



SNS COLLEGE OF TECHNOLOGY

Coimbatore-35
An Autonomous Institution



Accredited by NBA – AICTE and Accredited by NAAC – UGC with 'A+' Grade
Approved by AICTE, New Delhi & Affiliated to Anna University, Chennai

DEPARTMENT OF ELECTRONICS & COMMUNICATION ENGINEERING

19ECE351 – IMAGE PROCESSING AND COMPUTER VISION

III B.E. ECE / V SEMESTER

UNIT 5 – COMPUTER VISION

TOPIC – Pyramids



Pyramids

- Downsampling (decimation)
- Upsampling (interpolation)
- Pyramids
 - Gaussian Pyramids
 - Laplacian Pyramids
- Applications
 - Template matching (object detection)
 - Detecting stable points of interest
 - Image Registration
 - Compression
 - Image Blending
 - ...

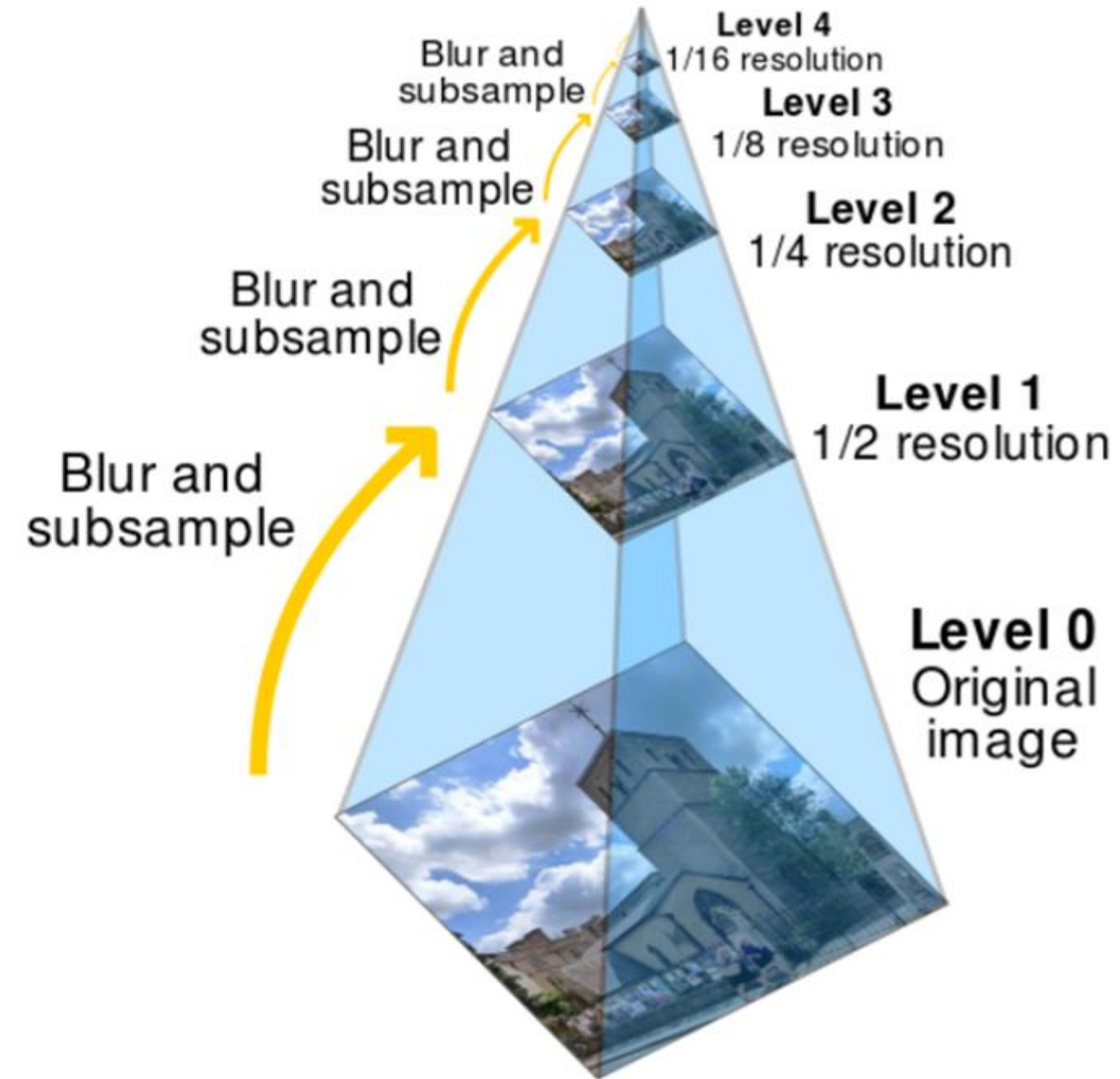




Image Scaling

- Assume that the image is too big for practical use:
 - Requires too much memory
 - Time consuming to process
 - Too big for the screen
 - ...
- A smaller image can be obtained by image sub-sampling





Image Downsampling



1/2



1/4



1/8

Throw away every other row and column → image reduced to $\frac{1}{2}$ size along each dimension.



Image Downsampling



1/2



1/4 (2x zoom)

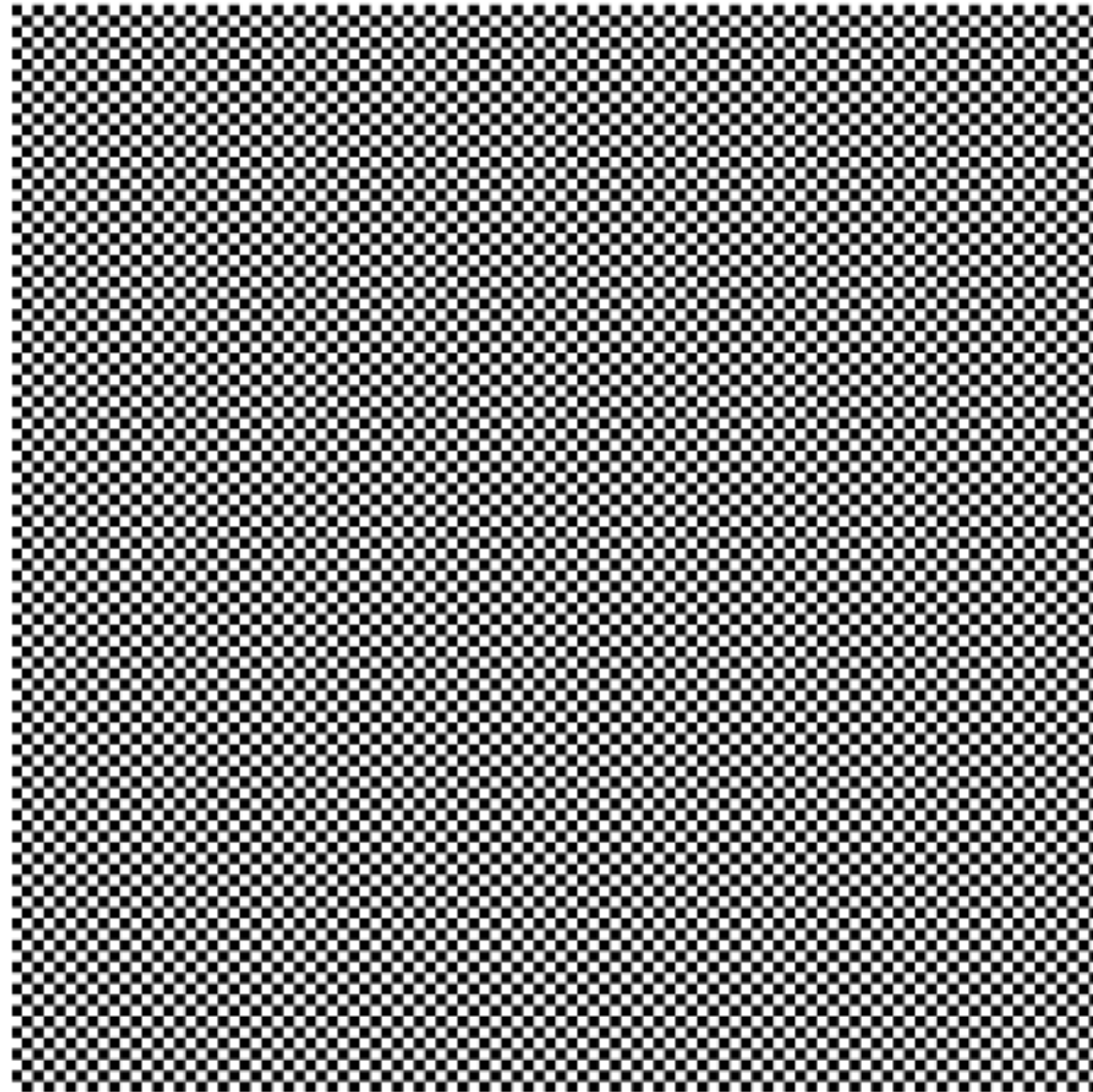


1/8 (4x zoom)

The subsampled images are of low quality. Why?

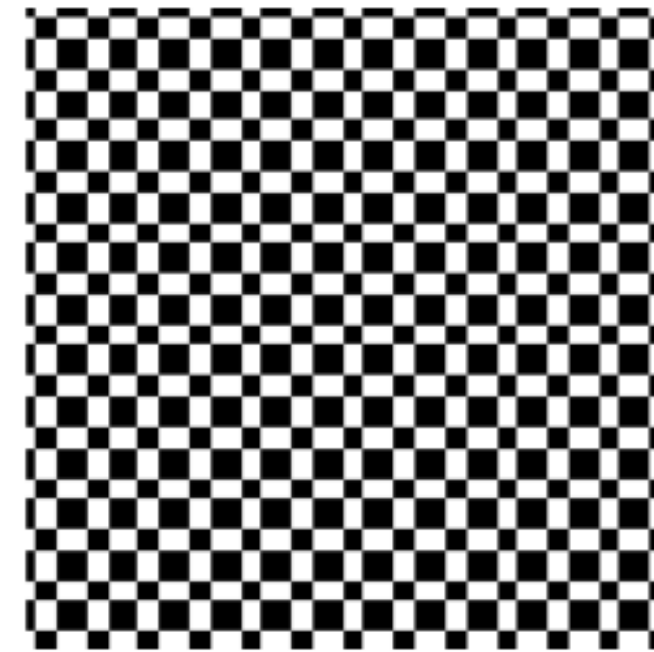


Spatial Undersampling

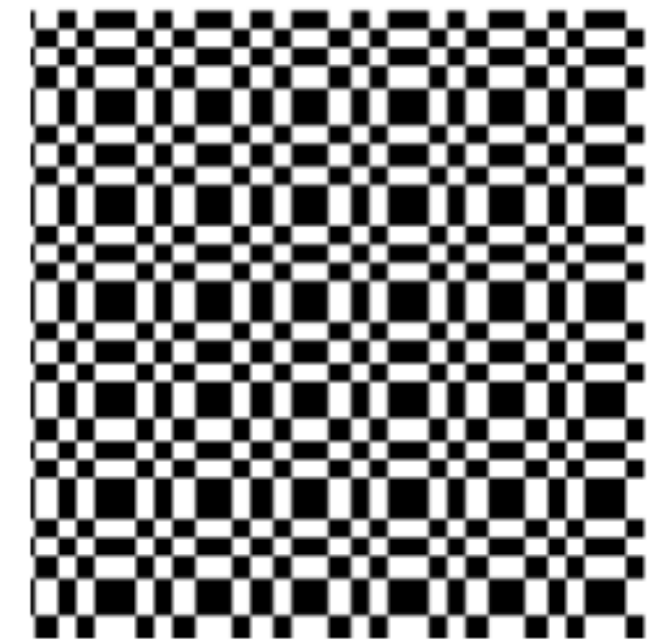


Checkerboard with 10 x 10 pixel squares

Downsampled images



1/10



1/16

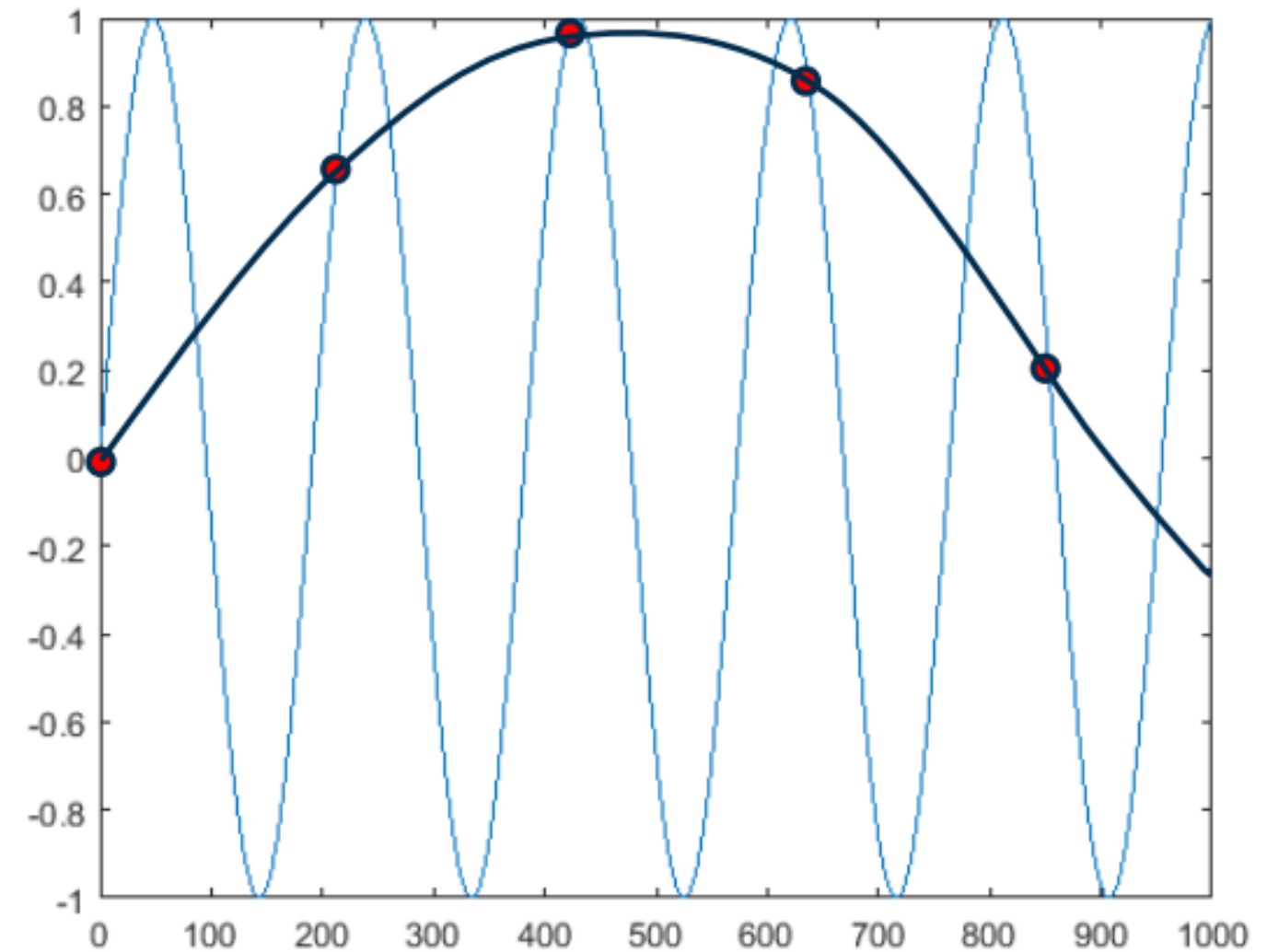




Aliasing

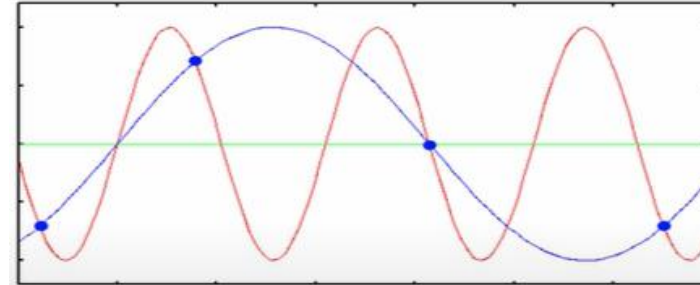
- Occurs when the (spatial) sampling rate is not high enough to capture the details in the image
- High frequencies are transformed to lower frequencies (i.e. aliases)
- To avoid aliasing the sampling rate must be at least two times the maximum frequency in the image (at least two samples per cycle)
- This minimum sampling rate is called the **Nyquist rate**.

Aliasing can be avoided by low-pass filtering the image before downsampling





Aliasing



Aliasing of a undersampled 1D sinusoidal signal



Aliasing of a undersampled 2D image



Aliasing problem of a downsampled image



Aliasing problem avoided by using band limiting Low Pass Filter.



Gaussian pre-filtering (low pass)



Gaussian 1/2



Gaussian 1/4



Gaussian 1/8





Gaussian pre-filtering (low pass)



Gaussian 1/2



Gaussian 1/4



Gaussian 1/8





Compared to downsampling without low-pass filtering...



1/2



1/4

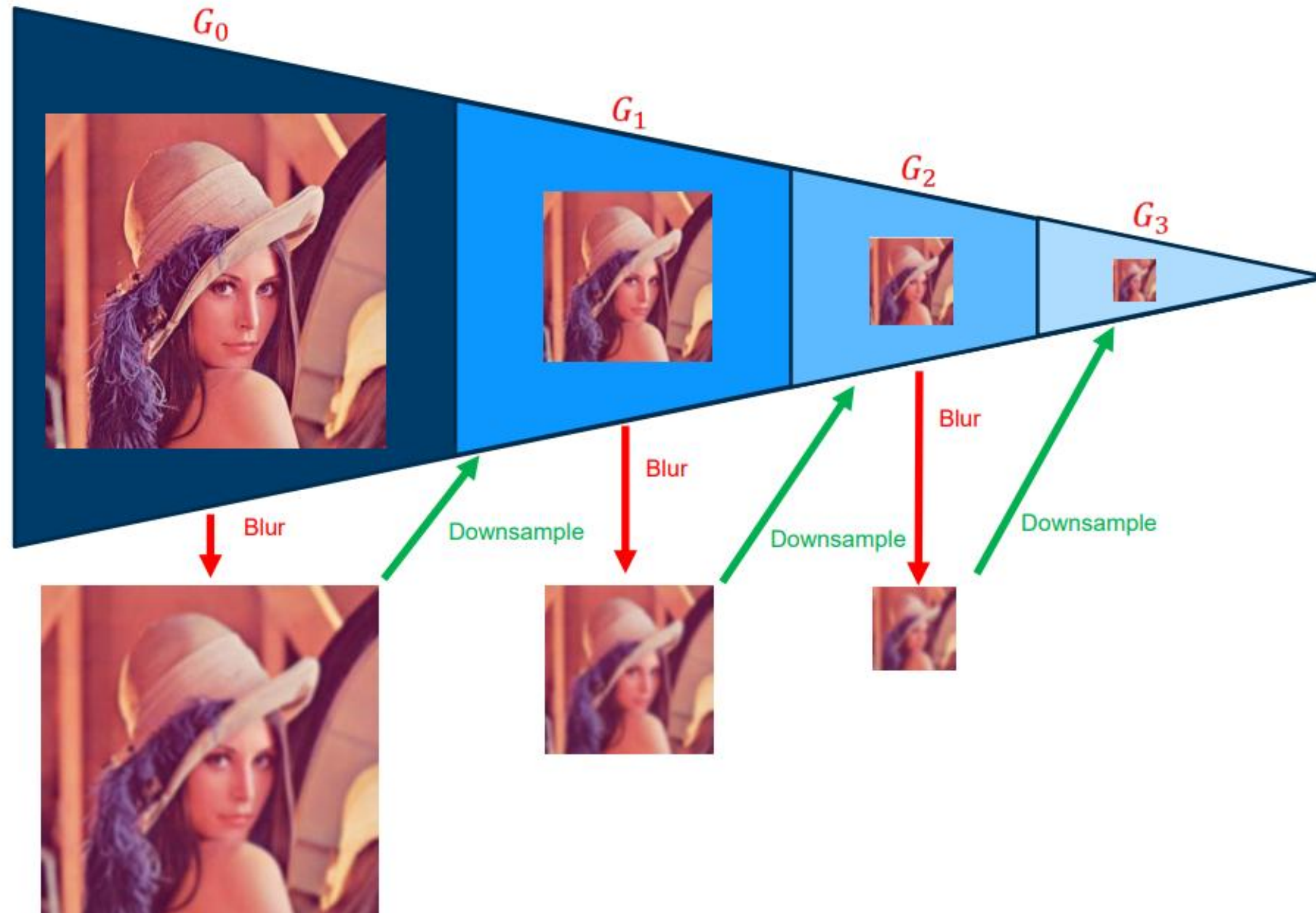


1/8

Conclusion: Low-pass filtering (i.e. smoothing with a Gaussian kernel) before subsampling the image!



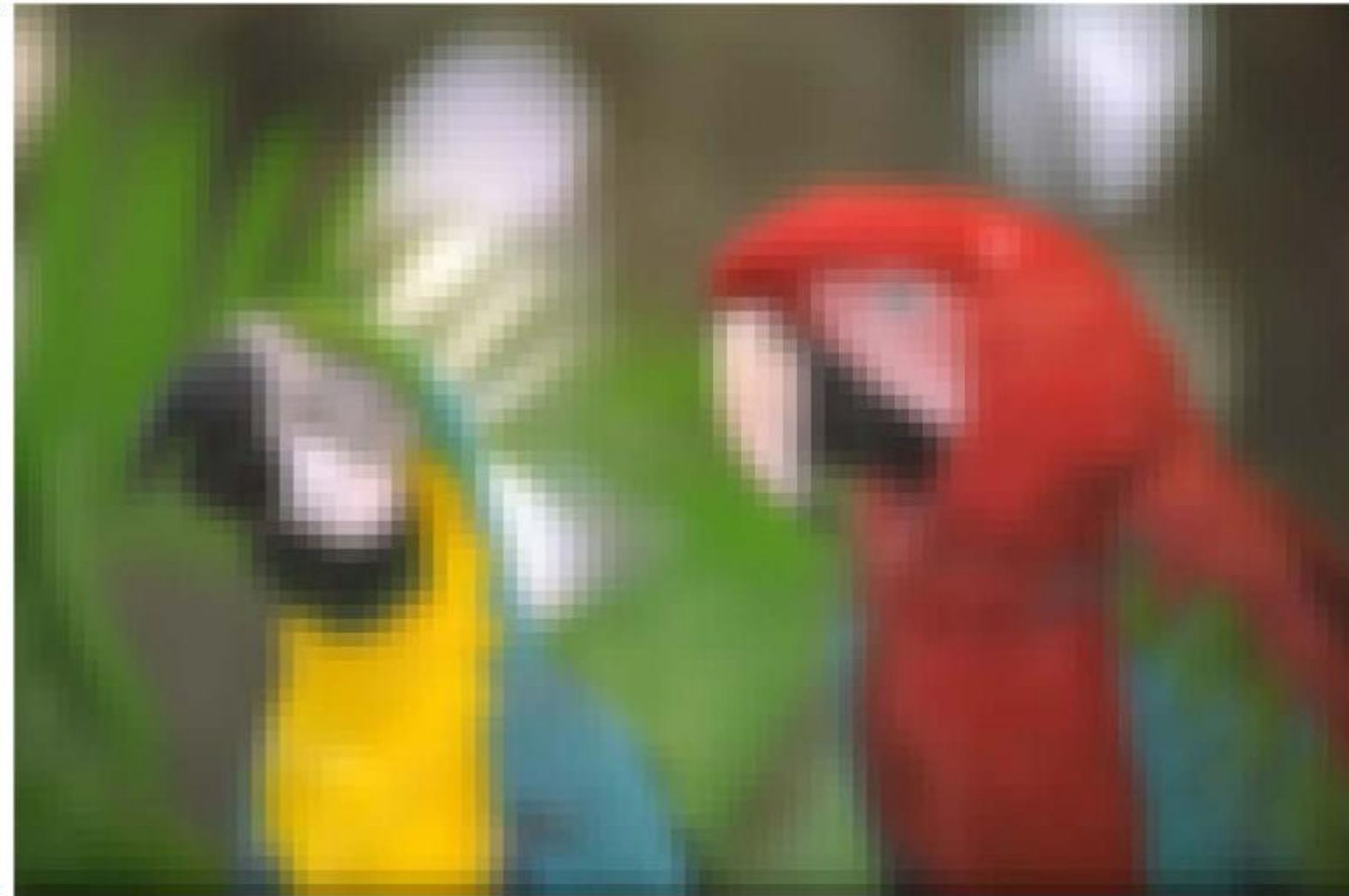
Gaussian Pyramid





Upsampling

10 x magnification



Nearest neighbor interpolation:

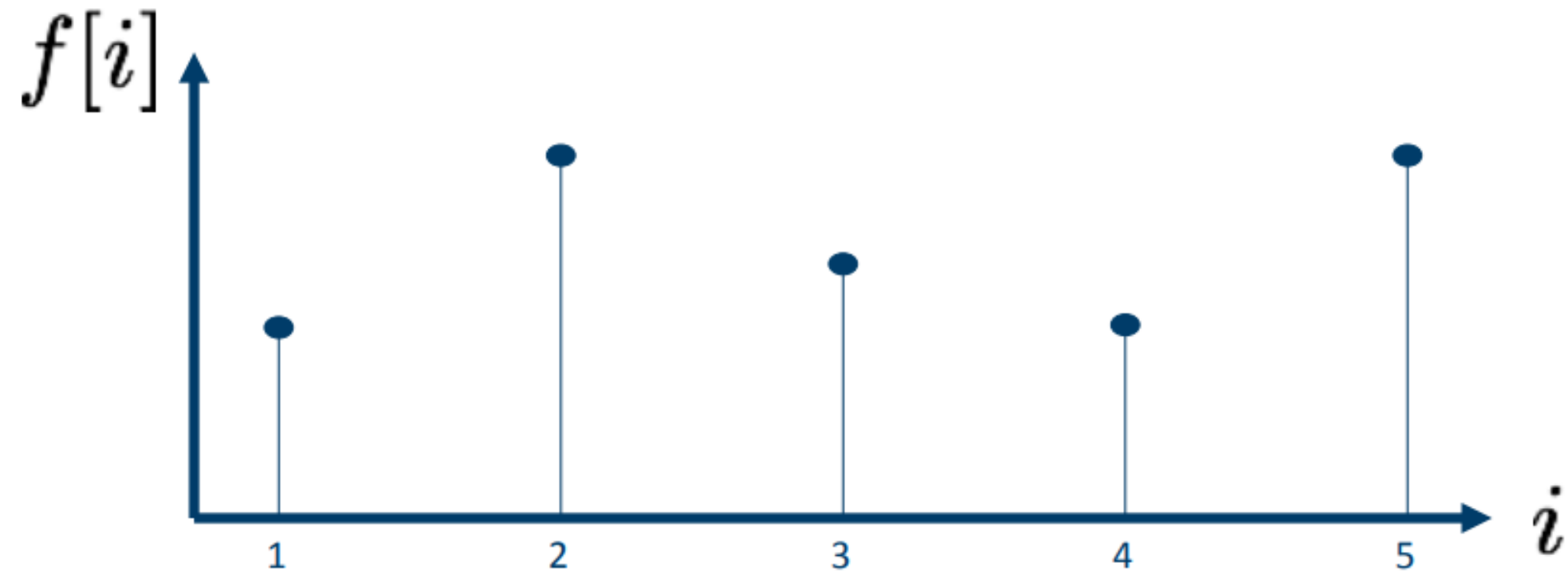
- Repeat each row and column 10 times
- Fast and simple approach.



Image interpolation

A digital image is a discrete point-sampling of a continuous function:

$$f[i, j] = \text{quantize}\{f(i\Delta x, j\Delta y)\} \quad \text{where } x = i\Delta x \text{ and } y = j\Delta y$$

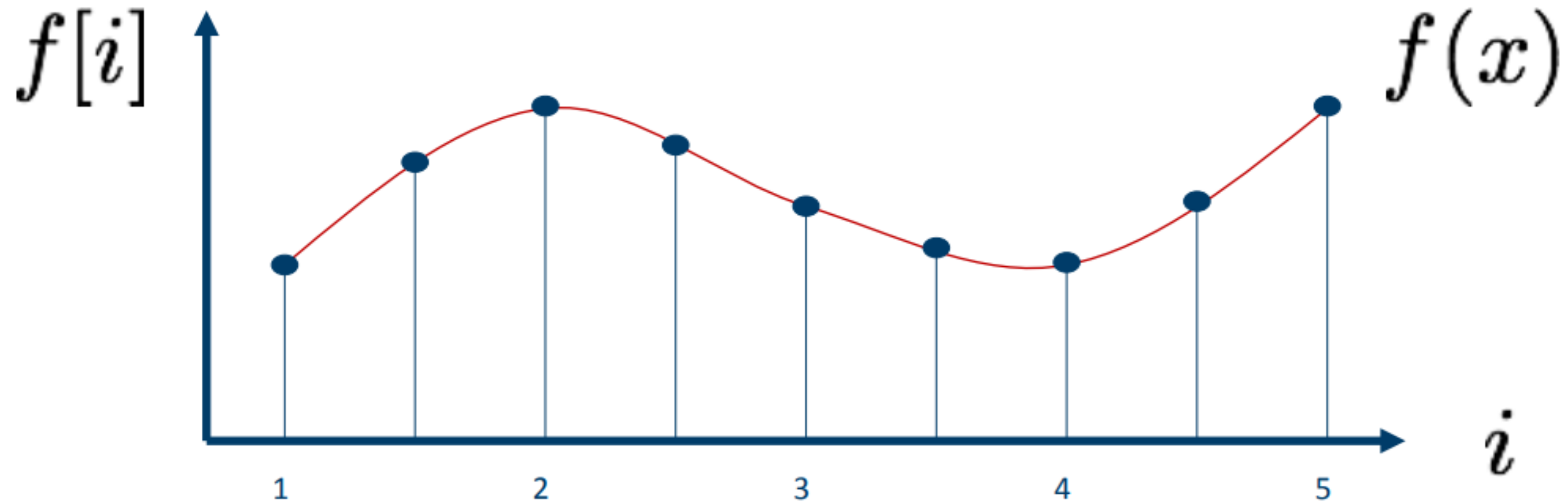


1D example

A new image could be generated, at any resolution and scale, if the original function could be reconstructed.



Interpolation by convolution

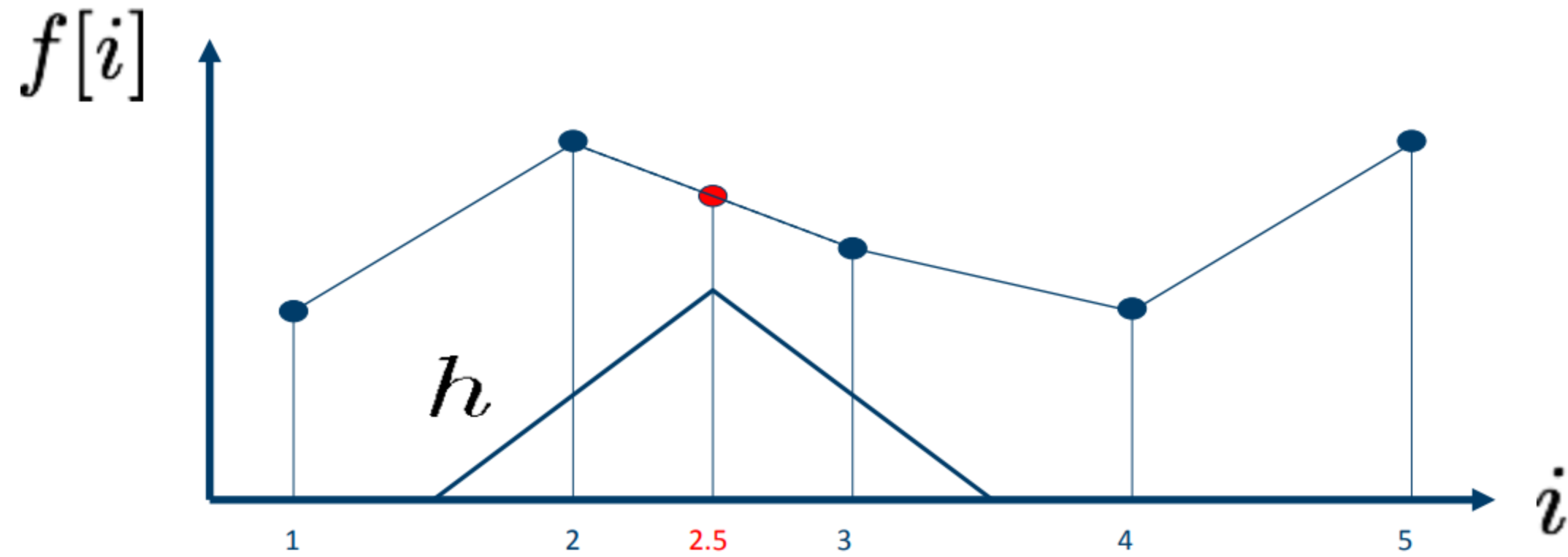


$$g[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k h[u, v] f[i - ru, j - rv]$$

$r =$ scale factor



Linear interpolation (bilinear for images)





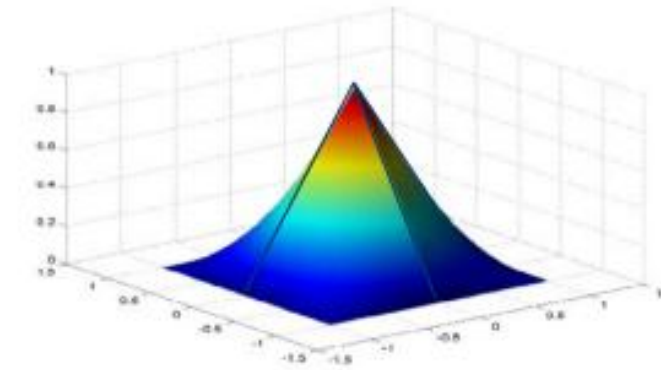
Some kernels for signal and image interpolation

Linear:

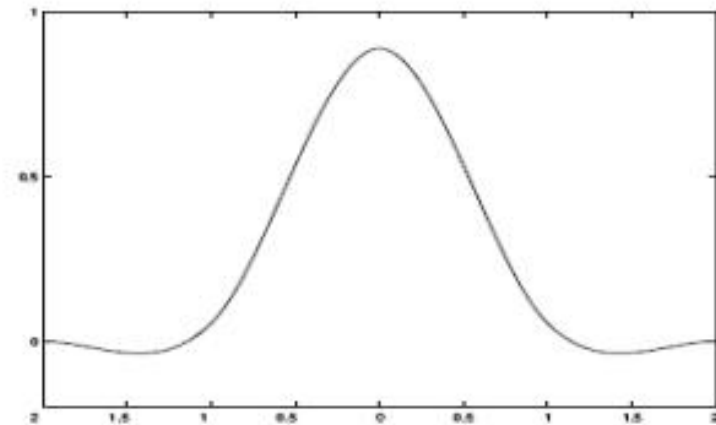


Bilinear:

$h(x, y)$



Bicubic (better choice for images):



Nearest neighbor:





Image interpolation - examples



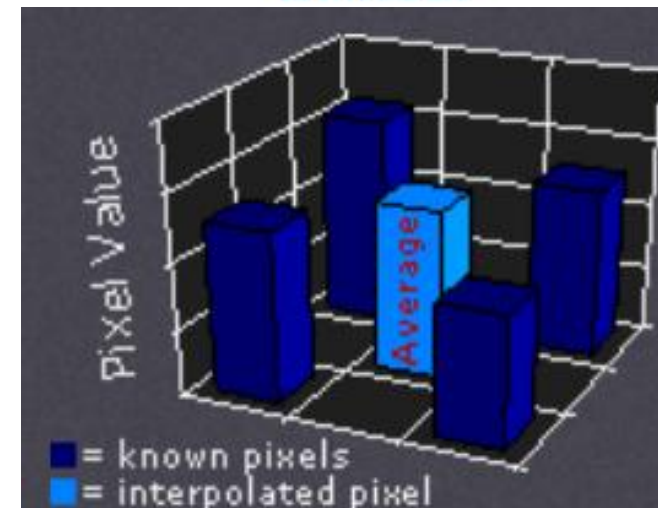
Nearest neighbor

It only considers one pixel — the closest one to the interpolated point.

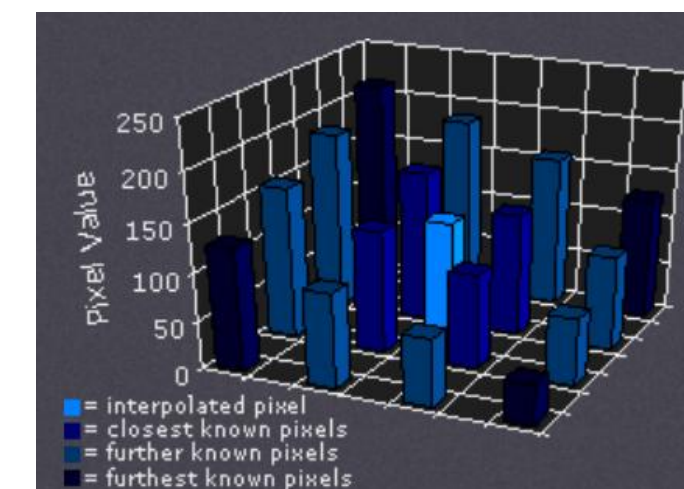
This has the effect of simply making each pixel bigger.



Bilinear



Bicubic





Application: Template Matching with Image Pyramids



Input: Image, Template

1. Match template at current scale
2. Downsample image
3. Repeat 1-2 until image is very small
4. Take responses above some threshold, perhaps with non-maxima suppression.



Thank
you!