# Image Pyramids

# Finding Waldo

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





image

# Finding Waldo

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





#### image

#### Idea: Re-size Image

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









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

Sanja Fidler

CSC420: Intro to Image Understanding

# 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 crufty?



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)



#### 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!

#### • 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)



#### 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)



### 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]

• 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]

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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% 93Shannon\_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]



1/2

1/4 (2x zoom)

1/8 (4x zoom)

[Source: S. Seitz]

### Where is the Rectangle?

My image



# Where is the Rectangle?

- My image
- Let's blur



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



#### 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 NxN image as a "pyramid" of 1x1, 2x2, 4x4,..., 2<sup>k</sup>x2<sup>k</sup> images (assuming N=2<sup>k</sup>)



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



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[Source: S. Seitz] Sanja Fidler CSC420: Intro to Image Understanding 21 / 35

### 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]



Recall how a digital image is formed

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

#### • 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



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d = 1 in this example

What if we don't know f?

#### Interpolation



What if we don't know f?

• Guess an approximation: for example nearest-neighbor



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]
```



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

### Linear Interpolation



• Linear interpolation:

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

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



Make a vector G with d times the size of F

#### • Let's make this signal triple length (d = 3)

G



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

G



- 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



• 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:

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

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

F(1) F(2)	F(n)
-----------	------

#### • Let's make this signal triple length

G'



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



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

$$h = \begin{bmatrix} 0, \frac{1}{3}, \frac{2}{3}, 1, \frac{2}{3}, \frac{1}{3}, 0 \end{bmatrix} \\ & * G'$$

$$F(1) \xrightarrow{F(2)} F(3) \xrightarrow{F(n)} \xrightarrow{F(n)} \xrightarrow{F(1)} 0 \xrightarrow{F(2)} 0 \xrightarrow{F(3)} 0 \xrightarrow{F(3)} 0 \xrightarrow{F(n)} \xrightarrow{F(n)} \xrightarrow{F(1)} 0 \xrightarrow$$

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# Interpolation via Convolution (1D)







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



- 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?



- 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



• 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)

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

h(x)

performs linear interpolation



(tent function) performs bilinear interpolation

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

performs linear interpolation



bilinear interpolation

• And filter for nearest neighbor interpolation?

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



performs linear interpolation



bilinear interpolation

• And filter for nearest neighbor interpolation?



• 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

# 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]

# 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