

# Digital Image Processing (CSE 478)

## Lecture 6: Image resampling

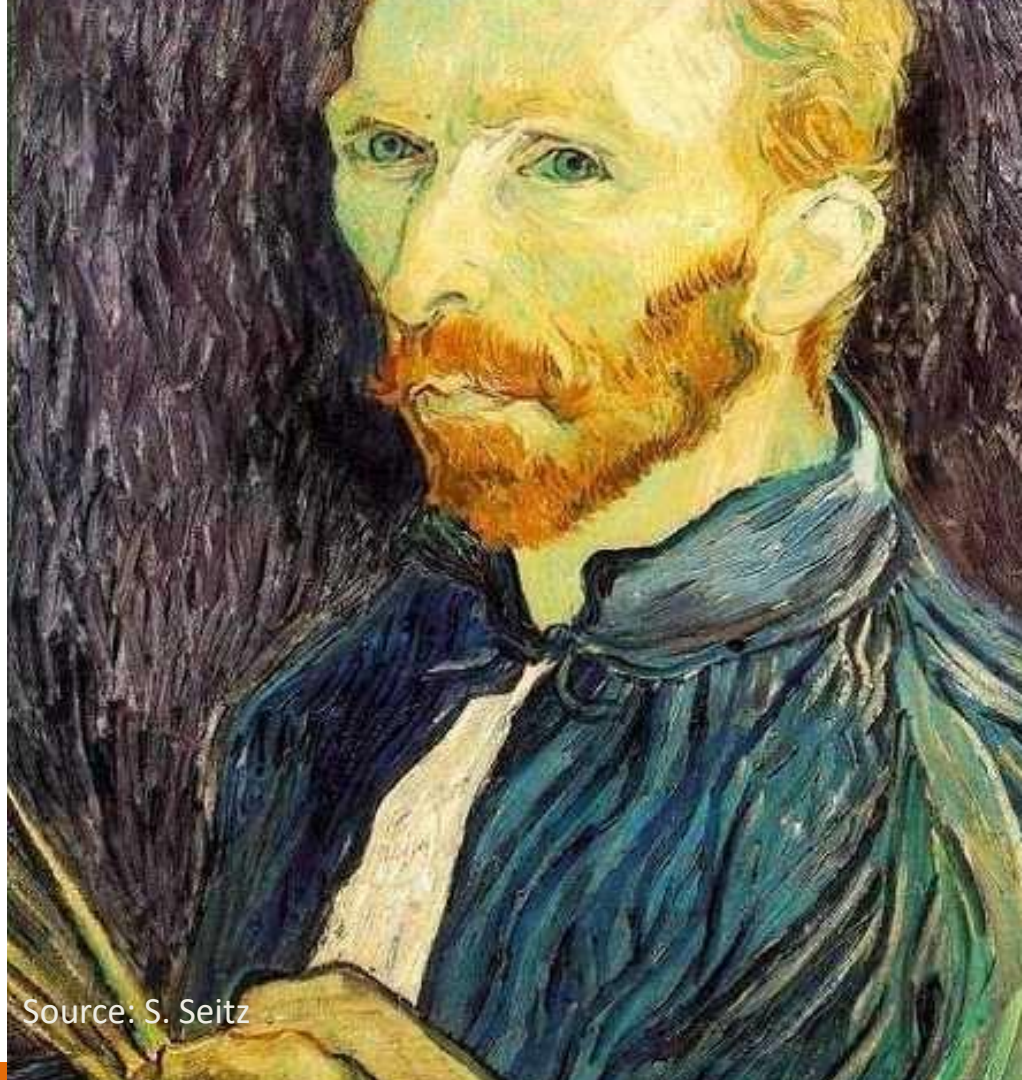
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Center for Visual Information Technology (CVIT), IIIT Hyderabad

# Today's Class

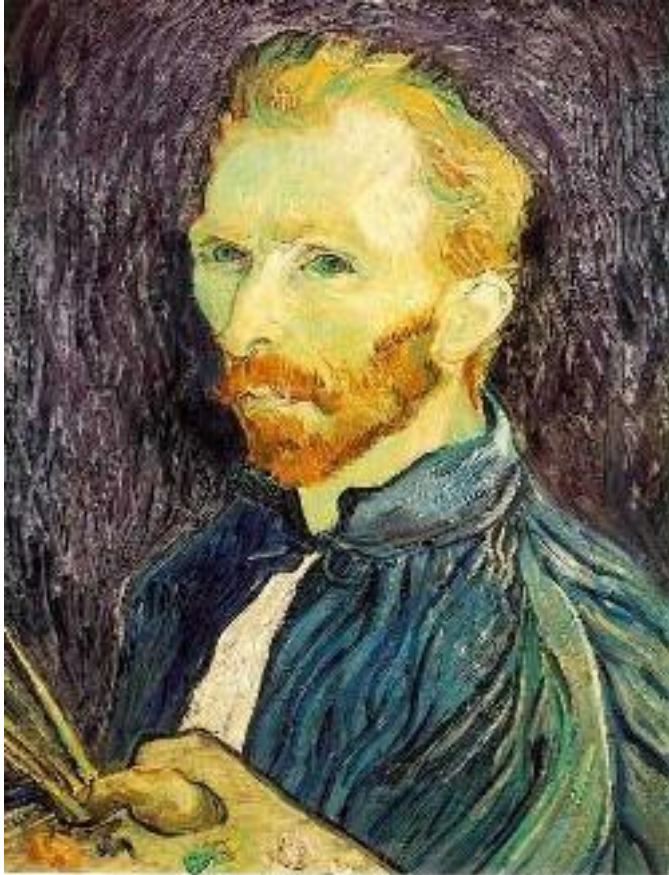
- Image down sampling
- Gaussian Pyramids
- Image up sampling

Image too big to fit screen. How  
can we resize?



Source: S. Seitz

# Image down-sampling

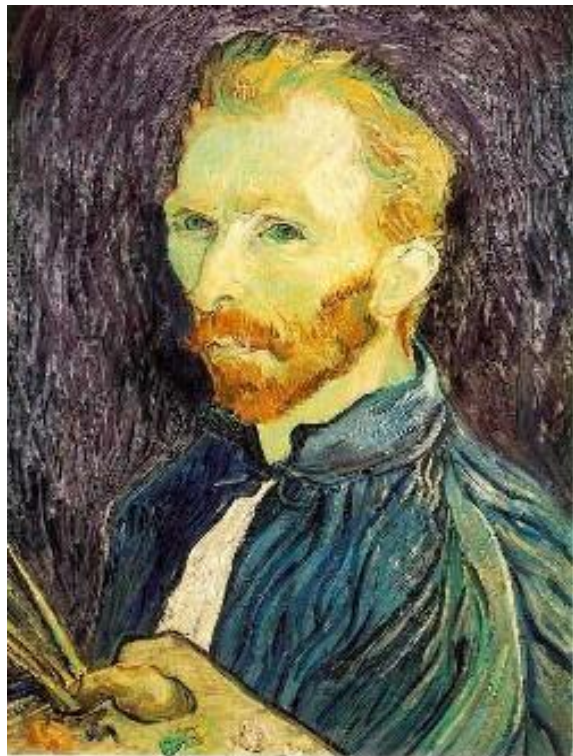


Throw away every other row and column to create a 1/2 size image





# Image down-sampling



1/2



1/4 (2x zoom)



1/8 (4x zoom)

# Image down-sampling



Courtesy: F. Durand

# Image down-sampling



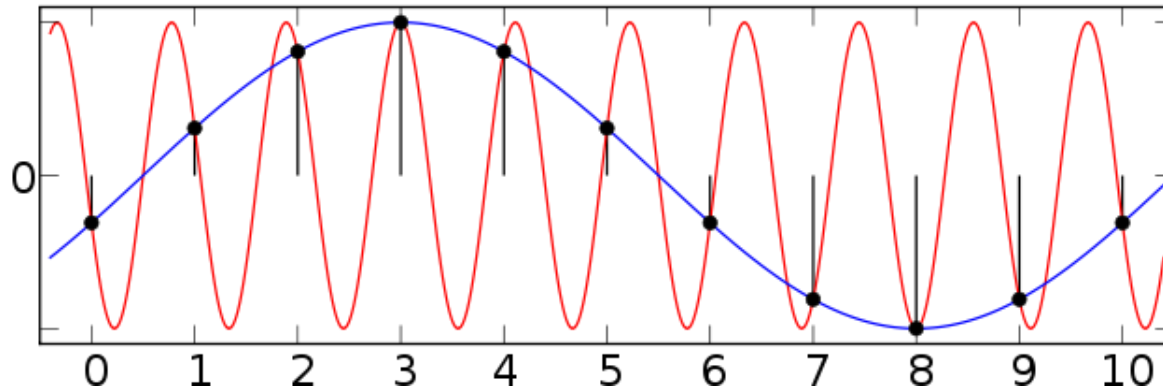
Original



1/2 (2x zoom)

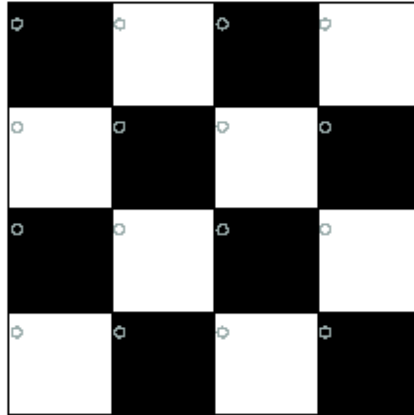
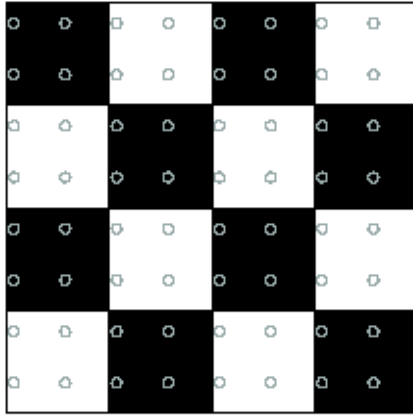
# Aliasing

- Sampling rate is not enough to capture the amount of detail
- To avoid aliasing
  - sampling rate  $\geq 2 * \text{max frequency in the image}$  (two samples per cycle)
  - minimum sampling rate is called Nyquist rate (on the basis of sampling theorem proposed by Harry Nyquist and Claude Shannon )

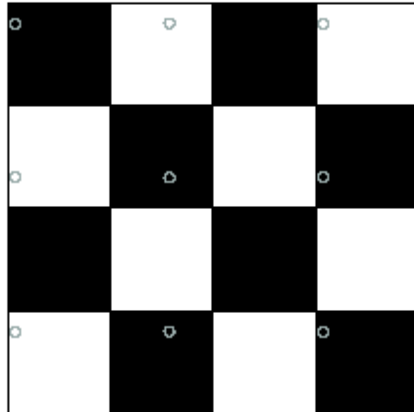
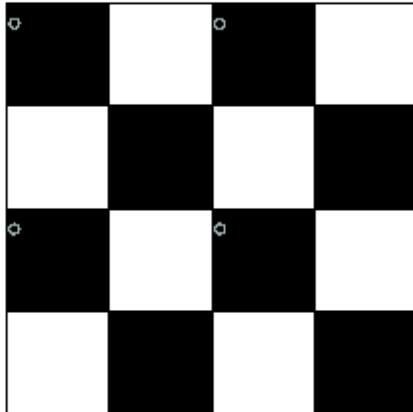




# Nyquist limit



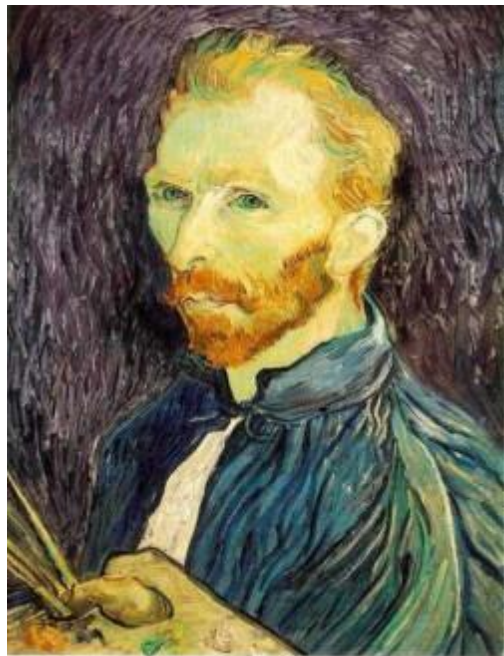
Good sampling



Bad sampling

# Gaussian pre-filtering

- Solution: filter the image, *then* subsample



Gaussian 1/2



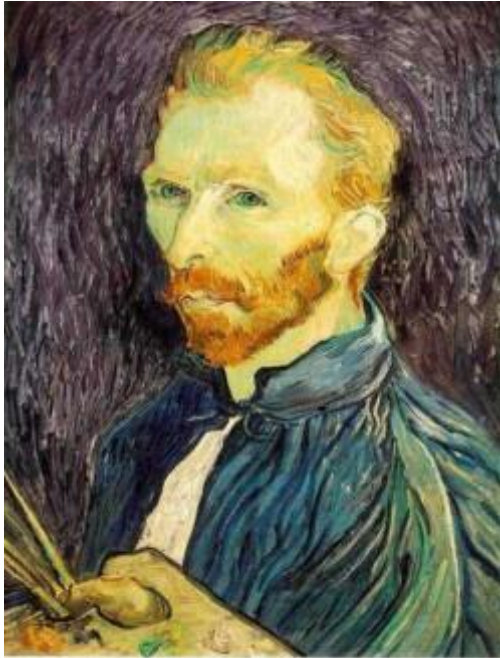
G 1/4



G 1/8

# Down-sampling with Gaussian pre-filtering

- Solution: filter the image, *then* subsample



Gaussian 1/2

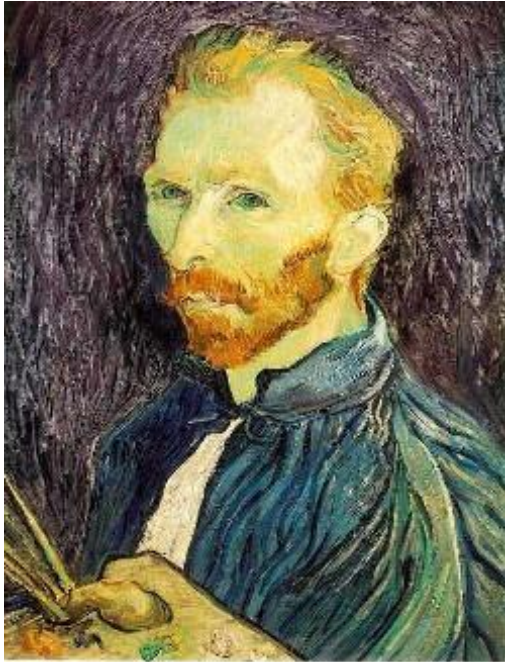


Gaussian 1/4 (2x zoom)



Gaussian 1/8 (4x zoom)

Compare with...



1/2



1/4 (2x zoom)



1/8 (4x zoom)

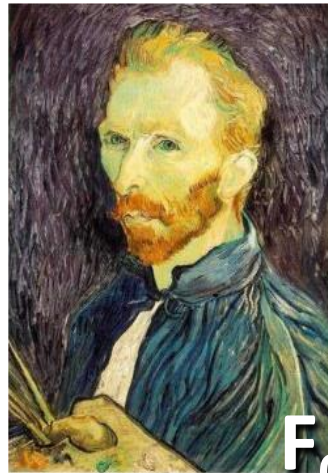
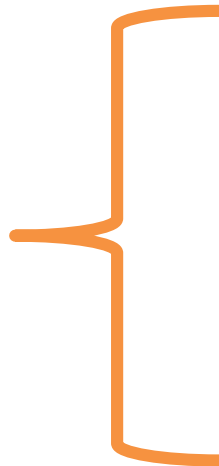


# Down-sampling with Gaussian pre-filtering

Smoothing removes high  
frequency components!



# Gaussian Pyramid

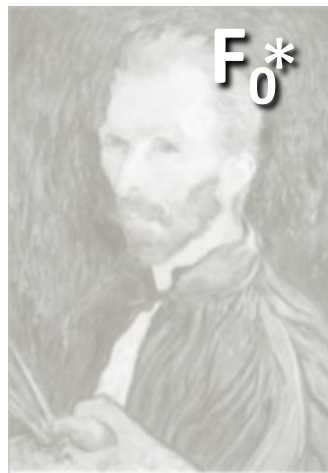


subsample



subsample ...

...



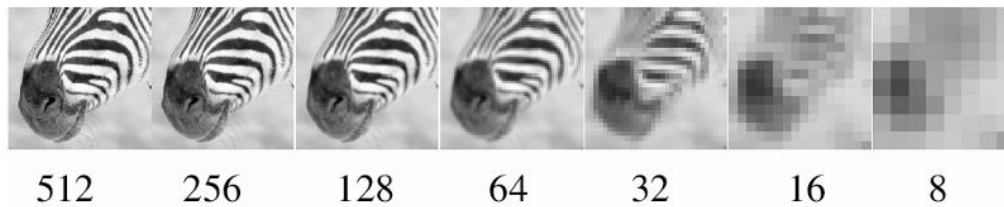
H



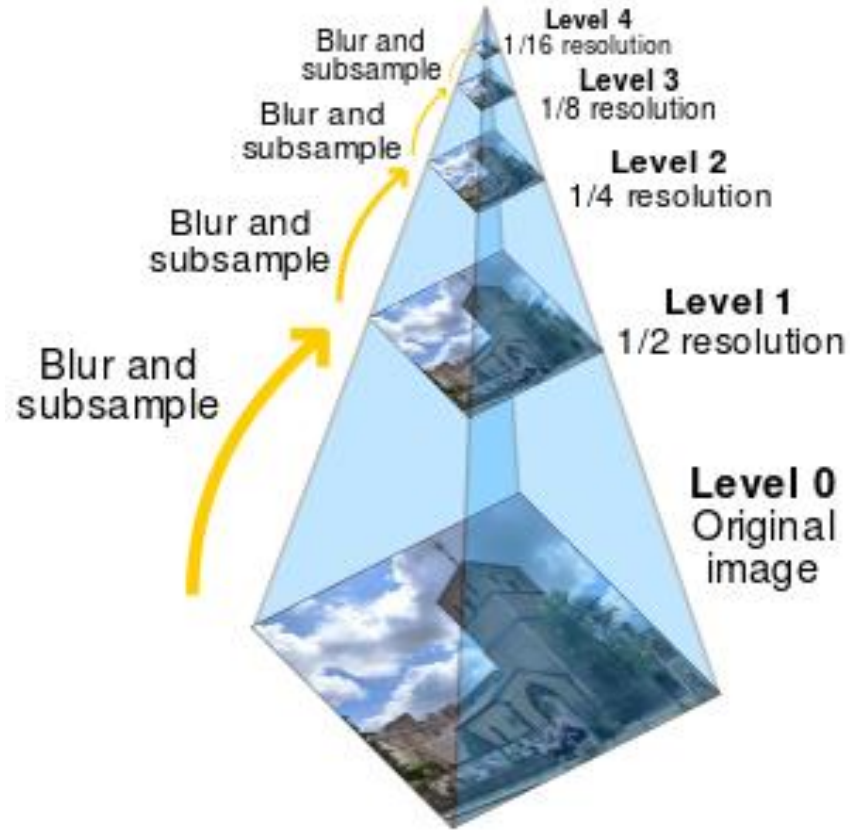
H



# Gaussian pyramid



# Gaussian pyramid





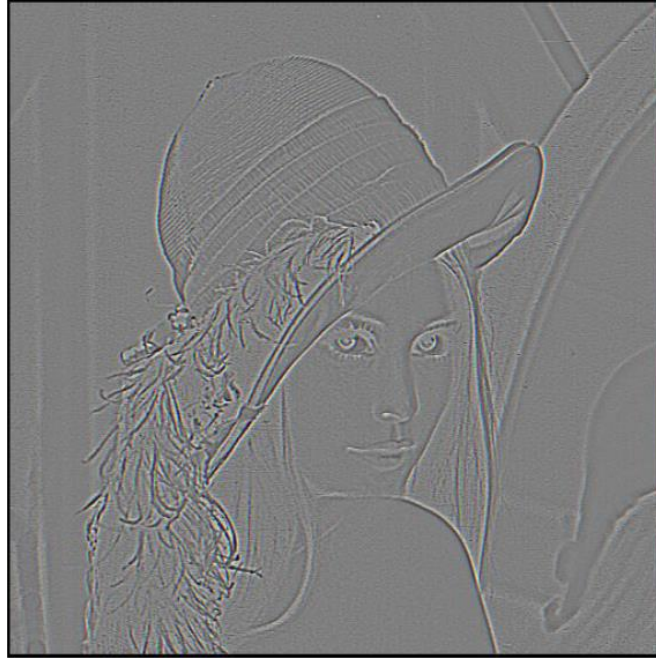
What does smoothing takes away?



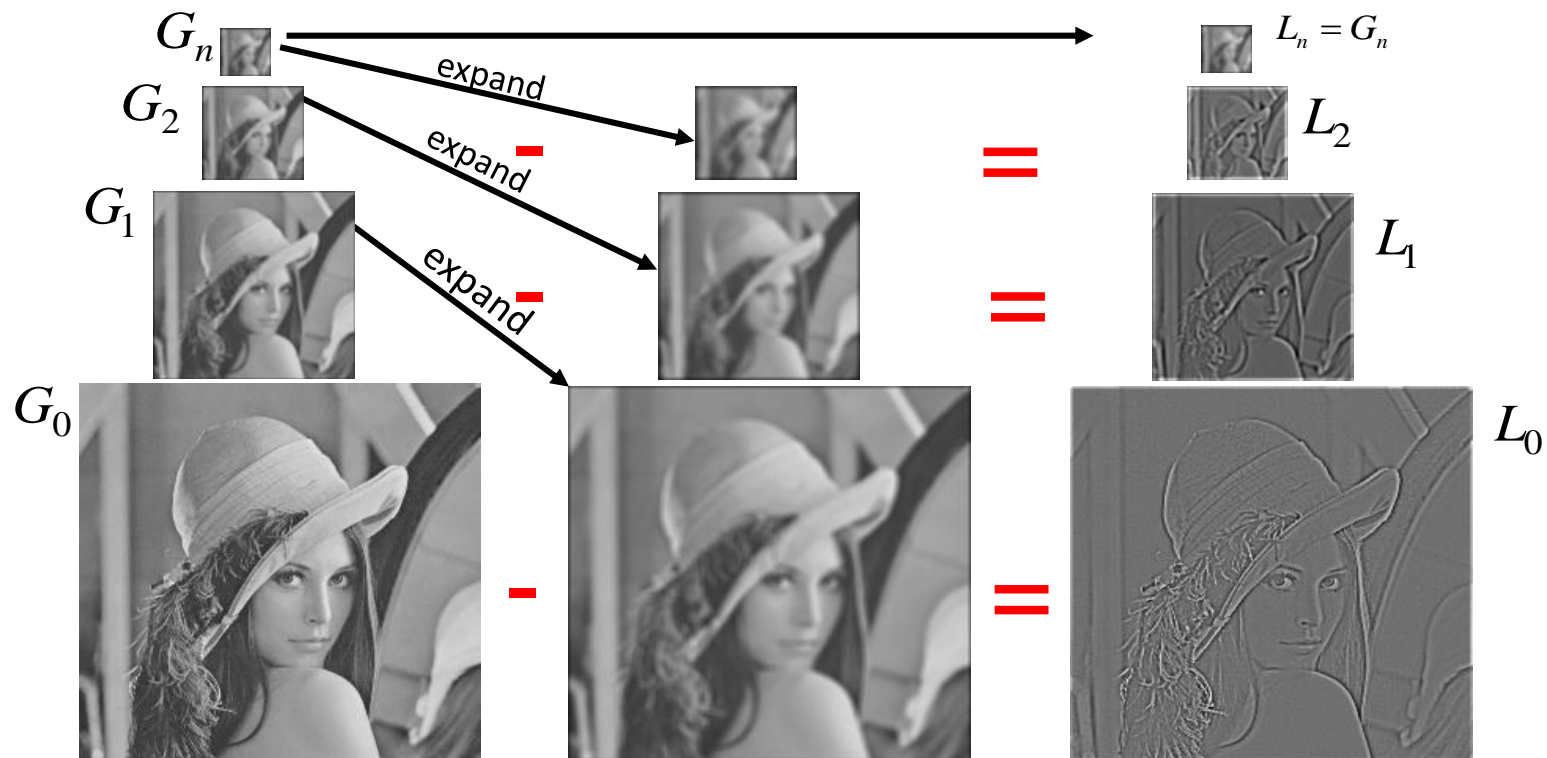
What does smoothing takes away?



# What does smoothing takes away?

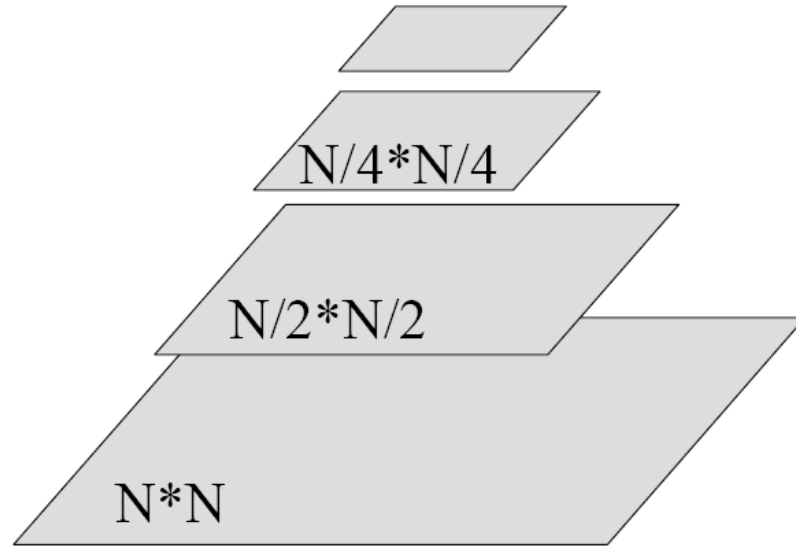


# Laplacian pyramid





# Space required for image pyramid



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

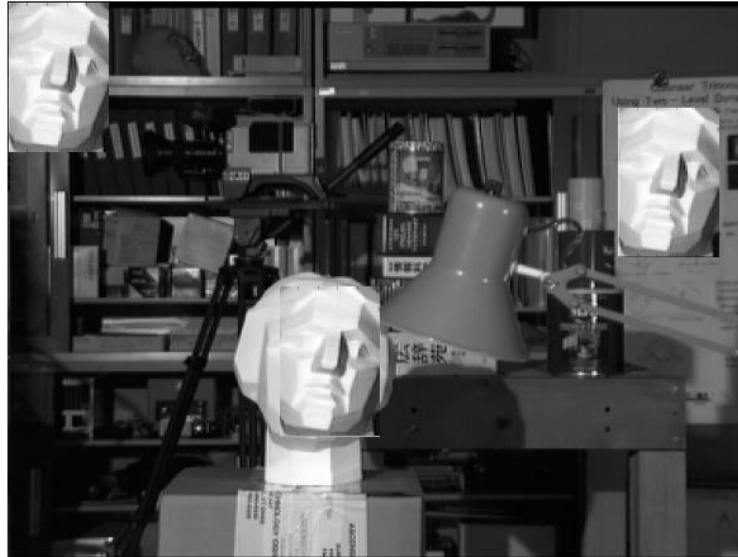
# Application Gaussian pyramid

- Efficient multi-scale detection

*Template*



*Search Region*



# Efficient multi-scale detection

*Template*



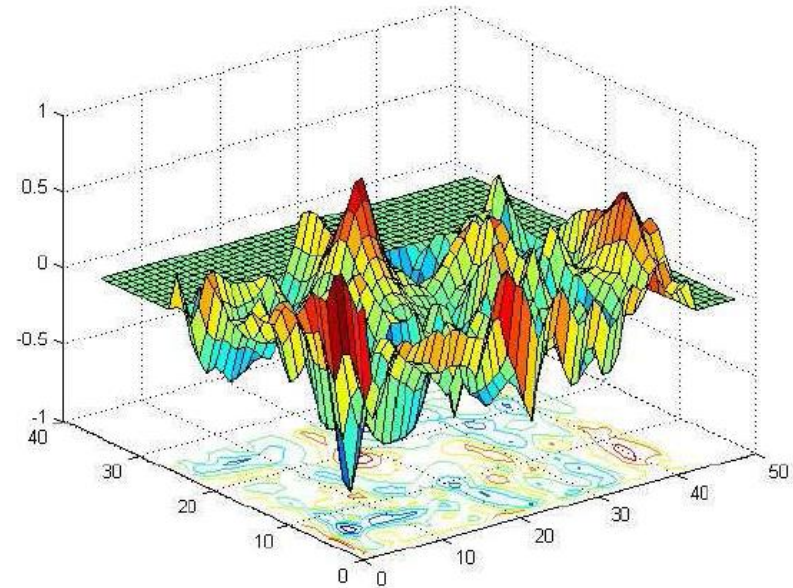
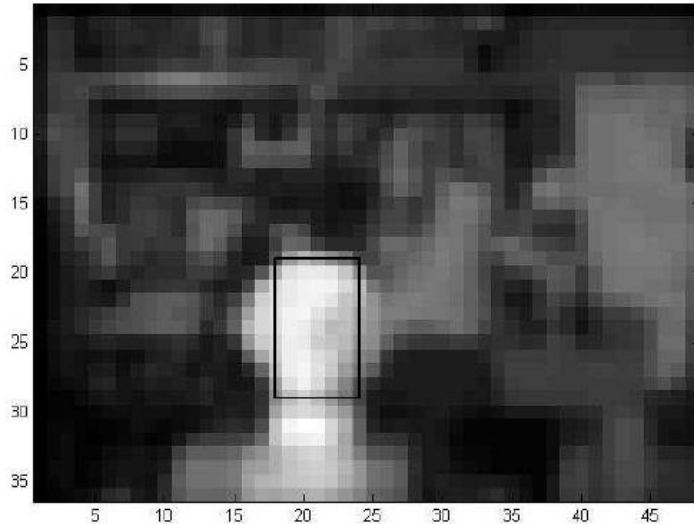
*Search Region*

Original Image



# Efficient multi-scale detection

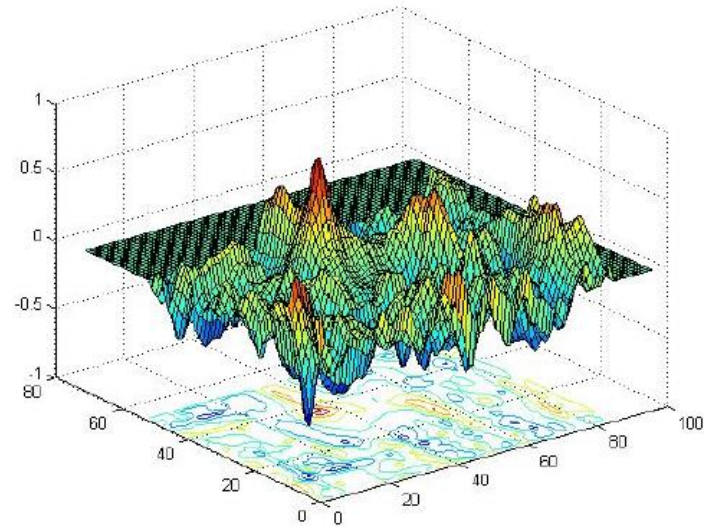
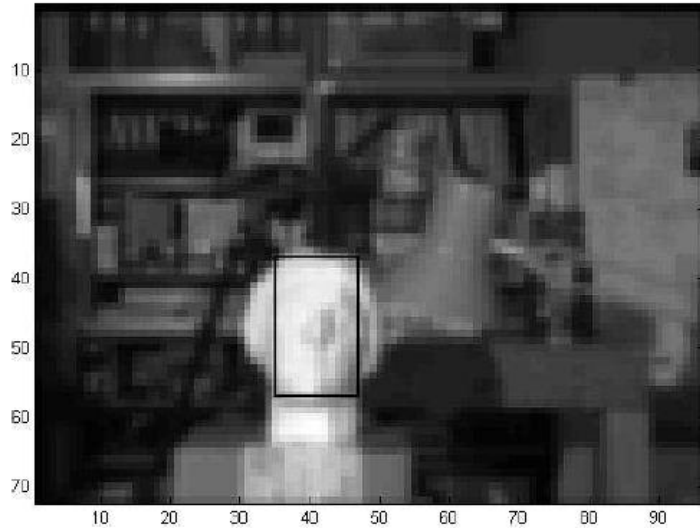
- Level 3 search: at the lowest level we search the entire image with correlation template





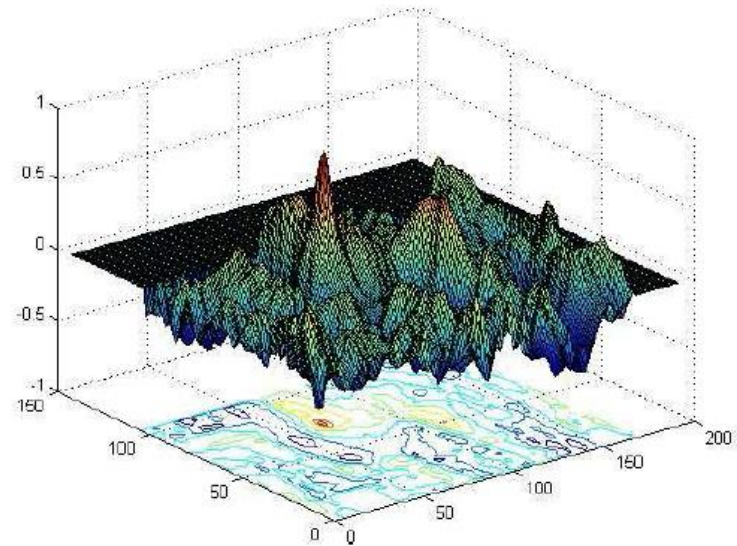
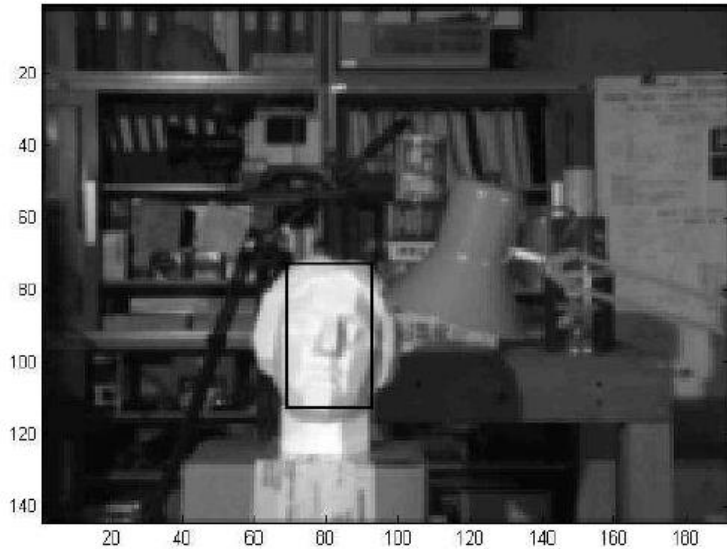
# Efficient multi-scale detection

- Level 2 search: constrained to a neighbourhood of high response centers in the previous level



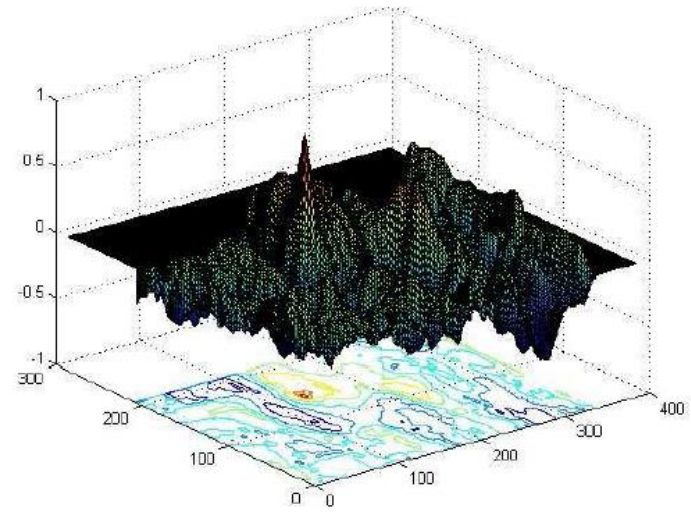
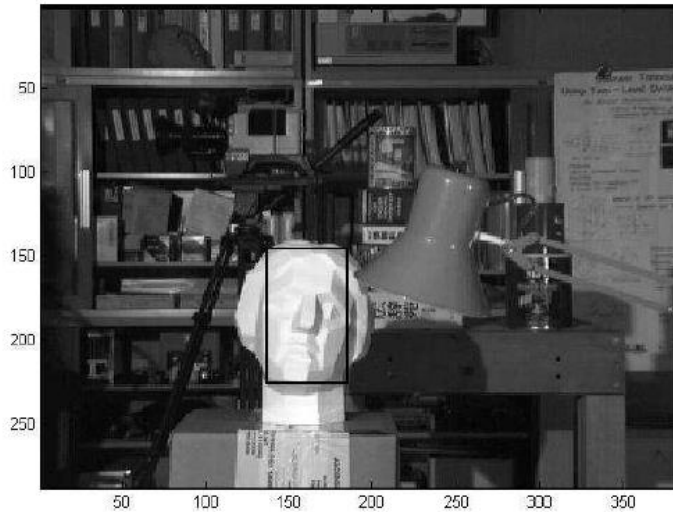
# Efficient multi-scale detection

- Level 1 search: again constrained based on results of level 1



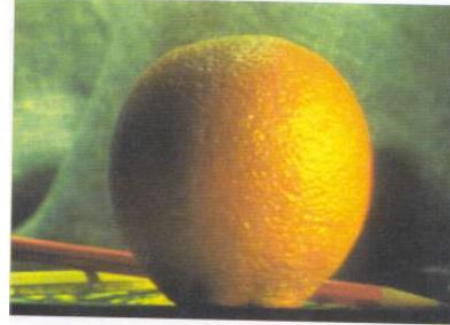
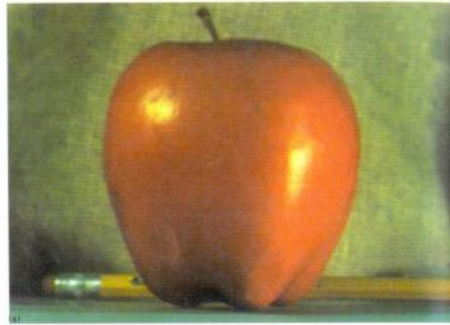
# Efficient multi-scale detection

- Level 0 search: total time reduced to 0.5 second from 31 seconds

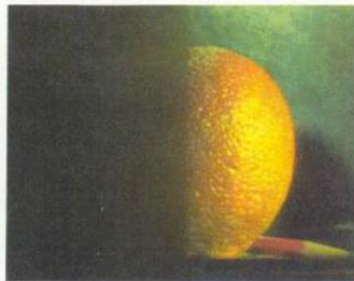


# Application Gaussian pyramid

- Blending Apples and Oranges



(d)



(h)

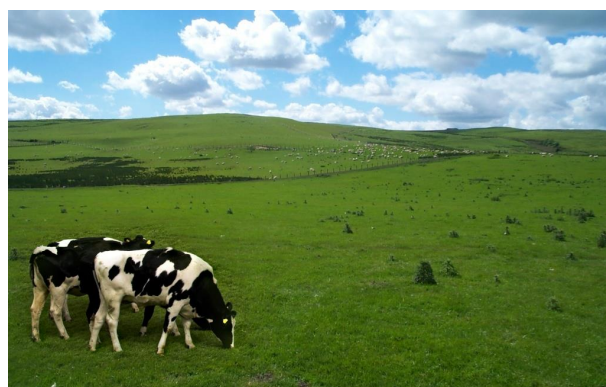


(l)

Burt and Adelson 1983



# Application Gaussian pyramid

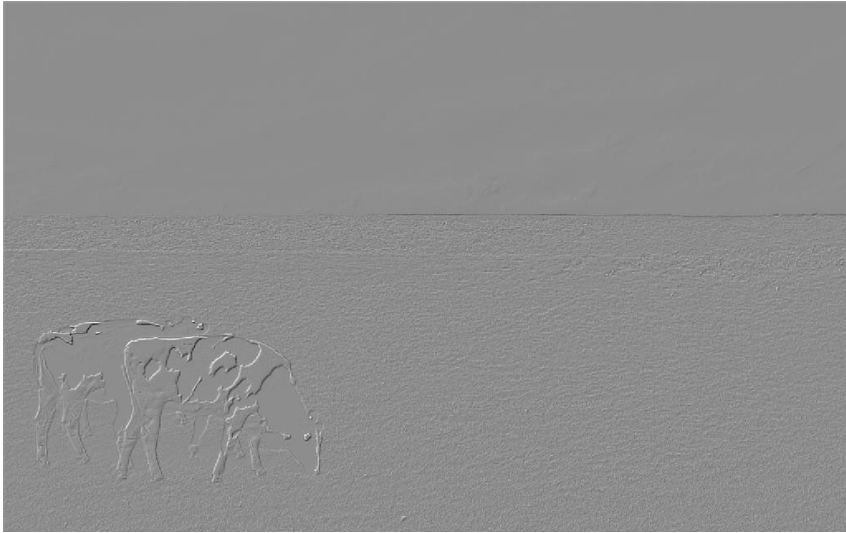




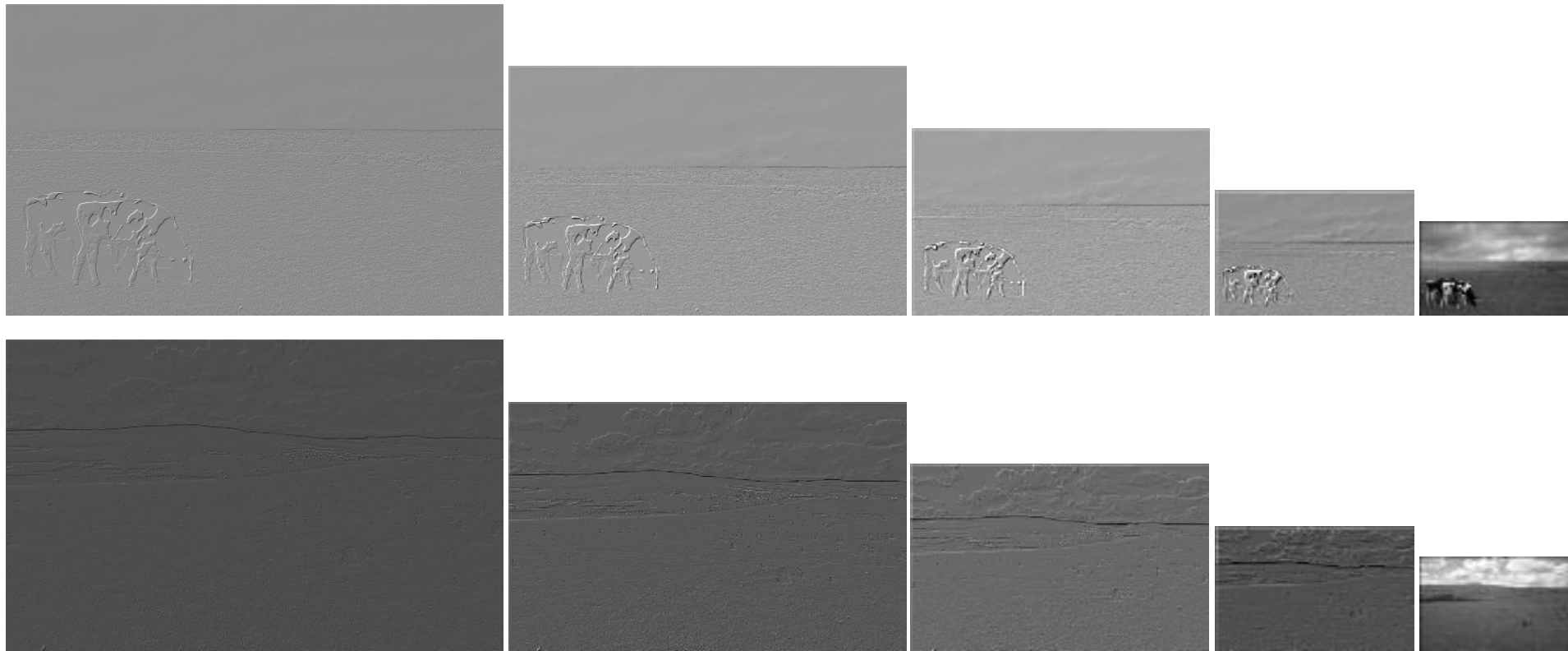




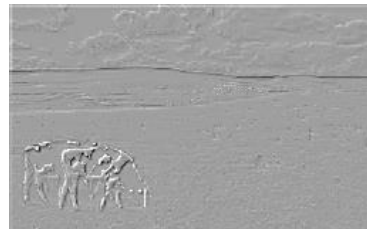
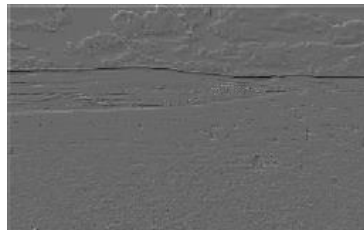
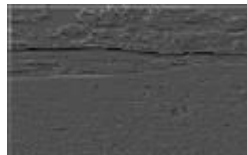
# Application Gaussian pyramid



# Application Gaussian pyramid



# Application Gaussian pyramid



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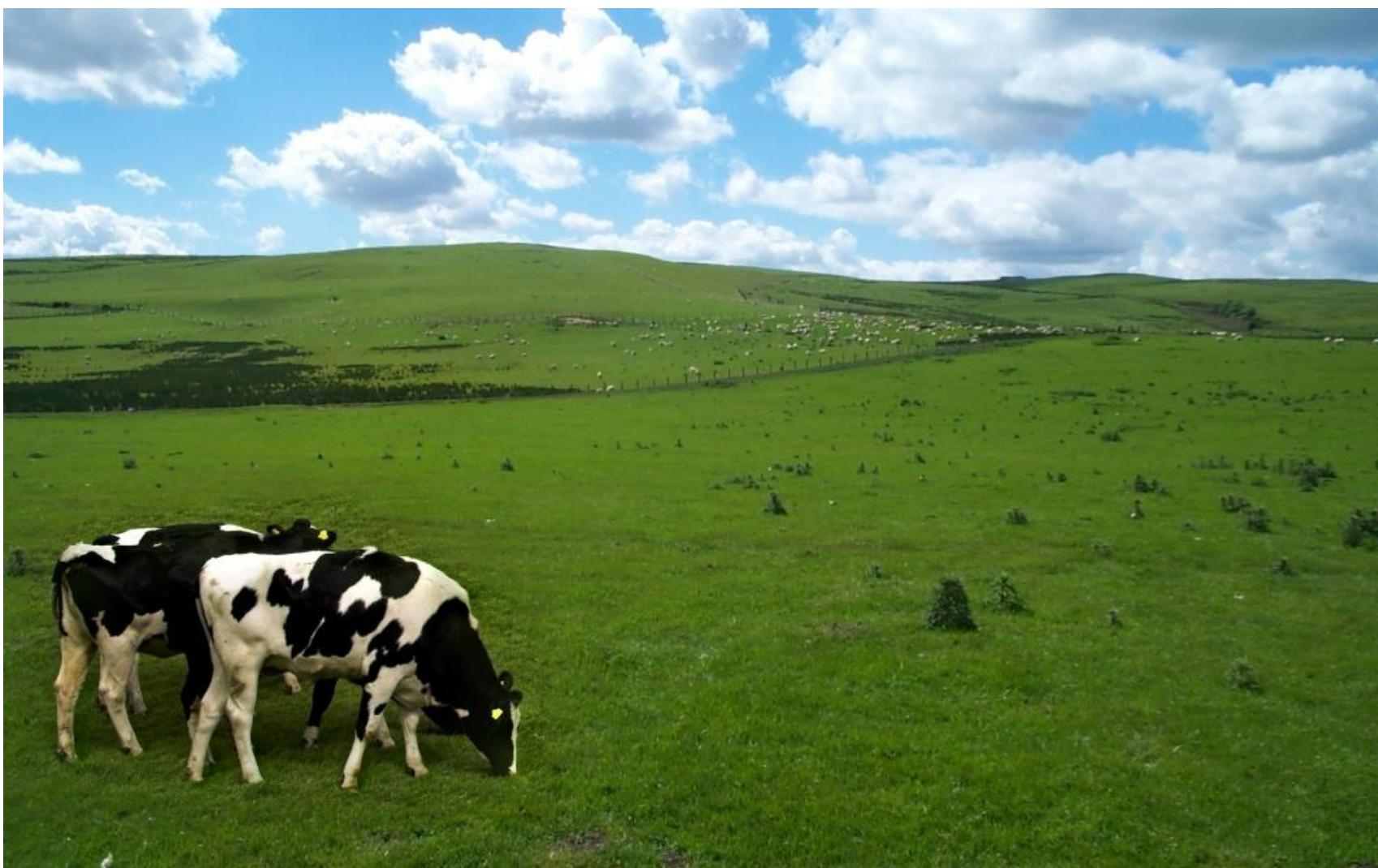
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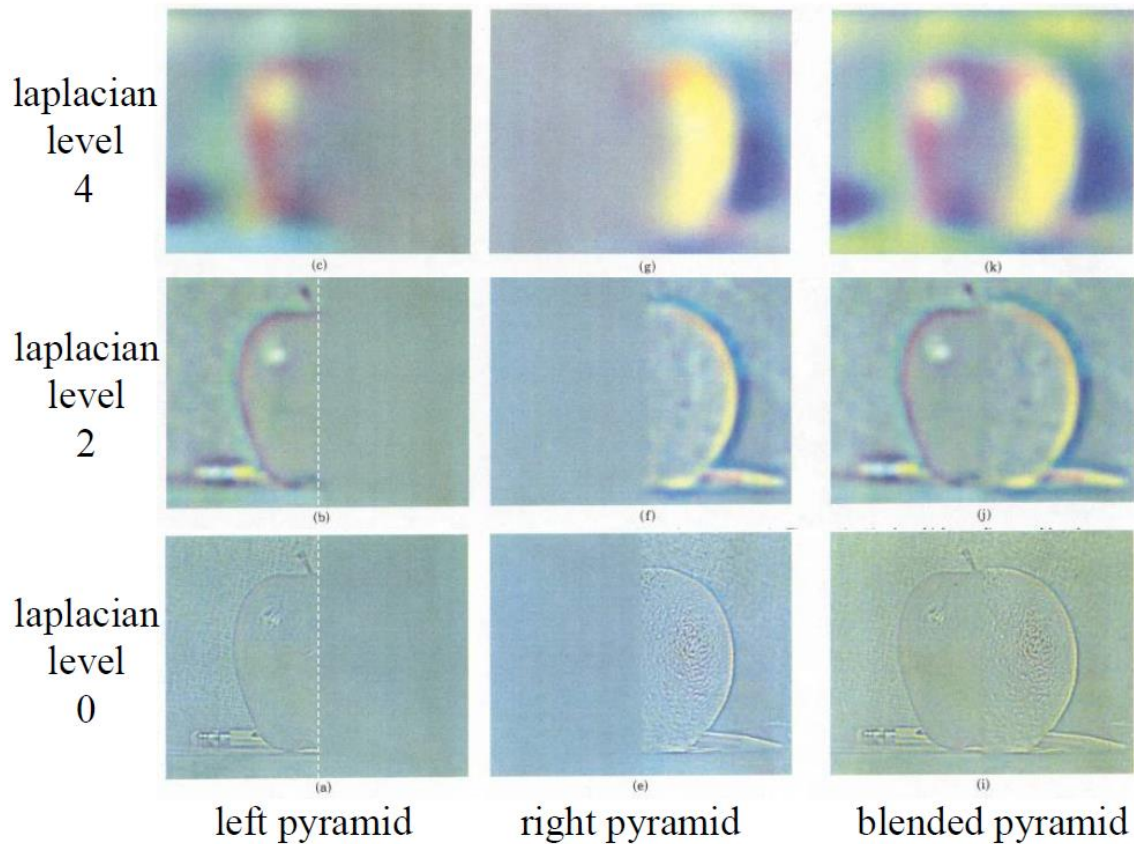




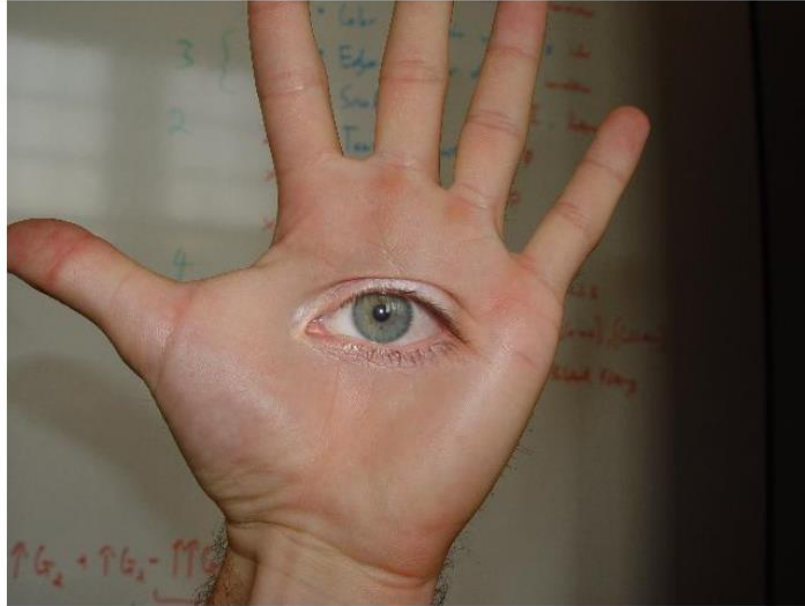




# Image Blending



# Image Blending



© prof. dmartin

# Questions?



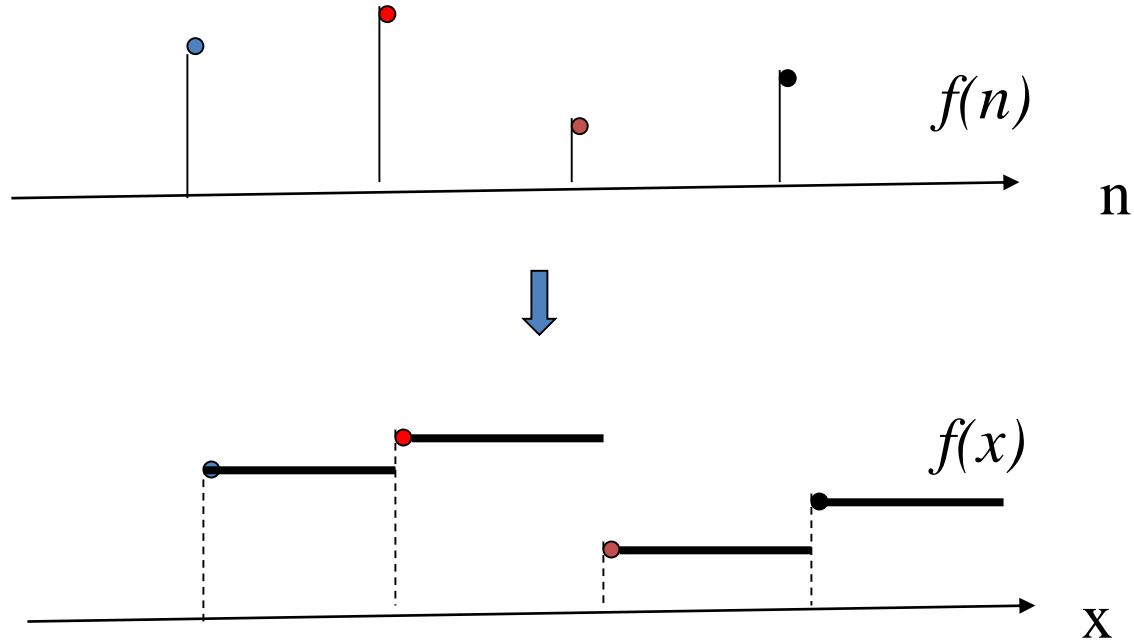


# Up-sampling



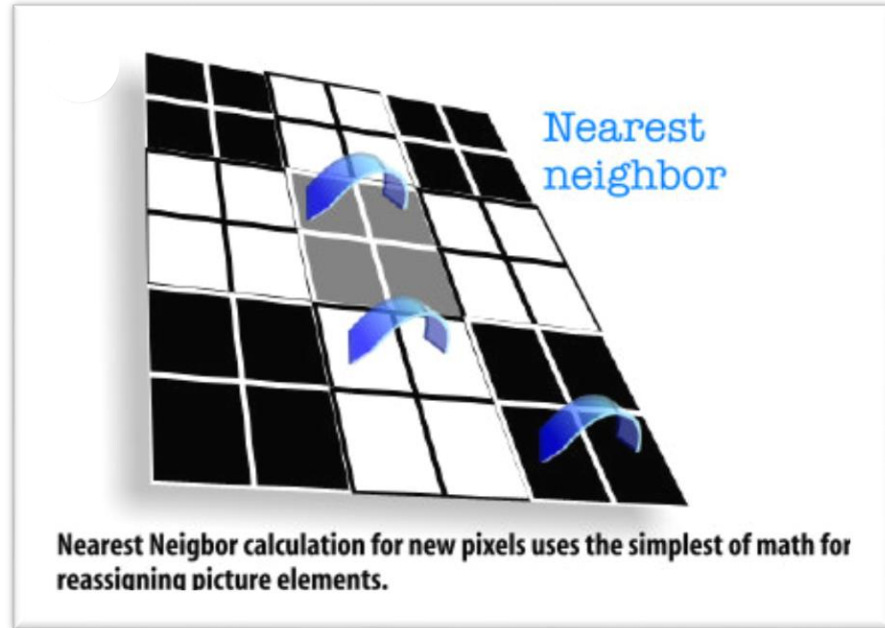
# Nearest neighbor interpolation

- Just repeat elements



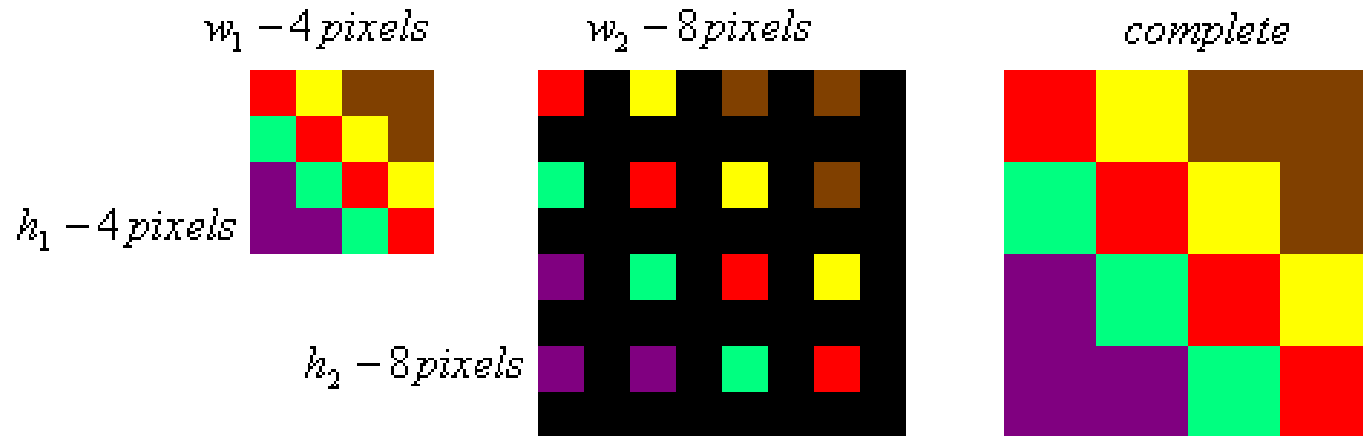
# Nearest Neighbour Interpolation

- Just repeat elements



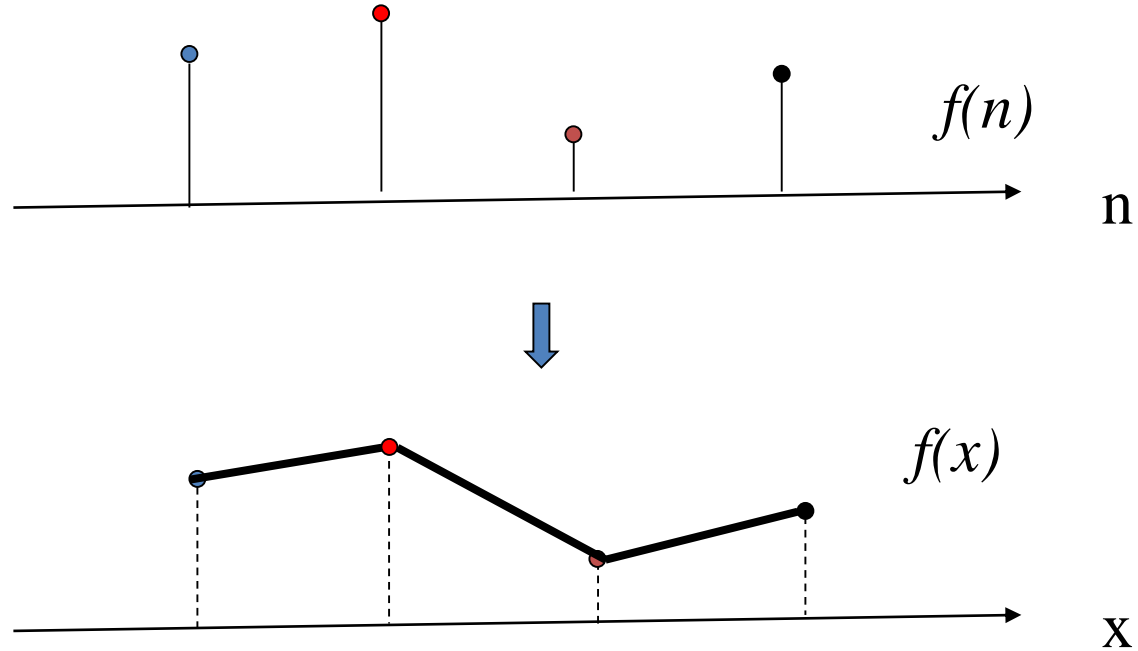
# Nearest Neighbour Interpolation

- Just repeat elements



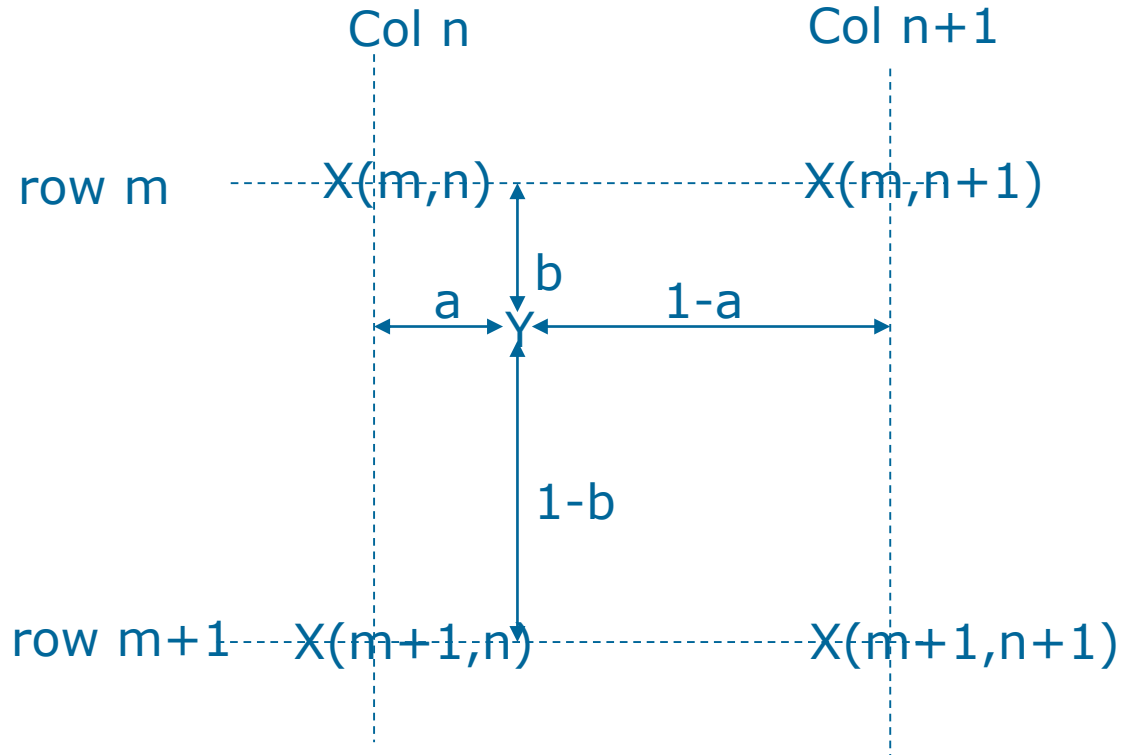
# Linear interpolation

- Linear combination



# Bilinear interpolation (2D)

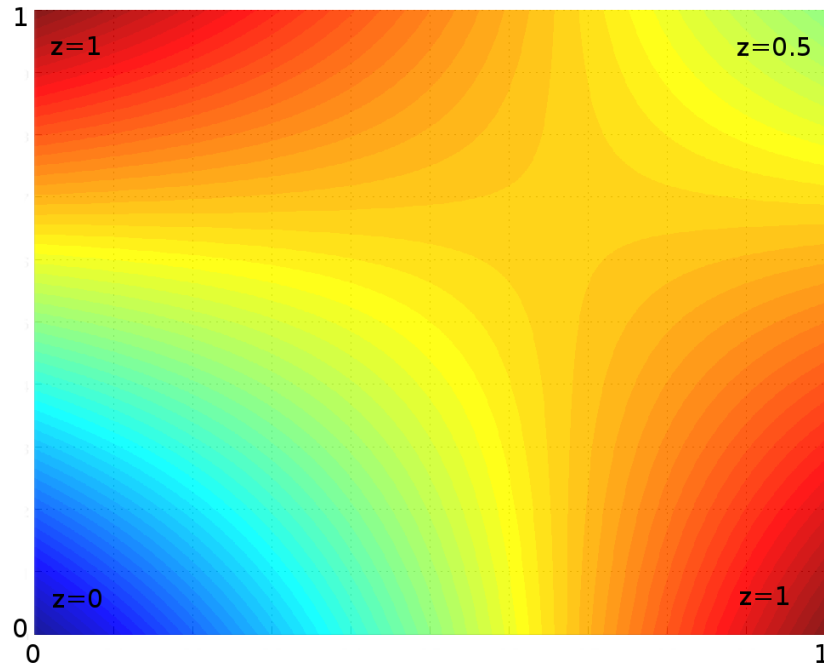
- Divide and conquer





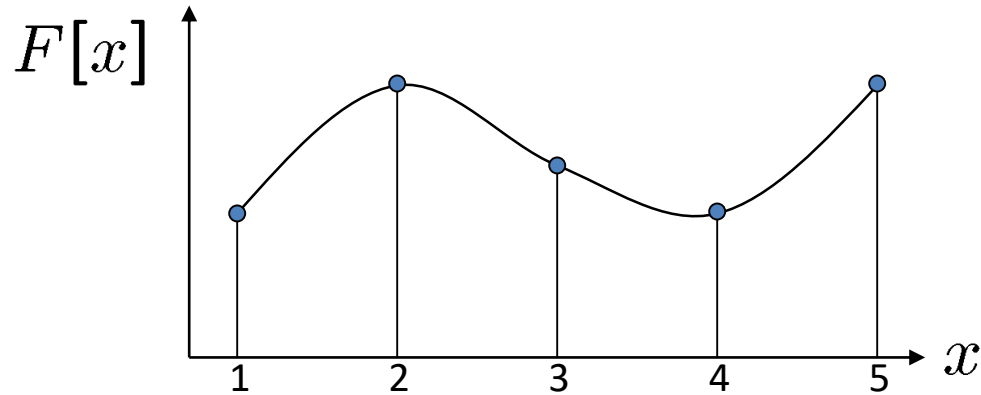
# Bilinear interpolation (2D)

- Although each step is linear in the sampled values and in the position, the interpolation as a whole is not linear but rather quadratic in the sample location.



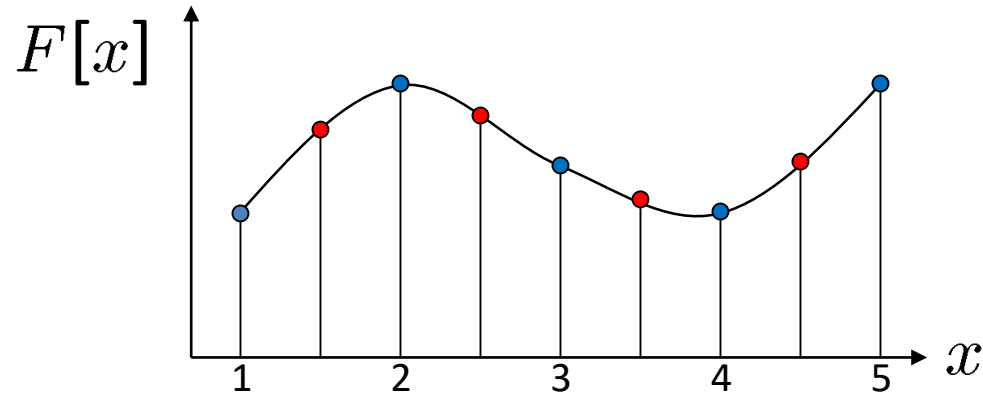
# Principled approach to interpolation

- Estimate the function from quantized values



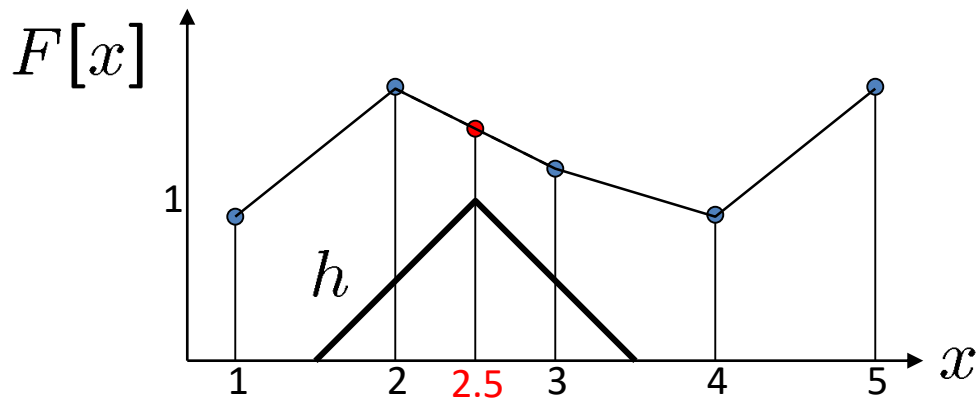
# Principled approach to interpolation

- Estimate the function from quantized values

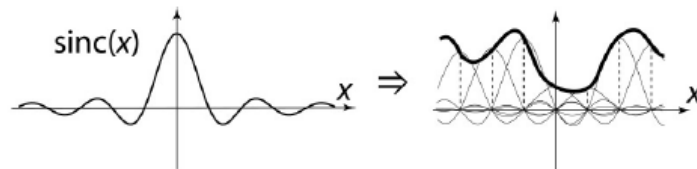


# Principled approach to interpolation

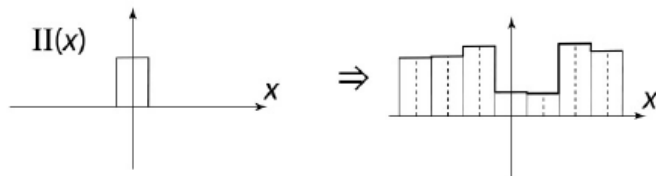
- Not always possible to estimate the function, what should we do?
- Approximation: Up-sampling as filtering



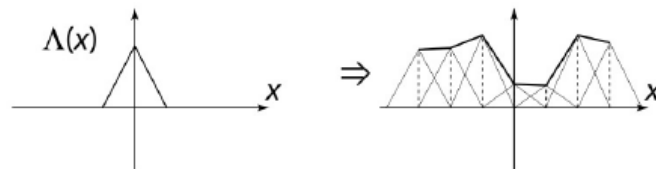
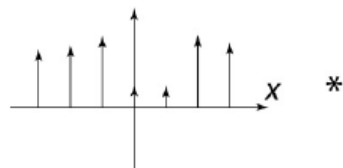
# Interpolation as filtering



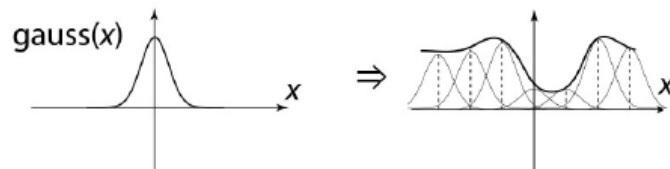
Sinc interpolation



Nearest-neighbor interpolation

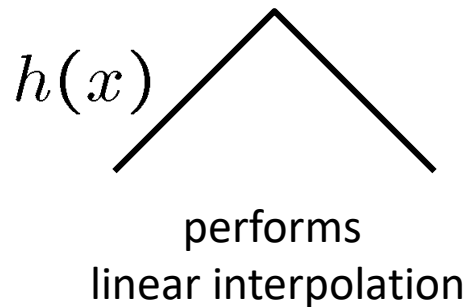


Linear interpolation

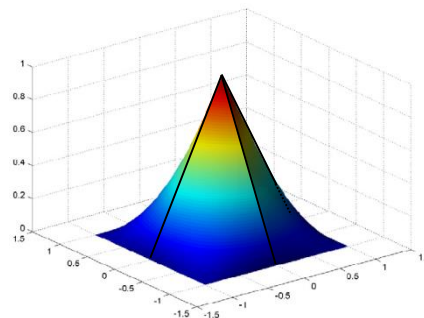


Gaussian reconstruction

# From 1D to 2D



$h(x, y)$

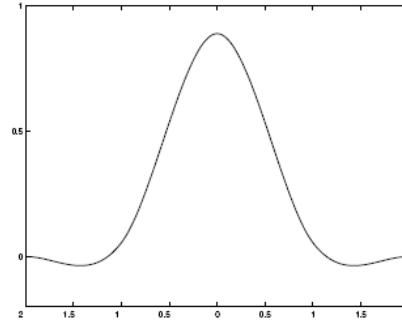


(tent function) performs  
**bilinear interpolation**



# Bicubic filter

- Commonly used



Cubic reconstruction filter

More advanced interpolation are adaptive, for example edge sensitive interpolation!

# Image interpolation

Original image:  x 10



Nearest-neighbor interpolation



Bilinear interpolation



Bicubic interpolation

# Image interpolation

Also used for *resampling*

