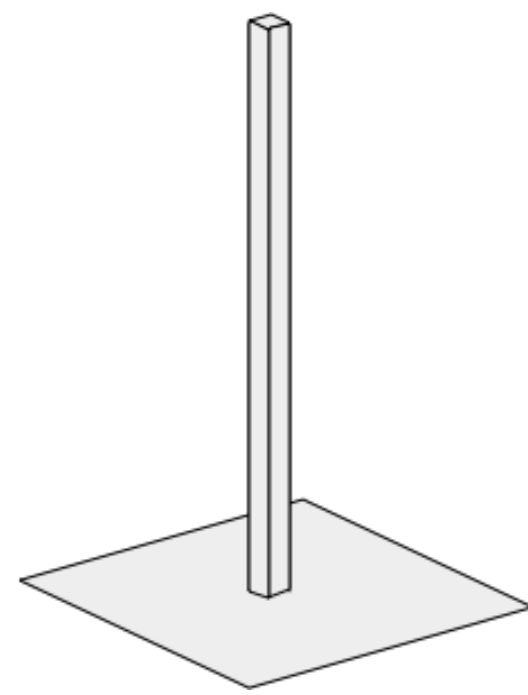


At each level, retain the residuals instead of the blurred images themselves.

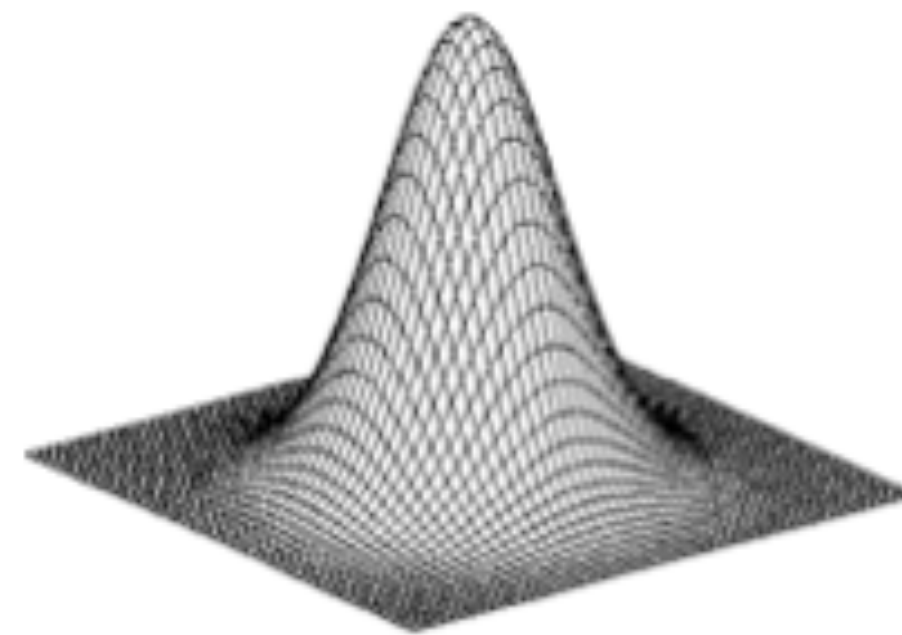
Why is it called Laplacian Pyramid?



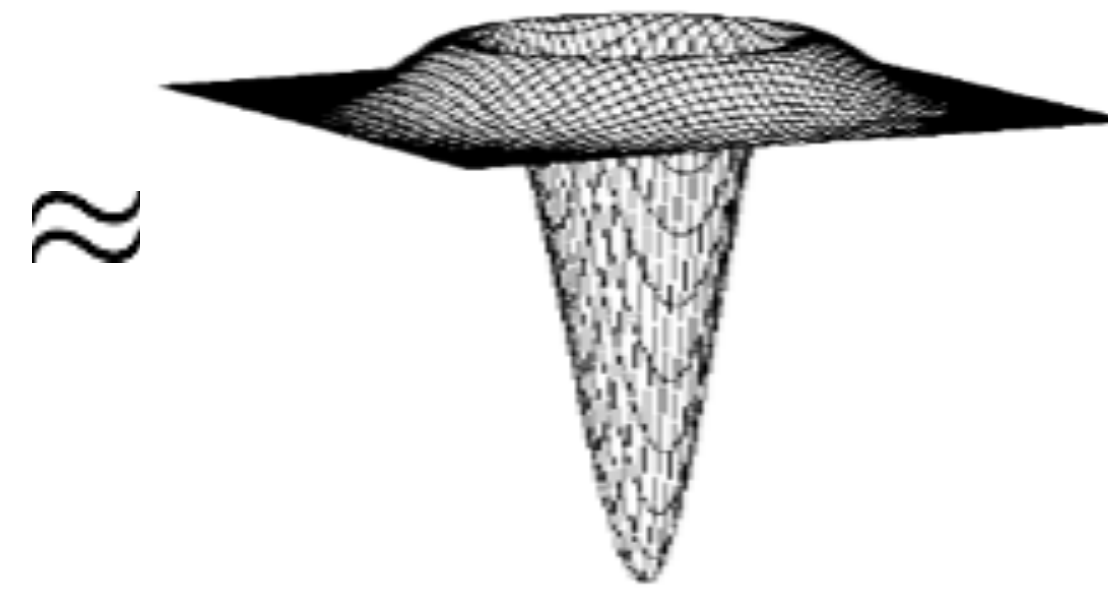
# Why **Laplacian** Pyramid?



unit

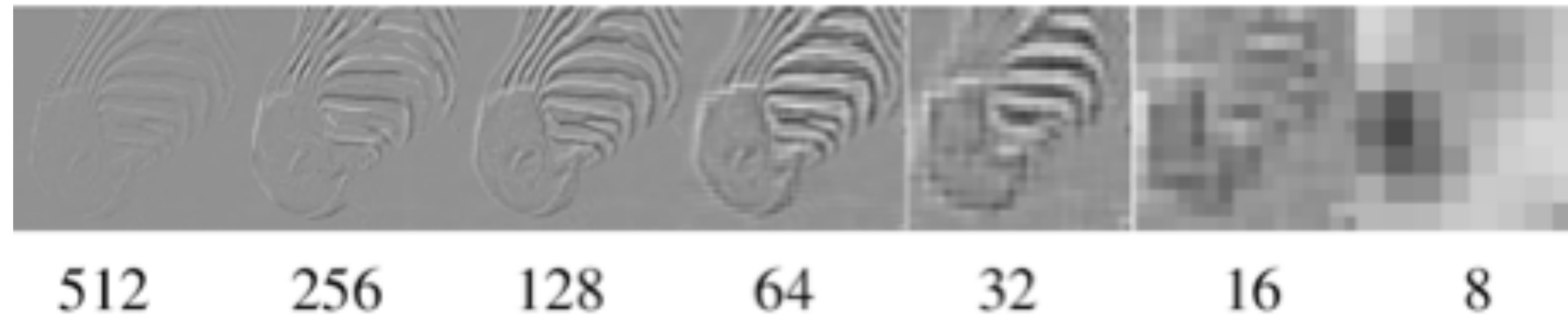


Gaussian



Laplacian

# Laplacian Pyramid

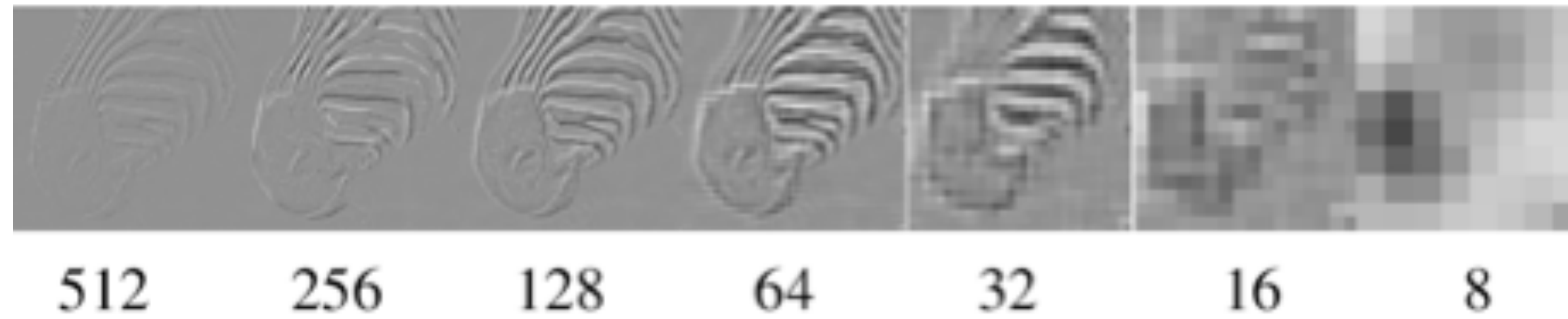


At each level, retain the residuals instead of the blurred images themselves.

**Why is it called Laplacian Pyramid?**

Can we reconstruct the original image using the pyramid?  
— Yes we can!

# Laplacian Pyramid



At each level, retain the residuals instead of the blurred images themselves.

**Why is it called Laplacian Pyramid?**

Can we reconstruct the original image using the pyramid?

— Yes we can!

What do we need to store to be able to reconstruct the original image?

# Let's start by just looking at **one level**



level 0

=



level 1 (upsampled)

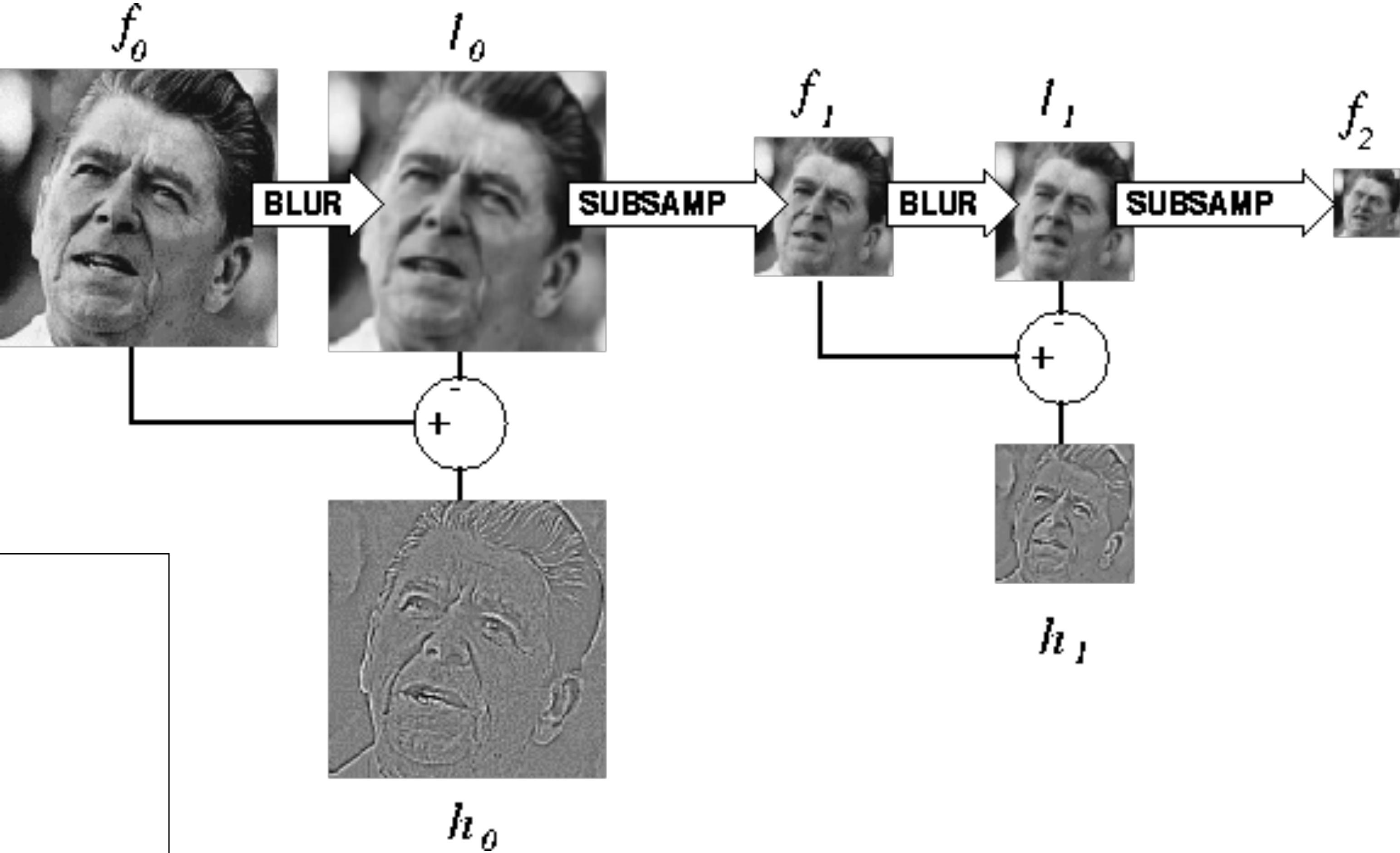
+



residual

Does this mean we need to store both residuals and the blurred copies of the original?

# Constructing a **Laplacian** Pyramid



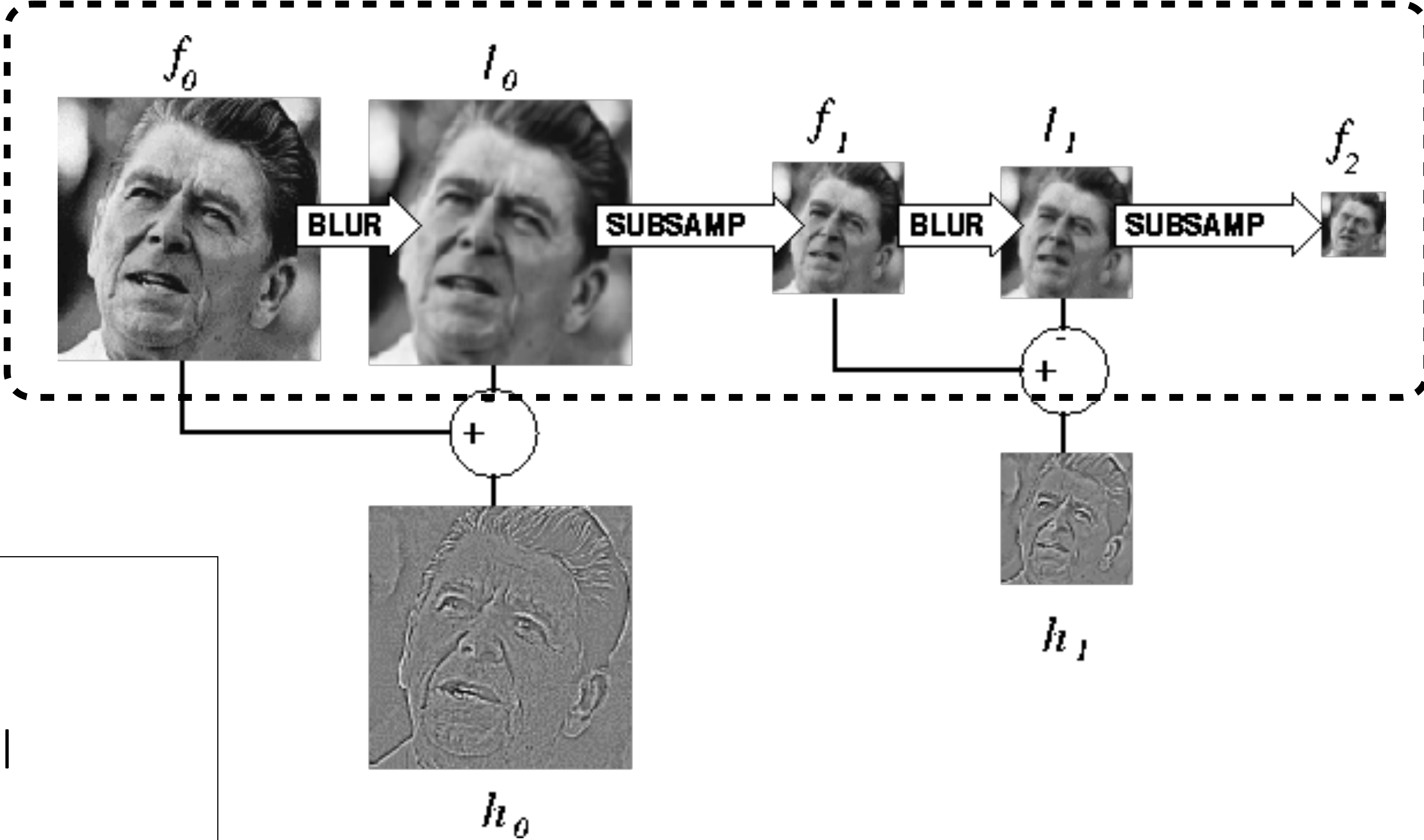
## Algorithm

repeat:  
  filter  
  compute residual  
  subsample  
until min resolution reached



# Constructing a **Laplacian** Pyramid

What is this part?

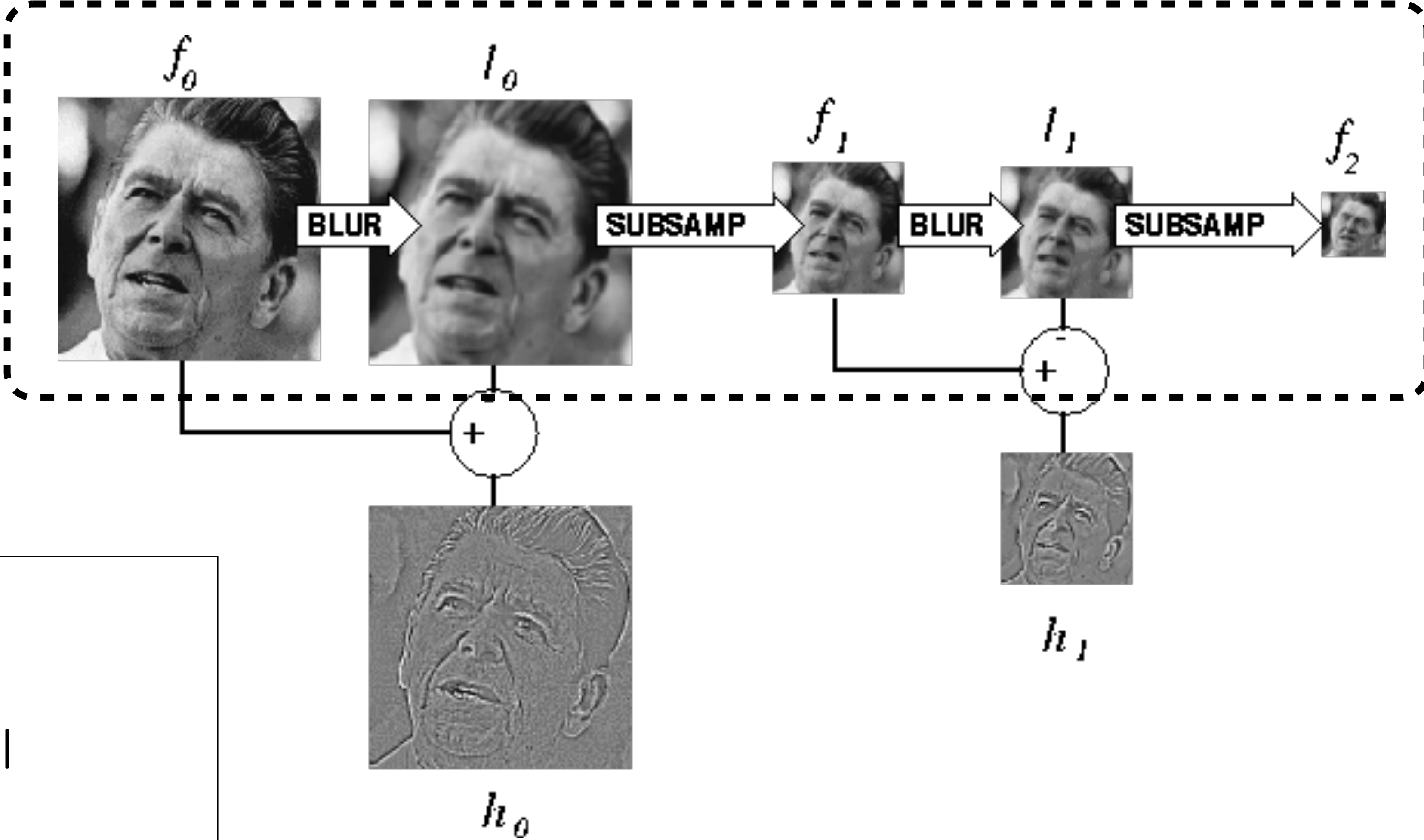


## Algorithm

repeat:  
  filter  
  compute residual  
  subsample  
until min resolution reached

# Constructing a **Laplacian** Pyramid

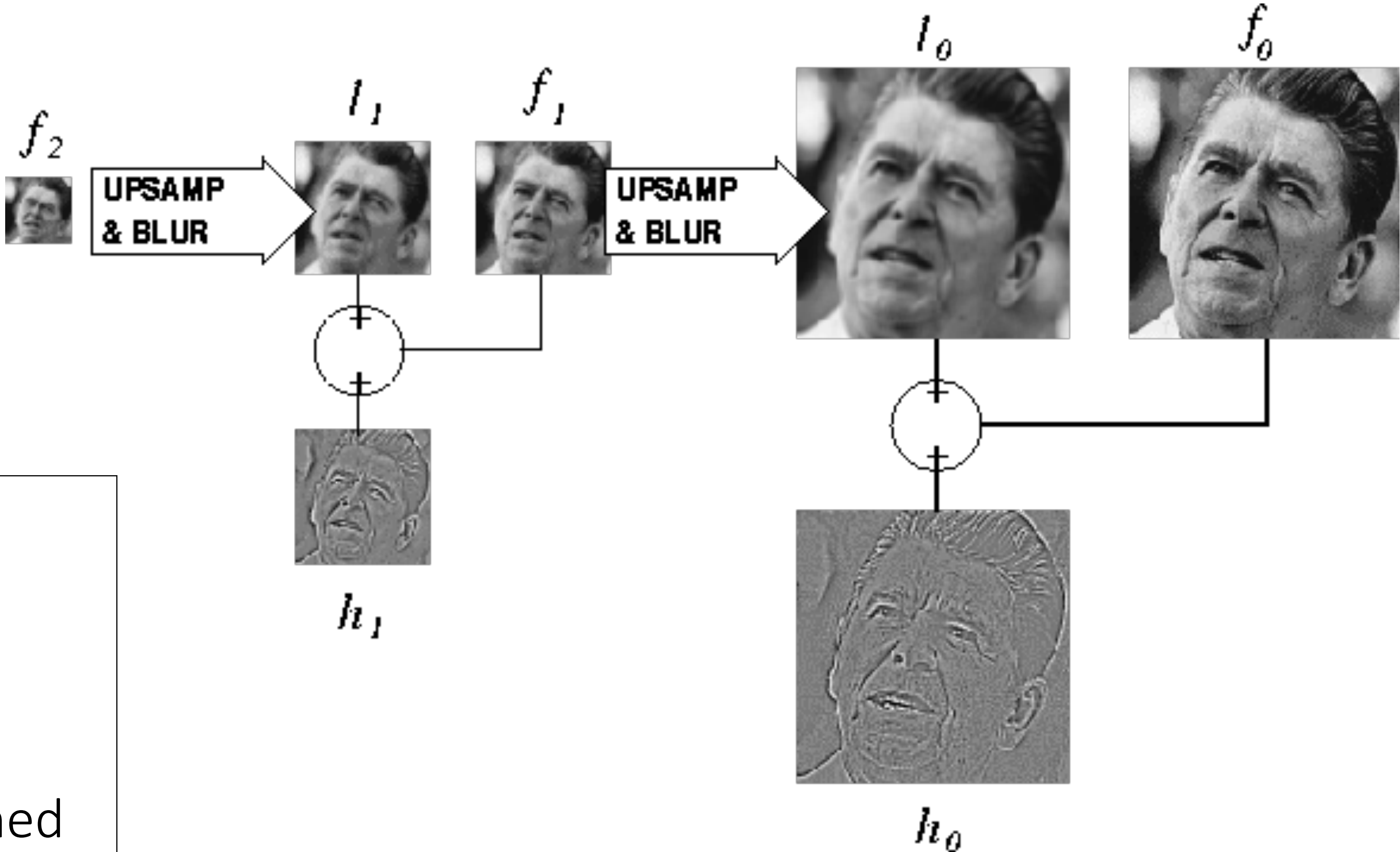
It's a Gaussian Pyramid



## Algorithm

repeat:  
  filter  
  compute residual  
  subsample  
until min resolution reached

# Reconstructing the Original Image



## Algorithm

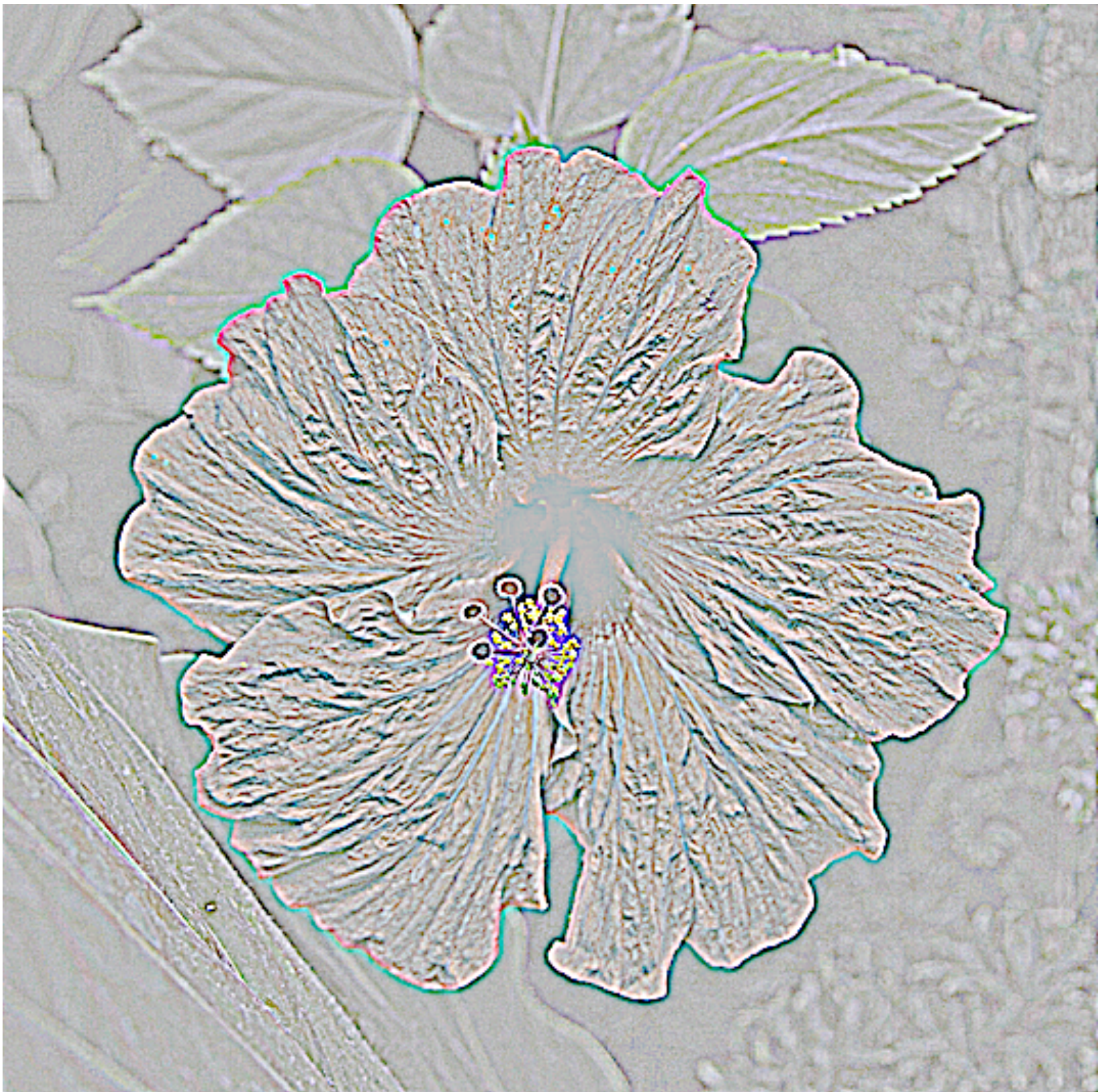
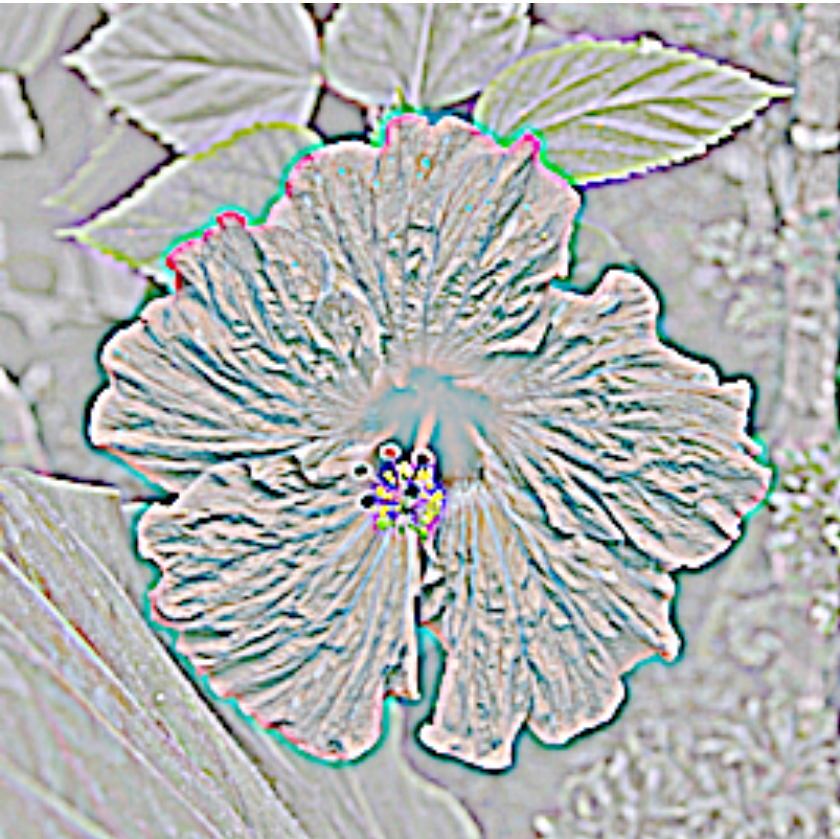
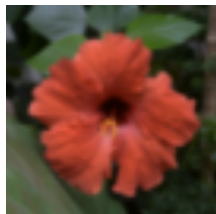
repeat:  
    upsample  
    sum with residual  
until orig resolution reached

# Gaussian vs Laplacian Pyramid

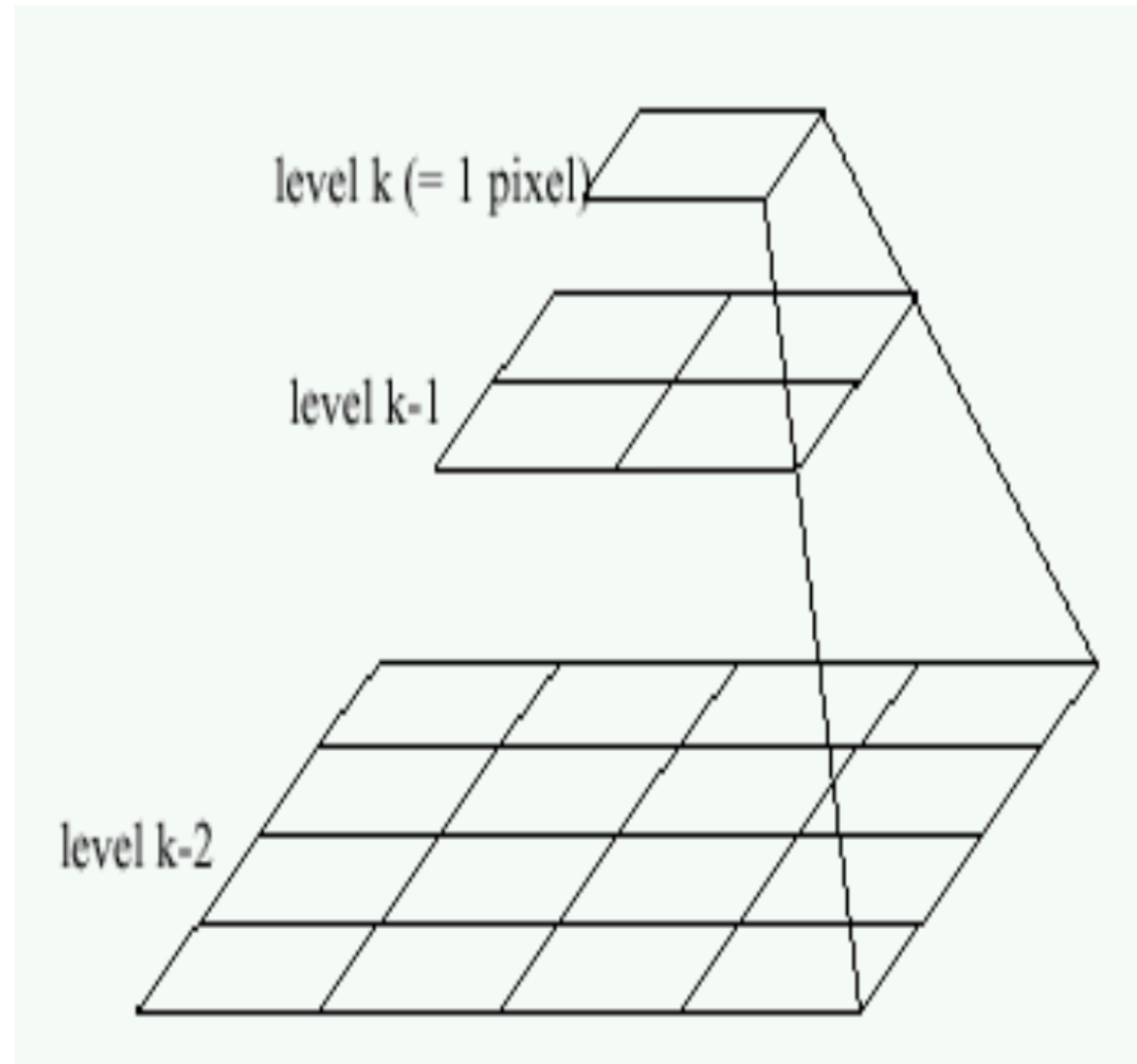


Shown in opposite order for space

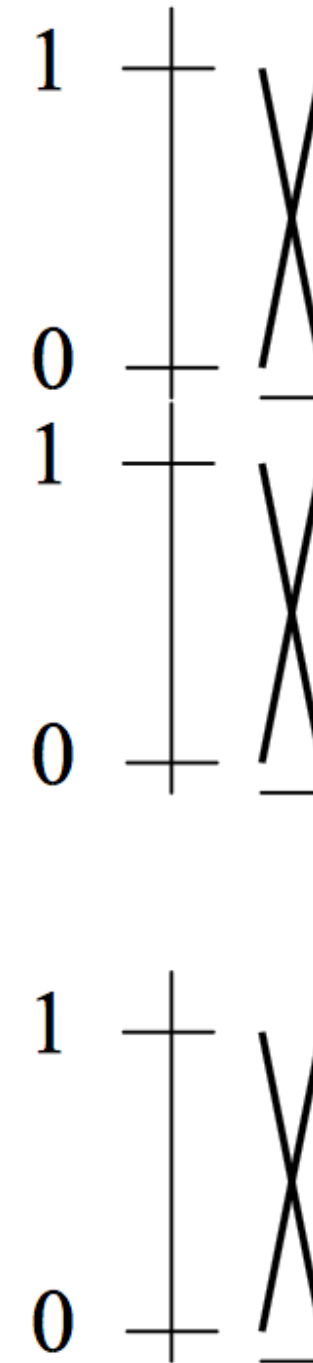
Which one takes more space to store?



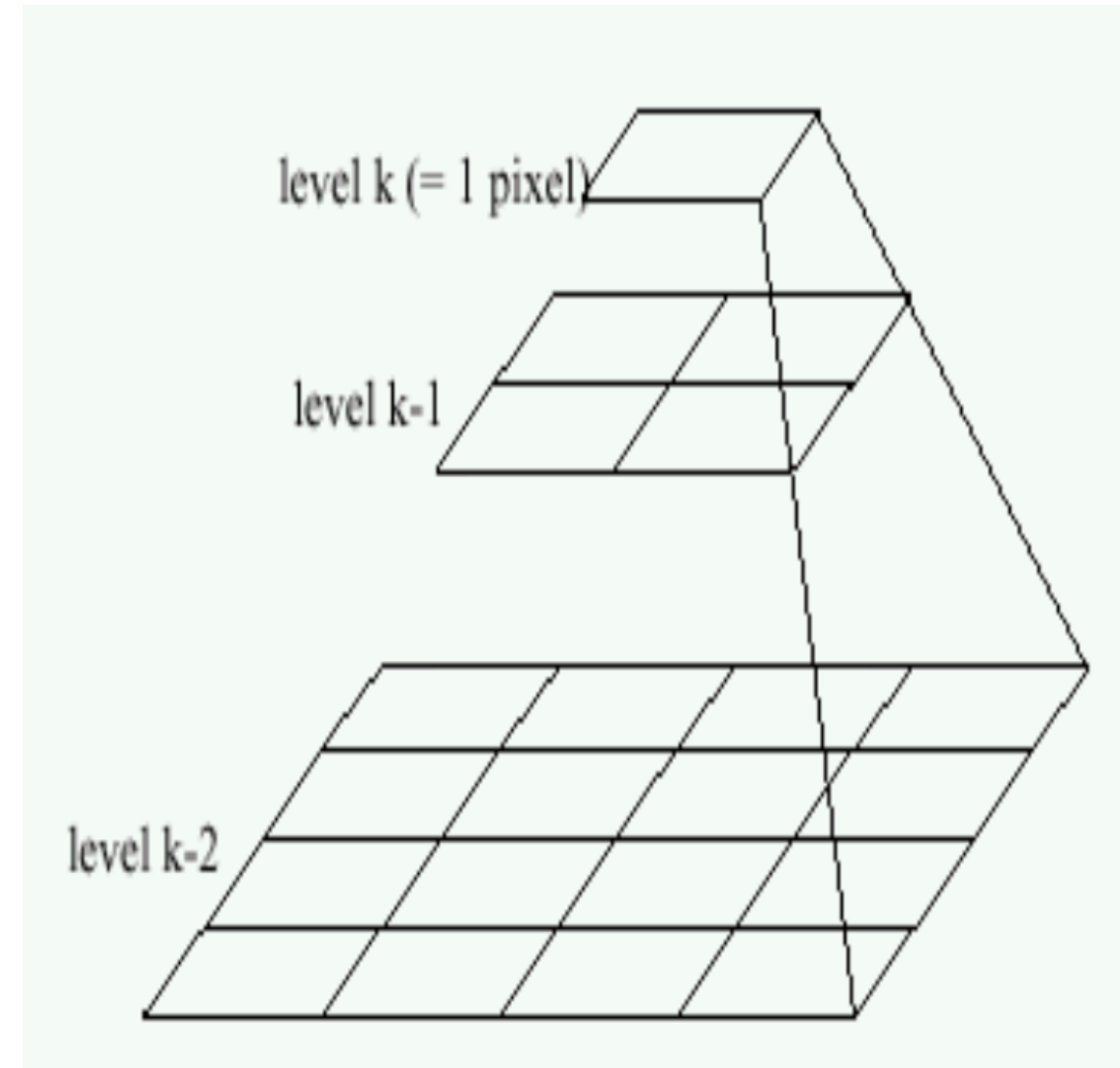
# Aside: Image Blending



Left pyramid



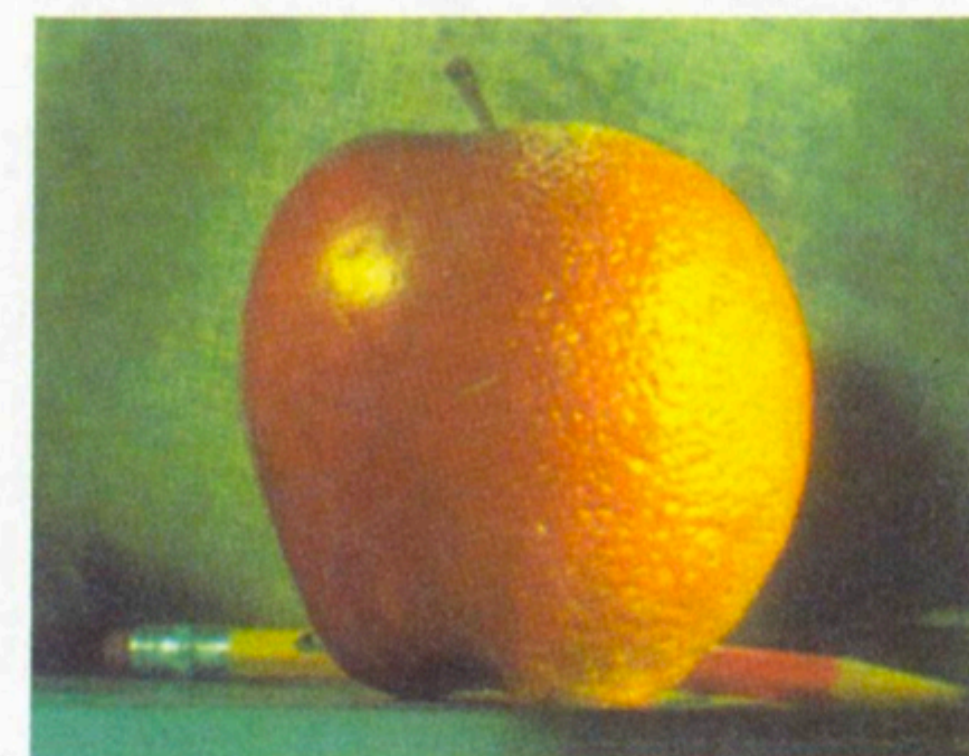
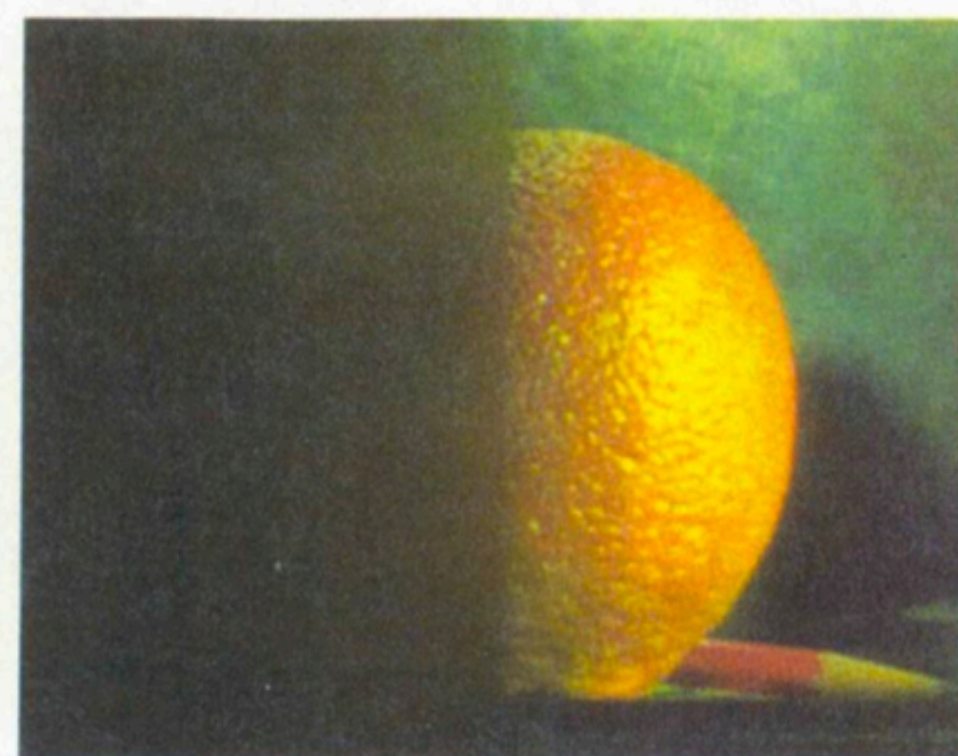
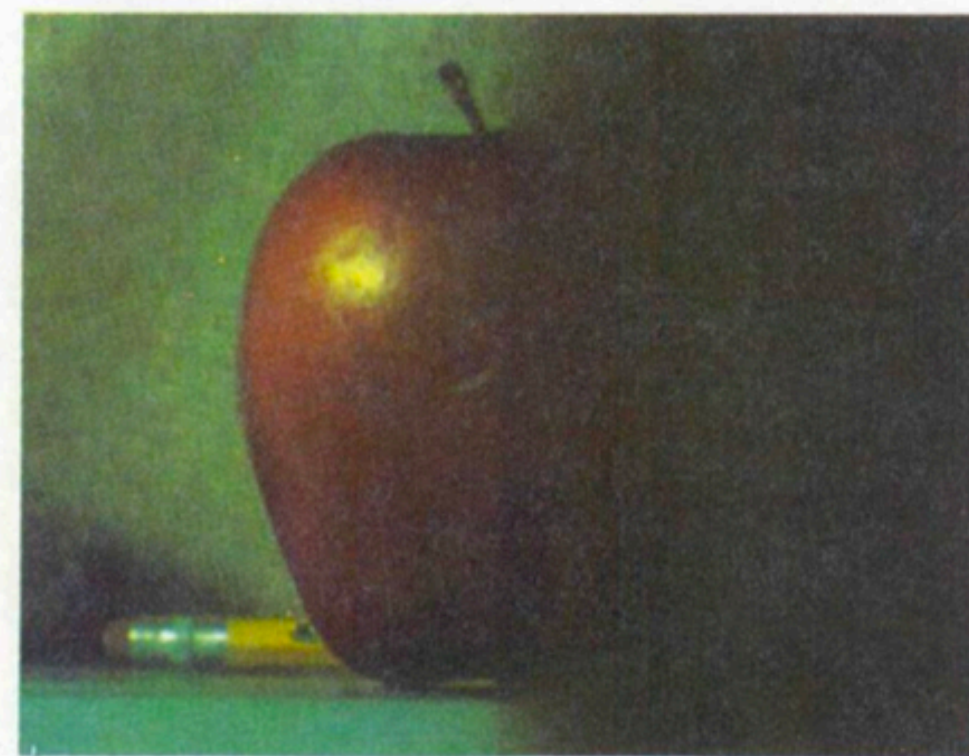
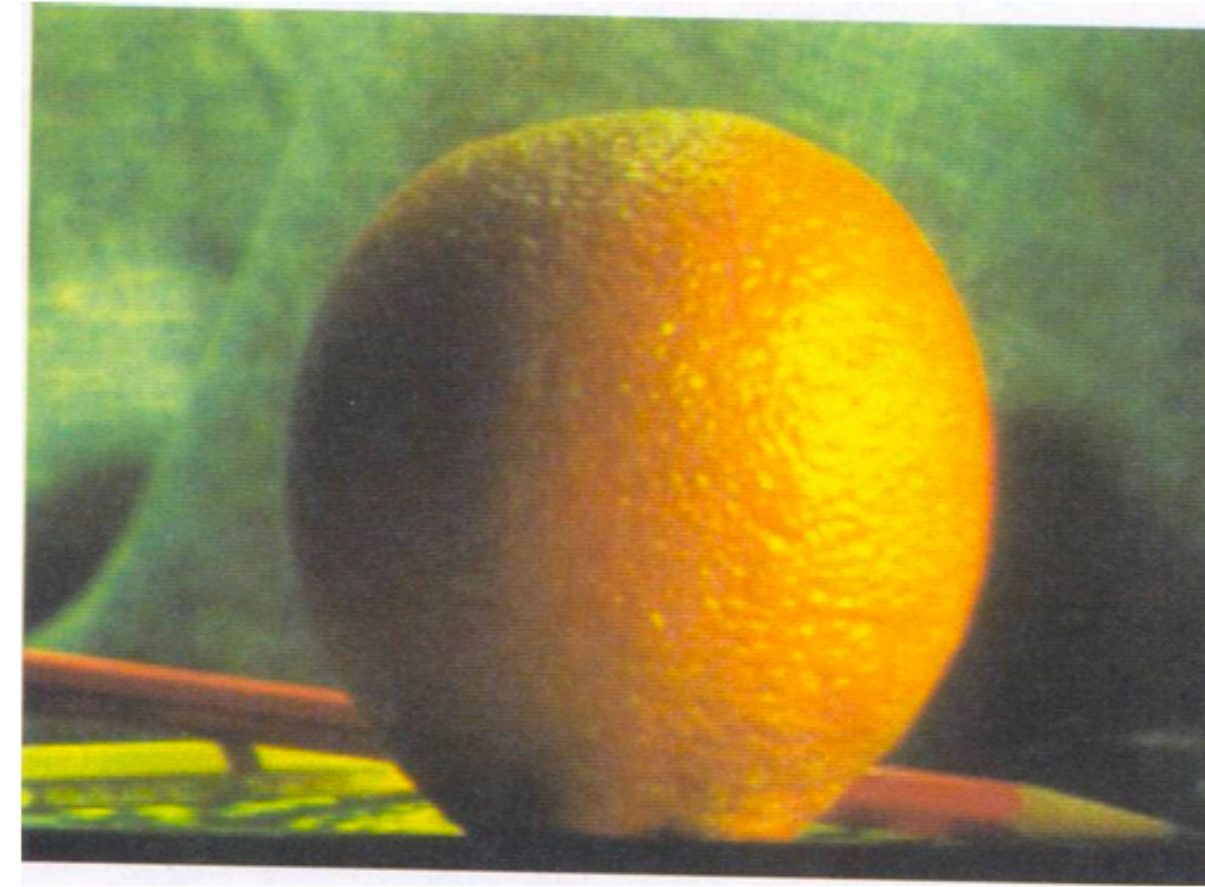
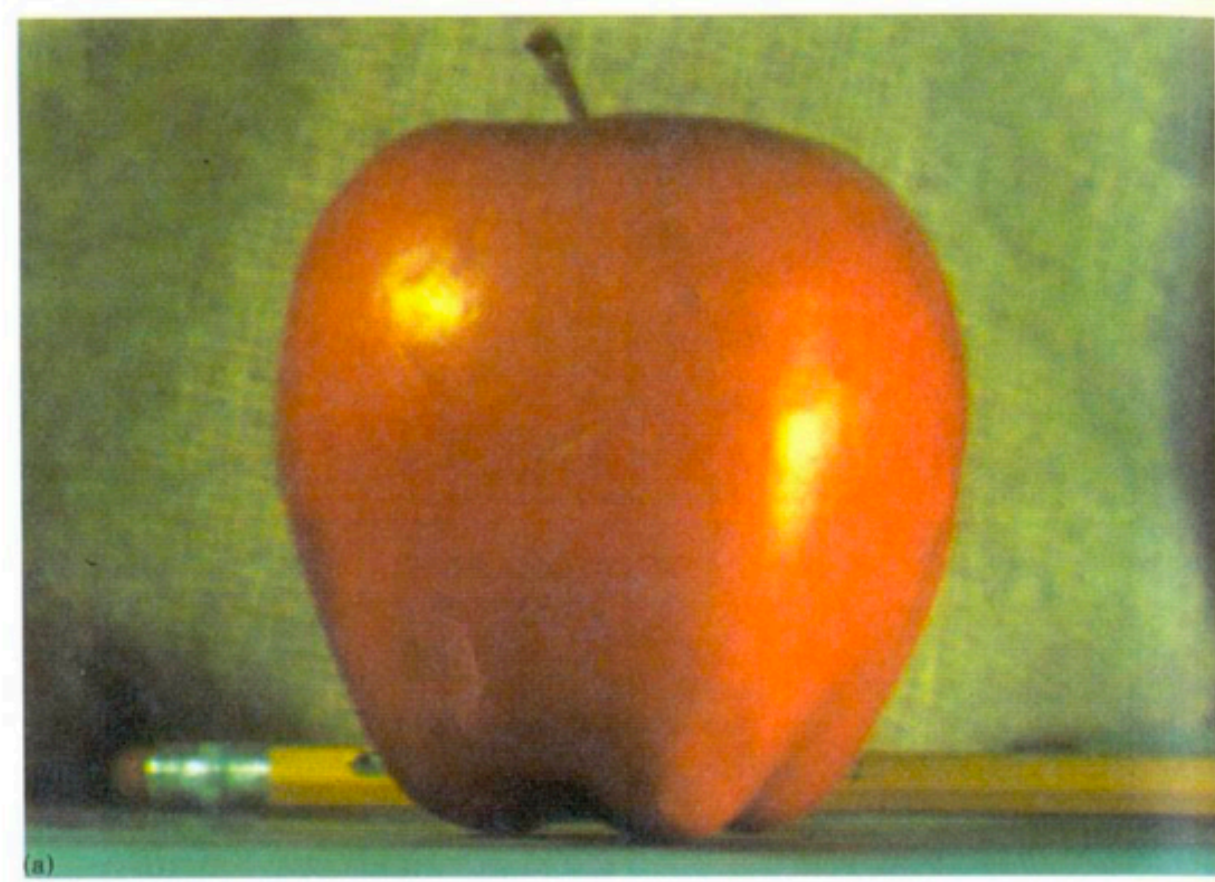
blend



Right pyramid

**Burt and Adelson**, "A multiresolution spline with application to image mosaics," ACM Transactions on Graphics, 1983, Vol.2, pp.217-236.

# Aside: Image Blending



**Burt and Adelson**, "A multiresolution spline with application to image mosaics," ACM Transactions on Graphics, 1983, Vol.2, pp.217-236.

# Aside: Image Blending

## Algorithm:

1. Build Laplacian pyramid  $LA$  and  $LB$  from images  $A$  and  $B$
2. Build a Gaussian pyramid  $GR$  from mask image  $R$  (the mask defines which image pixels should be coming from  $A$  or  $B$ )
3. Form a combined (blended) Laplacian pyramid  $LS$ , using nodes of  $GR$  as weights:  $LS(i,j) = GR(i,j) * LA(i,j) + (1-GR(i,j)) * LB(i,j)$
4. Reconstruct the final blended image from  $LS$

# Aside: Image Blending

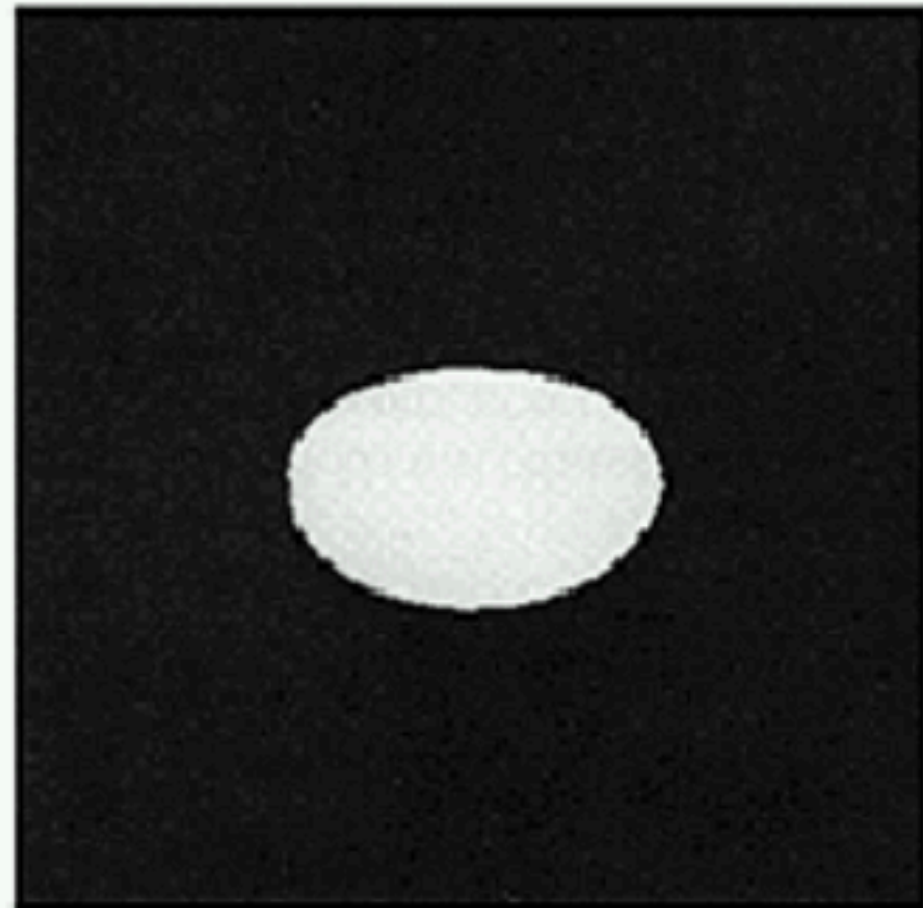
**left**



**right**



**mask**

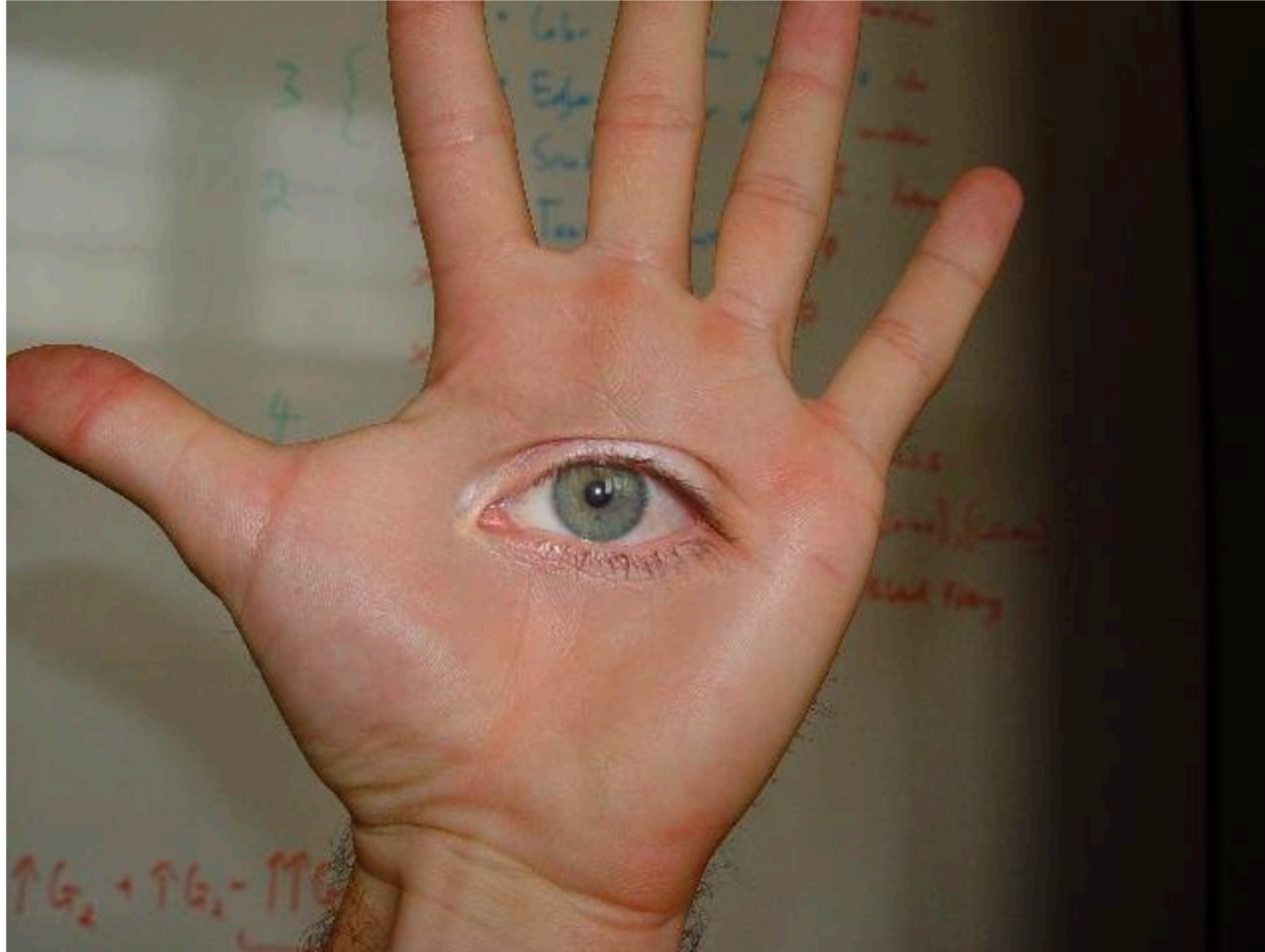


**blended**





# Aside: Image Blending



© david dmartin (Boston College)

# Aside: Image Blending



© Chris Cameron

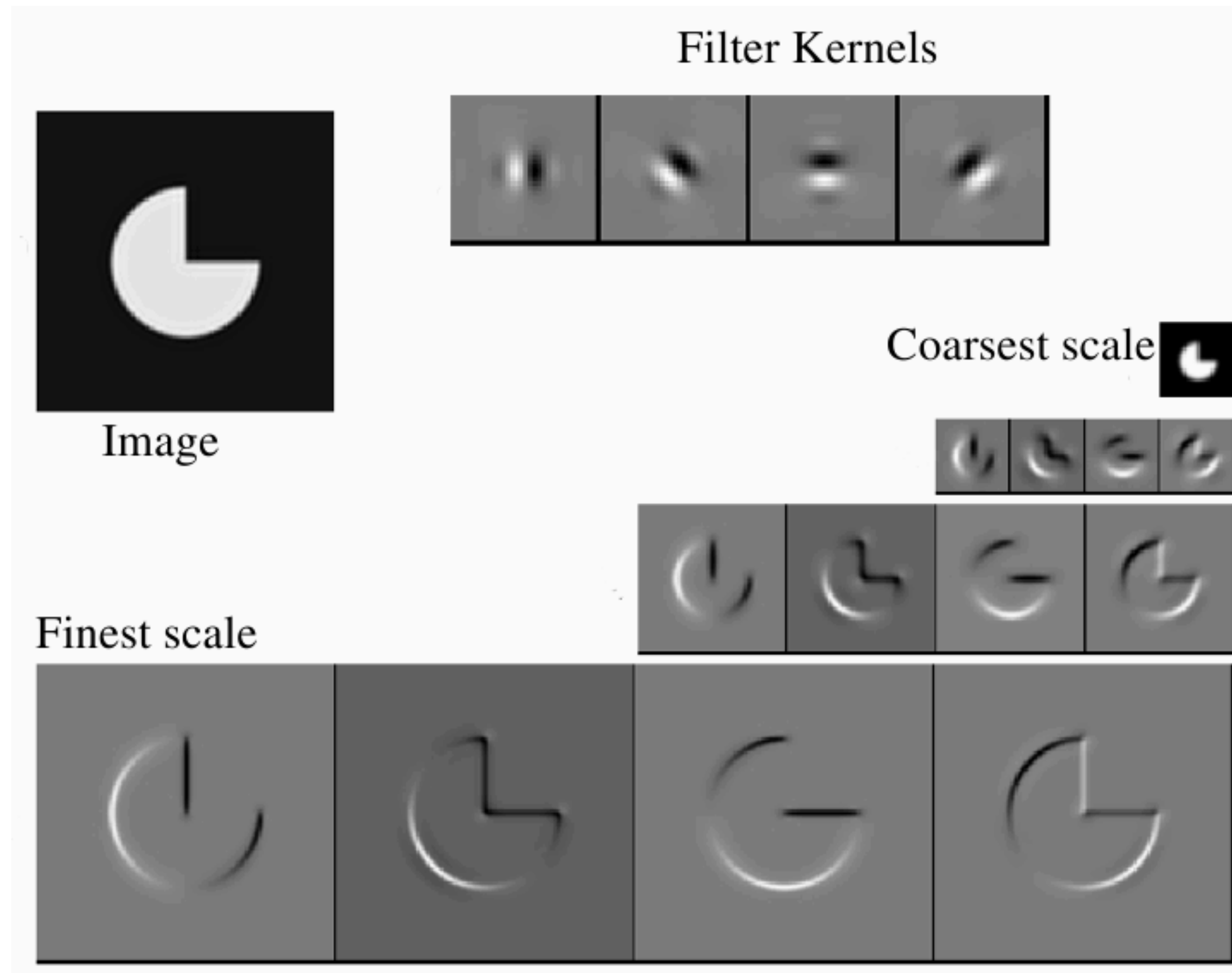
# Oriented Pyramids

Laplacian pyramid is orientation independent

**Idea:** Apply an oriented filter at each layer

- represent image at a particular scale and orientation
- *Aside:* We do not study details in this course

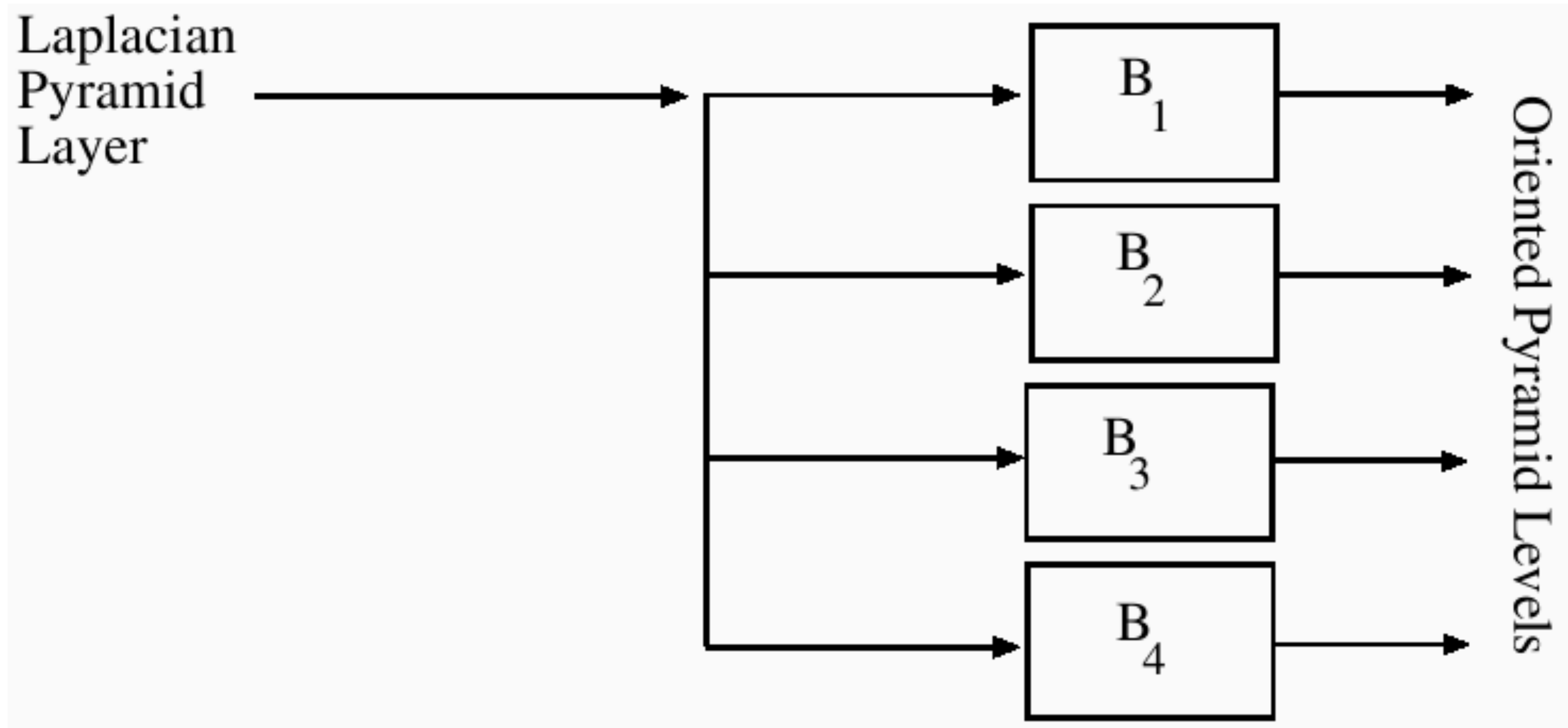
# Oriented Pyramids



Forsyth & Ponce (1st ed.) Figure 9.13

# Oriented Pyramids

Oriental Filters



Forsyth & Ponce (1st ed.) Figure 9.14

# Final Texture Representation

## Steps:

1. Form a Laplacian and oriented pyramid (or equivalent set of responses to filters at different scales and orientations)
2. Square the output (makes values positive)
3. Average responses over a neighborhood by blurring with a Gaussian
4. Take statistics of responses
  - Mean of each filter output
  - Possibly standard deviation of each filter

Please get your **iClickers** — Quiz