

# Image Pyramids

# Finding Waldo

- Let's revisit the problem of finding Waldo
- This time he is on the road



image



template (filter)

# Finding Waldo

- He comes closer but our filter doesn't know that
- How can we find Waldo?



image



template (filter)

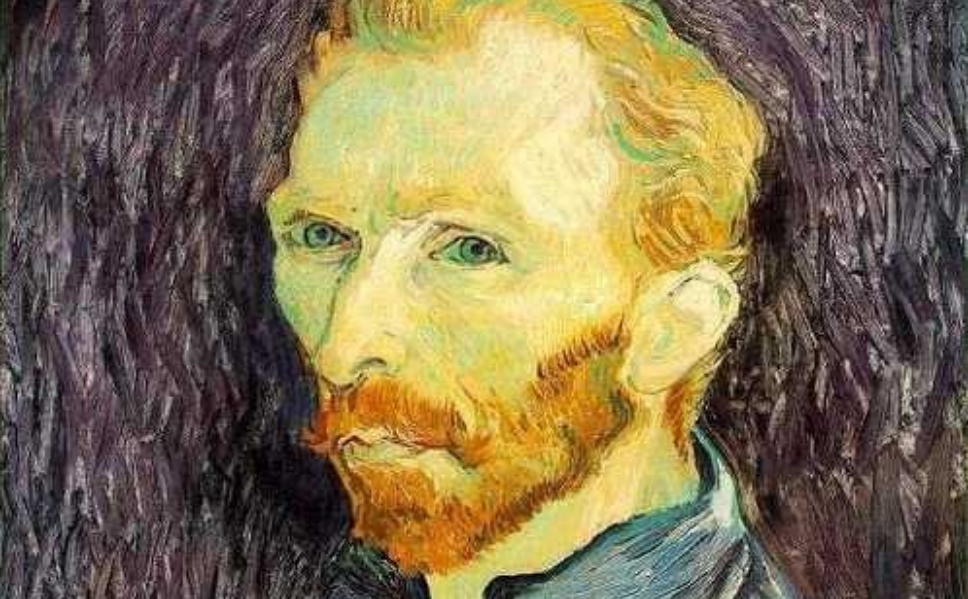
# Idea: Re-size Image

- Re-scale the image multiple times! Do correlation on every size!



template (filter)

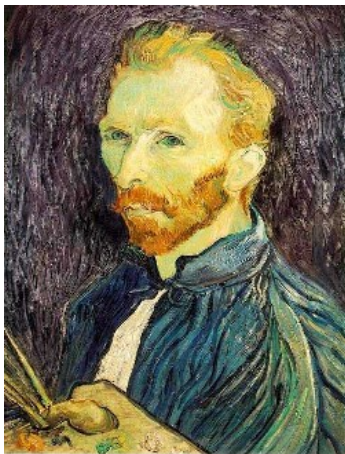




This image is huge. How can we make it smaller?

# Image Sub-Sampling

- **Idea:** Throw away every other row and column to create a  $1/2$  size image



1/4

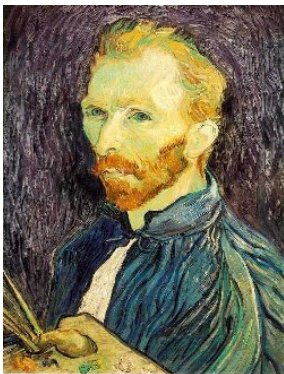


1/8

[Source: S. Seitz]

# Image Sub-Sampling

- Why does this look so cruffy?



$1/2$



$1/4$  (2x zoom)

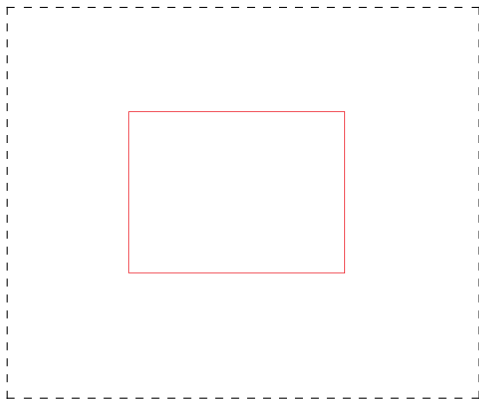


$1/8$  (4x zoom)

[Source: S. Seitz]

## Even worse for synthetic images

- I want to resize my image by factor 2
- And I take every other column and every other row (1st, 3rd, 5th, etc)



**Figure:** Dashed line denotes the border of the image (it's not part of the image)

# Even worse for synthetic images

- I want to resize my image by factor 2
- And I take every other column and every other row (1st, 3rd, 5th, etc)
- Where is the rectangle!



**Figure:** Dashed line denotes the border of the image (it's not part of the image)

# Even worse for synthetic images

- What's in the image?
- Now I want to resize my image by half in the width direction
- And I take every other column (1st, 3rd, 5th, etc)



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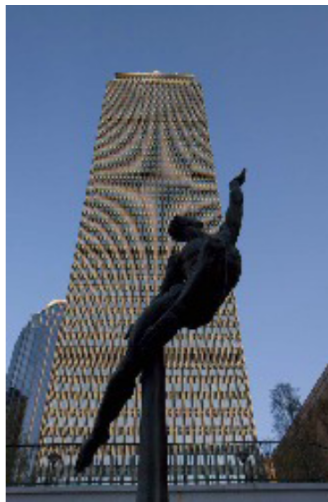
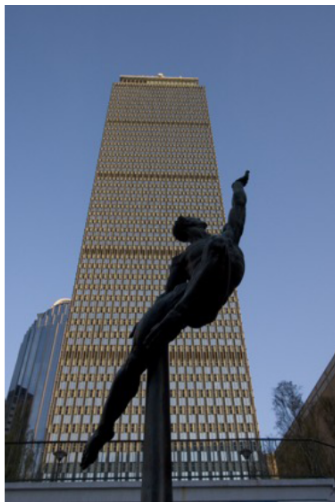
# Even worse for synthetic images

- What's in the image?
- Now I want to resize my image by half in the width direction
- And I take every other column (1st, 3rd, 5th, etc)
- Where is the chicken!





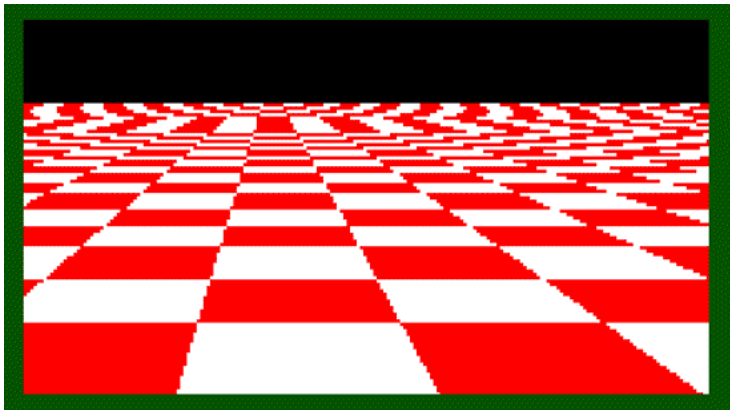
# Image Sub-Sampling



[Source: F. Durand]

# Even worse for synthetic images

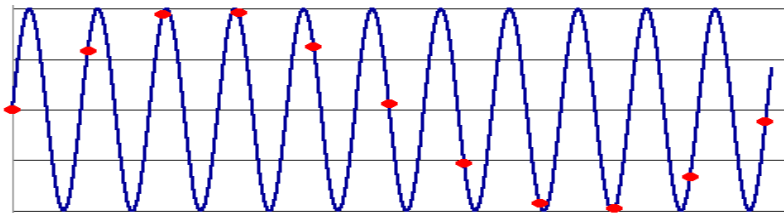
- What's happening?



[Source: L. Zhang]

# Aliasing

- Occurs when your sampling rate is not high enough to capture the amount of detail in your image

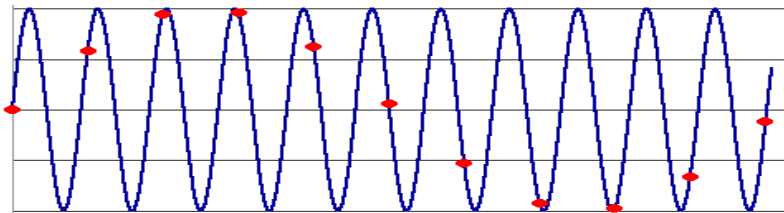


- To do sampling right, need to understand the structure of your signal/image

[Source: R. Urtasun]

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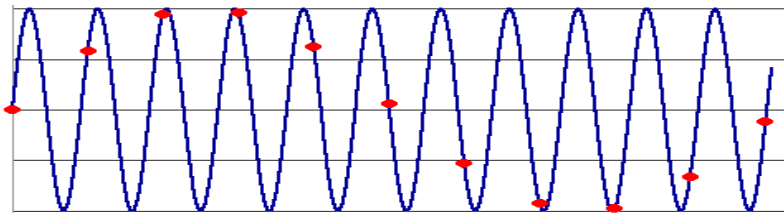


- To do sampling right, need to understand the structure of your signal/image
  - The minimum sampling rate is called the **Nyquist rate**

[Source: R. Urtasun]

# Aliasing

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- To do sampling right, need to understand the structure of your signal/image
- The minimum sampling rate is called the **Nyquist rate**

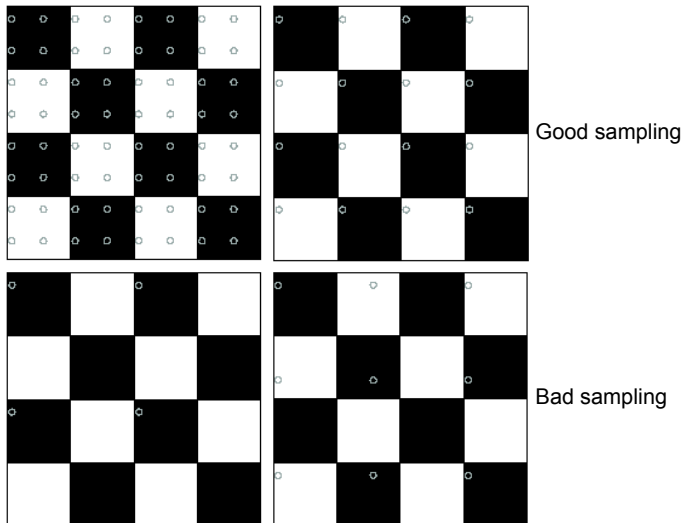
[Source: R. Urtasun]

# Mr. Nyquist

- Harry Nyquist says that one should look at the frequencies of the signal.
  - One should find the highest frequency (via Fourier Transform)
  - To sample properly you need to sample with at least twice that frequency
  - For those interested: [http://en.wikipedia.org/wiki/Nyquist%E2%80%99s\\_sampling\\_theorem](http://en.wikipedia.org/wiki/Nyquist%E2%80%99s_sampling_theorem)
- 
- He looks like a smart guy, we'll just believe him



## 2D example



[Source: N. Snavely]

# Going back to Downsampling ...

- When downsampling by a factor of two, the original image has frequencies that are too high
- High frequencies are caused by sharp edges
- How can we fix this?

[Adopted from: R. Urtasun]



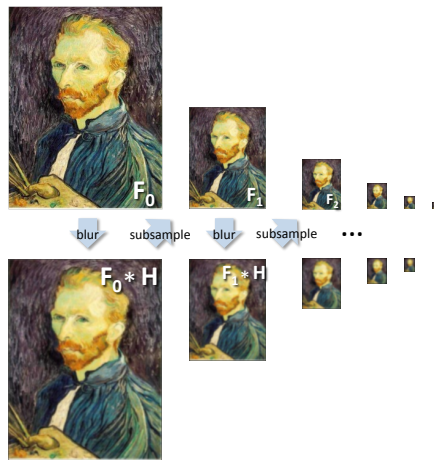
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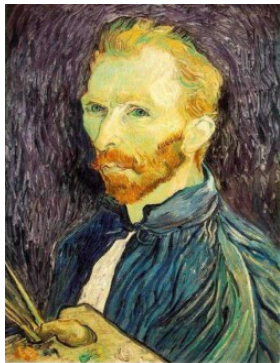
# Gaussian pre-filtering

- Solution: Blur the image via Gaussian, then subsample. Very simple!



[Source: N. Snavely]

# Subsampling with Gaussian pre-filtering



Gaussian  $1/2$



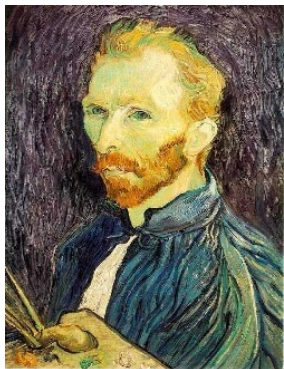
G  $1/4$



G  $1/8$

[Source: S. Seitz]

# Compare with ...



1/2



1/4 (2x zoom)

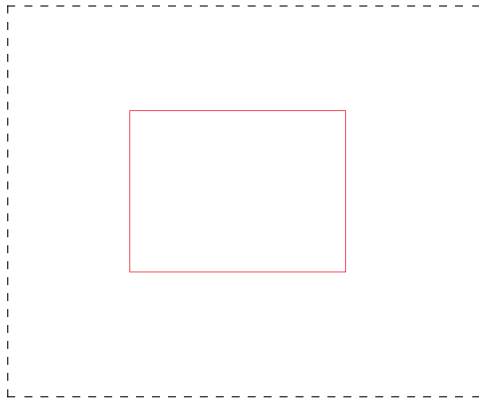


1/8 (4x zoom)

[Source: S. Seitz]

# Where is the Rectangle?

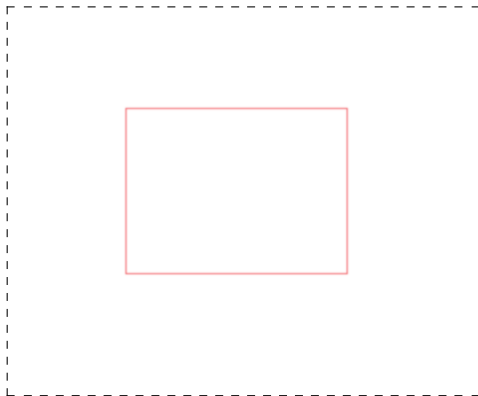
- My image



**Figure:** Dashed line denotes the border of the image (it's not part of the image)

# Where is the Rectangle?

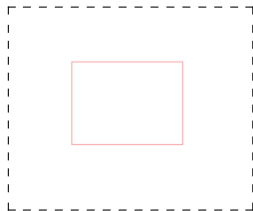
- My image
- Let's blur



**Figure:** Dashed line denotes the border of the image (it's not part of the image)

# Where is the Rectangle?

- My image
- Let's blur
- And now take every other row and column



**Figure:** Dashed line denotes the border of the image (it's not part of the image)

# Where is the Chicken?

- My image





# Where is the Chicken?

- My image
- Let's blur



# Where is the Chicken?

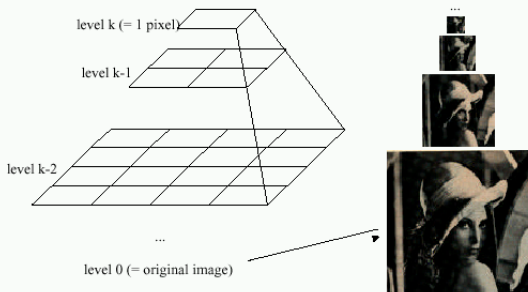
- My image
- Let's blur
- And now take every other column



# Gaussian Pyramids [Burt and Adelson, 1983]

- A sequence of images created with Gaussian blurring and downsampling is called a **Gaussian Pyramid**
- In computer graphics, a *mip map* [Williams, 1983]

Idea: Represent  $N \times N$  image as a “pyramid” of  $1 \times 1, 2 \times 2, 4 \times 4, \dots, 2^k \times 2^k$  images (assuming  $N=2^k$ )



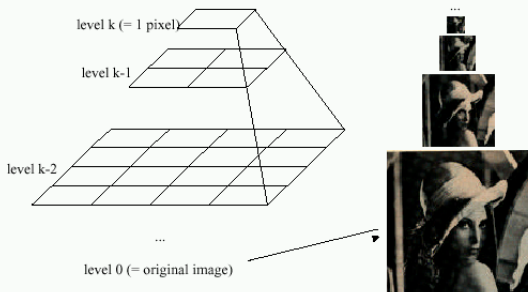
- How much space does a Gaussian pyramid take compared to original image?

[Source: S. Seitz]

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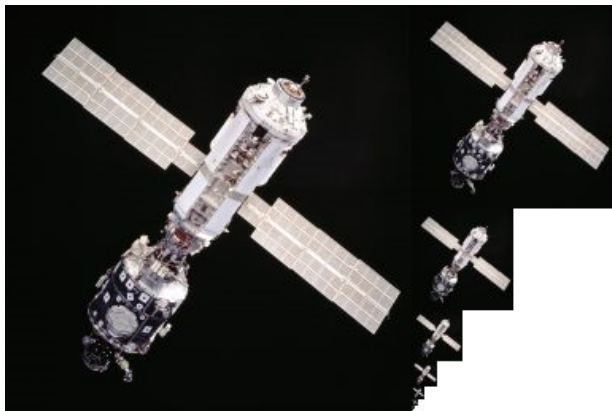
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- How much space does a Gaussian pyramid take compared to original image?

[Source: S. Seitz]

# Example of Gaussian Pyramid



[Source: N. Snavely]

# Image Up-Sampling

- This image is too small, how can we make it 10 times as big?



[Source: N. Snavely, R. Urtasun]

# Image Up-Sampling

- This image is too small, how can we make it 10 times as big?

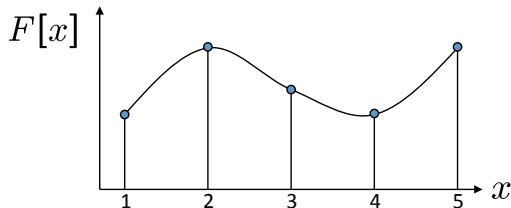


- Simplest approach: repeat each row and column 10 times



[Source: N. Snavely, R. Urtasun]

# Interpolation



$d = 1$  in this example

Recall how a digital image is formed

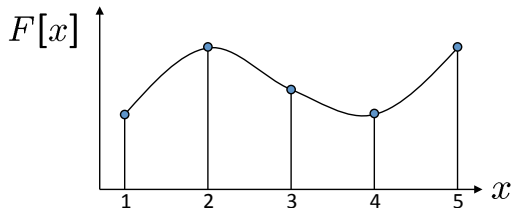
$$F[x, y] = \text{quantize}\{f(xd, yd)\}$$

- It is a discrete point-sampling of a continuous function
- If we could somehow reconstruct the original function, any new image could be generated, at any resolution and scale

[Source: N. Snavely, S. Seitz]



# Interpolation



$d = 1$  in this example

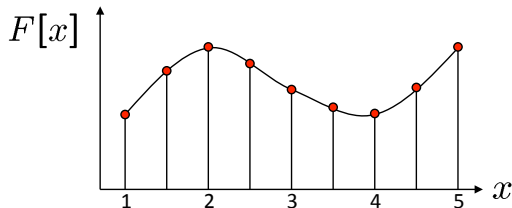
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[Source: N. Snavely, S. Seitz]

# Interpolation



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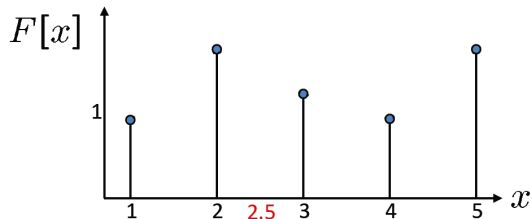
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[Source: N. Snavely, S. Seitz]

# Interpolation

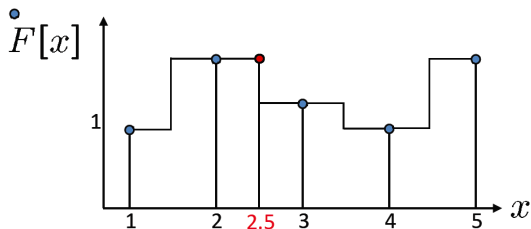


$d = 1$  in this  
example

What if we don't know  $f$ ?

[Source: N. Snavely, S. Seitz]

# Interpolation



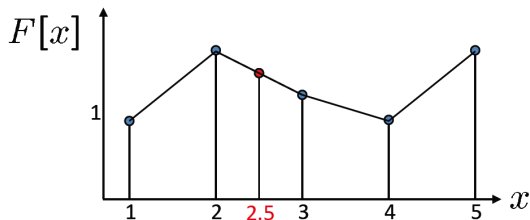
$d = 1$  in this example

What if we don't know  $f$ ?

- Guess an approximation: for example nearest-neighbor

[Source: N. Snavely, S. Seitz]

# Interpolation



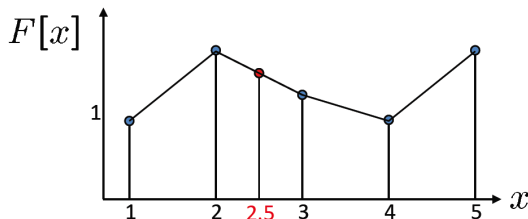
$d = 1$  in this example

What if we don't know  $f$ ?

- Guess an approximation: for example nearest-neighbor
- Guess an approximation: for example linear

[Source: N. Snavely, S. Seitz]

# Interpolation



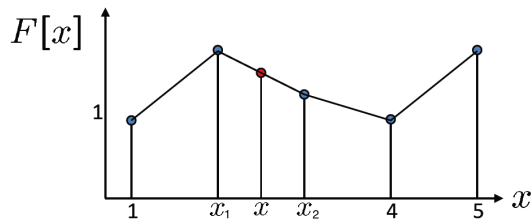
$d = 1$  in this example

What if we don't know  $f$ ?

- Guess an approximation: for example nearest-neighbor
- Guess an approximation: for example linear
- More complex approximations: cubic, B-splines

[Source: N. Snavely, S. Seitz]

# Linear Interpolation



$d = 1$  in this example

- Linear interpolation:

$$G(x) = \frac{x_2 - x}{x_2 - x_1} F(x_1) + \frac{x - x_1}{x_2 - x_1} F(x_2)$$

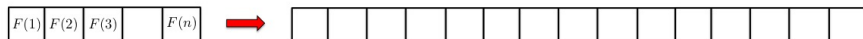
# Interpolation: 1D Example

$F(1)$	$F(2)$			$F(n)$
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- Let's make this signal triple length



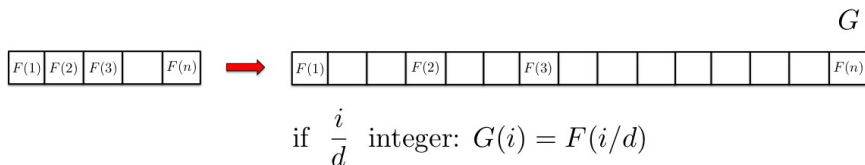
# Interpolation: 1D Example



Make a vector  $G$  with  $d$  times the size of  $F$

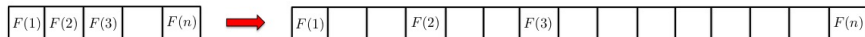
- Let's make this signal triple length ( $d = 3$ )

# Interpolation: 1D Example



- Let's make this signal triple length ( $d = 3$ )
- If  $i/d$  is an integer, just copy from the signal

# Interpolation: 1D Example



if  $\frac{i}{d}$  integer:  $G(i) = F(i/d)$

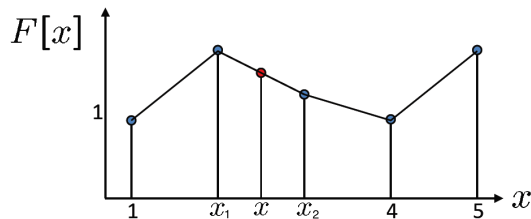
otherwise:  $G(i) = \frac{x_2 - x}{x_2 - x_1} F(x_1) + \frac{x - x_1}{x_2 - x_1} F(x_2)$

where

$$x = i/d$$
$$x_1 = \lfloor i/d \rfloor$$
$$x_2 = \lceil i/d \rceil$$

- Let's make this signal triple length ( $d = 3$ )
- If  $i/d$  is an integer, just copy from the signal
- Otherwise use the interpolation formula

# Linear Interpolation via Convolution



$d = 1$  in this example

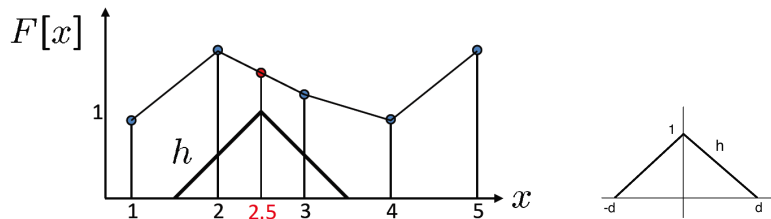
- Linear interpolation:

$$G(x) = \frac{x_2 - x}{x_2 - x_1} F(x_1) + \frac{x - x_1}{x_2 - x_1} F(x_2)$$

- With  $t = x - x_1$  and  $d = x_2 - x_1$  we can get:

$$G(x) = \frac{d - t}{d} F(x - t) + \frac{t}{d} F(x + d - t)$$

# Linear Interpolation via Convolution



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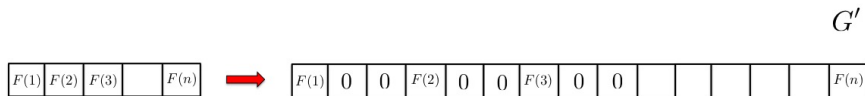
( Kind of looks like convolution:  $G(x) = \sum_t h(t) F(x - t)$  ) )

# Interpolation via Convolution: 1D Example

$F(1)$	$F(2)$			$F(n)$
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- Let's make this signal triple length

# Interpolation via Convolution: 1D Example

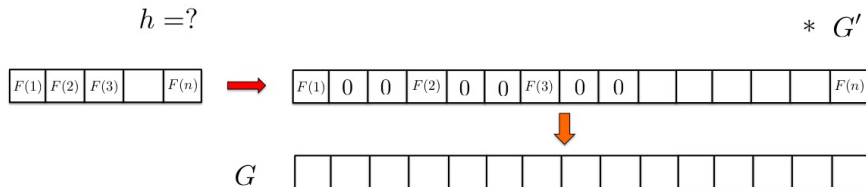


if  $\frac{i}{d}$  integer:  $G'(i) = F(i/d)$

otherwise: 0

- Let's make this signal triple length ( $d = 3$ )

# Interpolation via Convolution: 1D Example

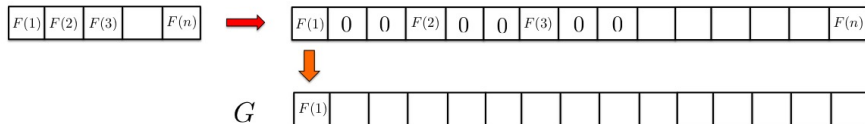


- Let's make this signal triple length ( $d = 3$ )
- What should be my “reconstruction” filter  $h$  (such that  $G = h * G'$ )?



# Interpolation via Convolution: 1D Example

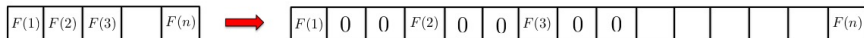
$$h = \left[0, \frac{1}{3}, \frac{2}{3}, 1, \frac{2}{3}, \frac{1}{3}, 0\right] \quad * \quad G'$$



- Let's make this signal triple length ( $d = 3$ )
- What should be my “reconstruction” filter  $h$  (such that  $G = h * G'$ )?
- $h = [0, \frac{1}{d}, \dots, \frac{d-1}{d}, 1, \frac{d-1}{d}, \dots, \frac{1}{d}, 0]$ , where  $d$  my upsampling factor

# Interpolation via Convolution: 1D Example

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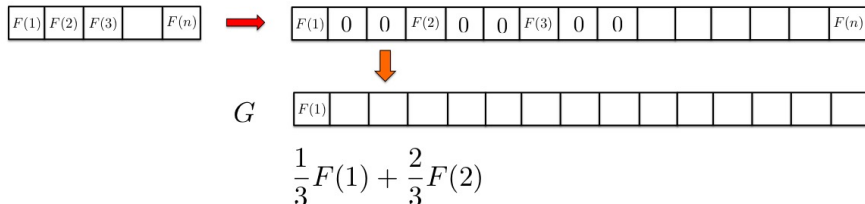


$$\frac{2}{3}F(1) + \frac{1}{3}F(2)$$

- Let's make this signal triple length ( $d = 3$ )
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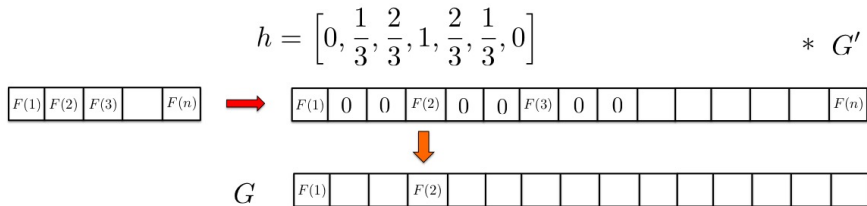
# Interpolation via Convolution: 1D Example

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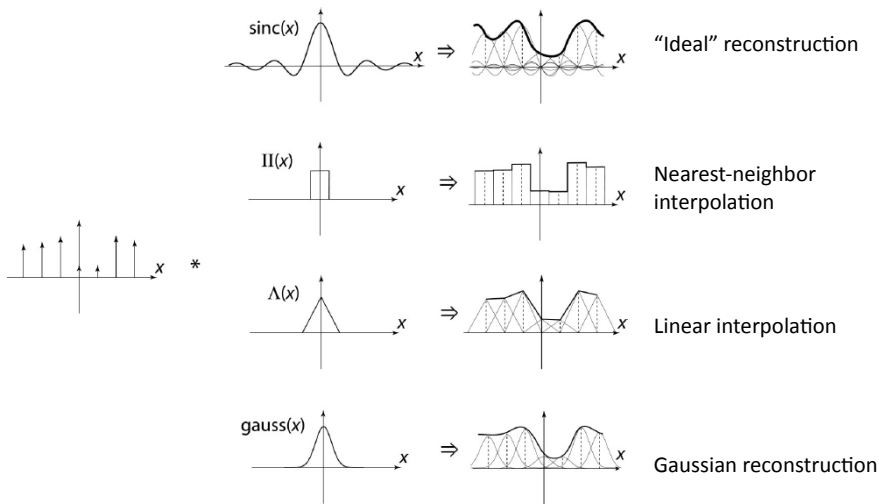
- Let's make this signal triple length ( $d = 3$ )
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# Interpolation via Convolution: 1D Example



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# Interpolation via Convolution (1D)



Source: B. Curless

# Image Interpolation (2D)

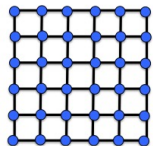
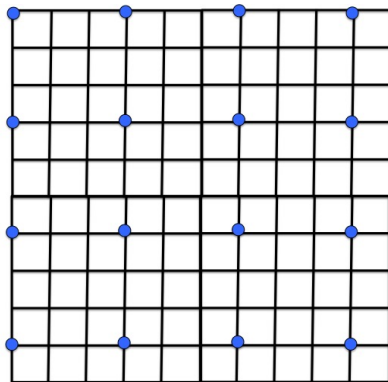
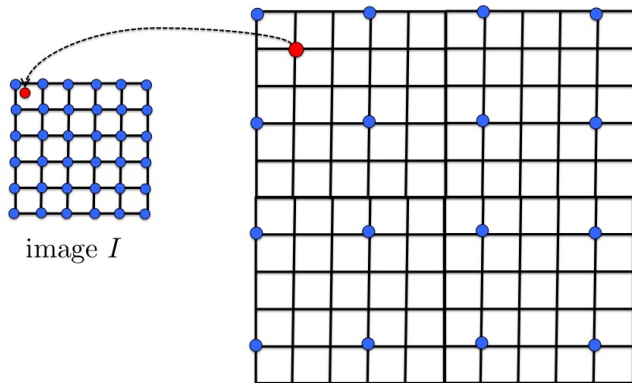


image  $I$



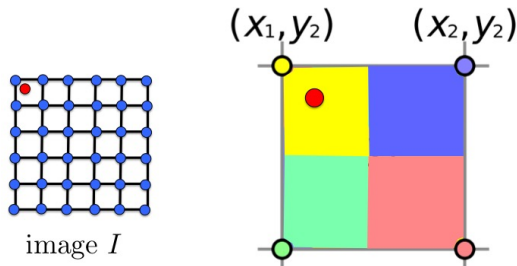
- Let's make this image triple size
- Copy image in every third pixel. What about the remaining pixels in  $G$ ?

# Image Interpolation (2D)



- Let's make this image triple size
- Copy image in every third pixel. What about the remaining pixels in  $G$ ?
- How shall we compute this value?

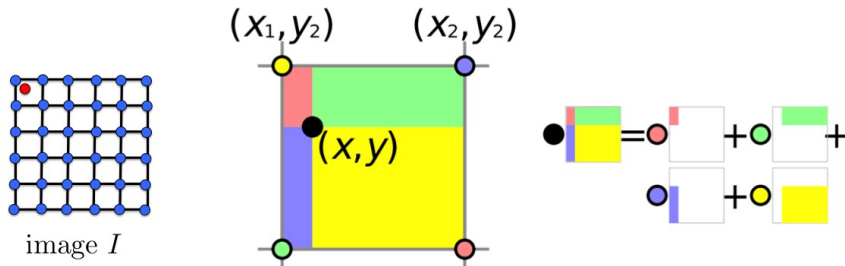
# Image Interpolation (2D)



- Let's make this image triple size
- Copy image in every third pixel. What about the remaining pixels in  $G$ ?
- One possible way: nearest neighbor interpolation



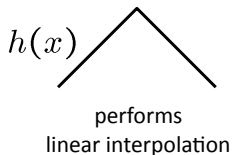
# Image Interpolation (2D)



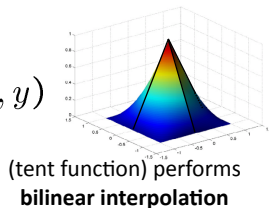
- Let's make this image triple size
- Copy image in every third pixel. What about the remaining pixels in  $G$ ?
- Better: bilinear interpolation (check out details:  
[http://en.wikipedia.org/wiki/Bilinear\\_interpolation](http://en.wikipedia.org/wiki/Bilinear_interpolation))

# Reconstruction Filters

- What does the 2D version of this hat function look like?

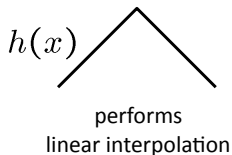


$$h(x, y)$$

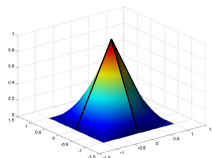


# Reconstruction Filters

- What does the 2D version of this hat function look like?



$h(x, y)$

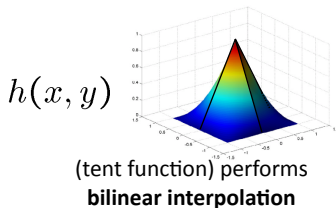
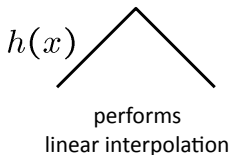


(tent function) performs  
**bilinear interpolation**

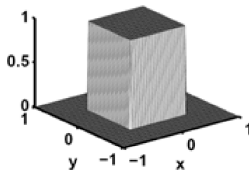
- And filter for nearest neighbor interpolation?

# Reconstruction Filters

- What does the 2D version of this hat function look like?

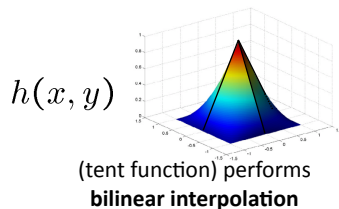
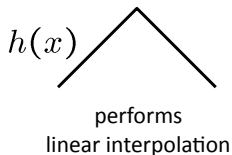


- And filter for nearest neighbor interpolation?



# Reconstruction Filters

- What does the 2D version of this hat function look like?



- Better filters give better resampled images: Bicubic is a common choice

# Image Interpolation via Convolution (2D)

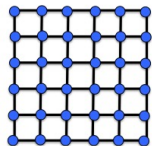
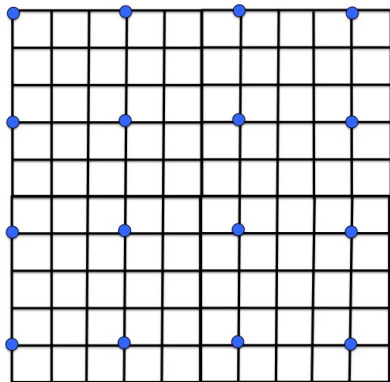
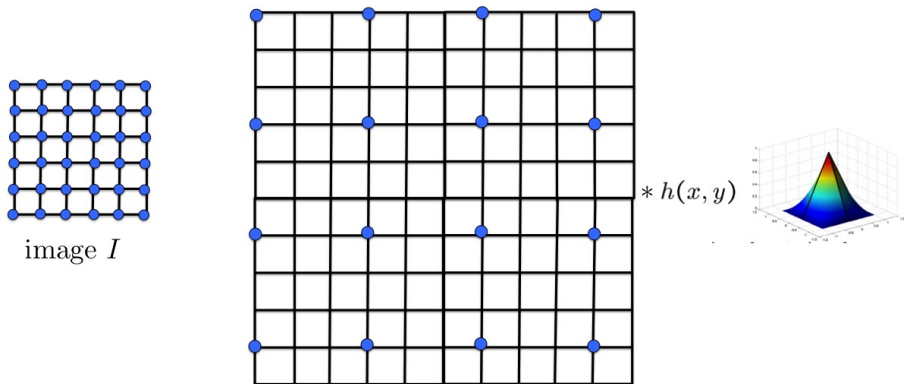


image  $I$



- Let's make this image triple size: copy image values in every third pixel, place zeros everywhere else

# Image Interpolation via Convolution (2D)



- Let's make this image triple size: copy image values in every third pixel, place zeros everywhere else
- Convolution with a reconstruction filter (e.g., bilinear) and you get the interpolated image

# Image Interpolation

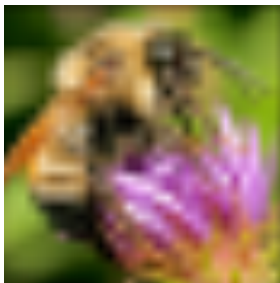
Original image



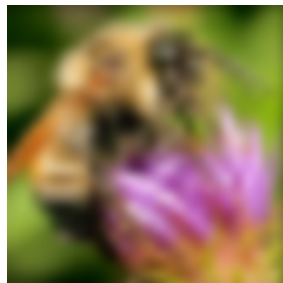
Interpolation results



Nearest-neighbor interpolation



Bilinear interpolation



Bicubic interpolation

[Source: N. Snavely]



# Deep Learning for Image Superresolution

- You can use DL to increase image resolution!



<https://www.youtube.com/watch?v=pZXFxtfd-Ak>

Pic credit:

<https://medium.com/beyondminds/an-introduction-to-super-resolution-using-deep-learning-f60aff9a499d>

# Deep Learning Super Sampling (DLSS)

- You can use DL to increase image resolution!



<https://news.developer.nvidia.com/dlss-three-things-you-need-to-know/>

# Summary – Stuff You Should Know

- To down-scale an image: blur it with a small Gaussian (e.g.,  $\sigma = 1.4$ ) and downsample
- To up-scale an image: interpolation (nearest neighbor, bilinear, bicubic, etc)
- Gaussian pyramid: Blur with Gaussian filter, downsample result by factor 2, blur it with the Gaussian, downsample by 2...

## Matlab functions:

- `IMRESIZE(IMAGE, SCALE, METHOD)`: Matlab's function for resizing the image, where `METHOD`= “nearest”, “bilinear”, “bicubic” (works for downsampling and upsampling)
- `SKIMAGE.TRANSFORM.RESIZE` and `SKIMAGE.TRANSFORM.RESCALE`: Python's function for resizing, where `ORDER` is in the range 0-5 with the following semantics: 0: Nearest-neighbor 1: Bi-linear (default) 2: Bi-quadratic 3: Bi-cubic