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?





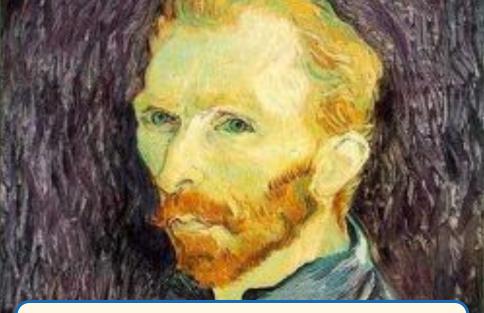
image

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?

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CSC420: Intro to Image Understanding

Image Sub-Sampling

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





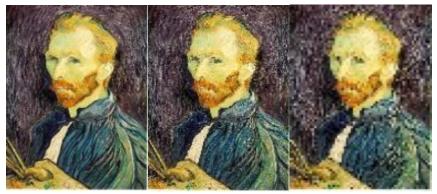
1/4

[Source: S. Seitz]

1/8

Image Sub-Sampling

• Why does this look so crufty?



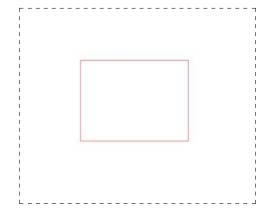
1/2

1/4 (2x zoom)

1/8 (4x zoom)

[Source: S. Seitz]

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



- 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)



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- 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!



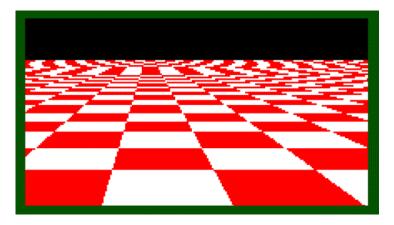
Image Sub-Sampling



[Source: F. Durand]

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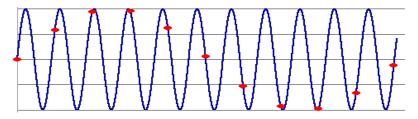
• What's happening?



[Source: L. Zhang]

Aliasing

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

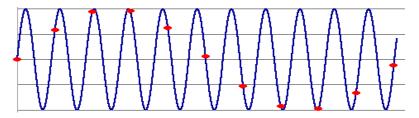


• To do sampling right, need to understand the structure of your signal/image

[Source: R. Urtasun]

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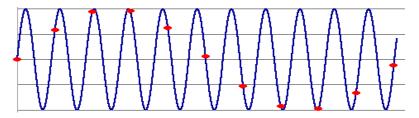


- To do sampling right, need to understand the structure of your signal/image
- The minimum sampling rate is called the Nyquist rate

[Source: R. Urtasun]

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[Source: R. Urtasun]

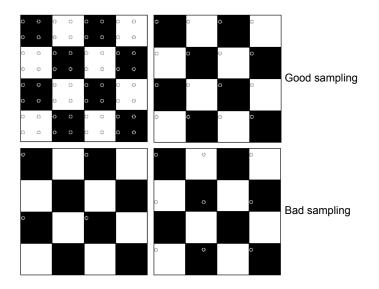
Mr. Nyquist

- 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]

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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]

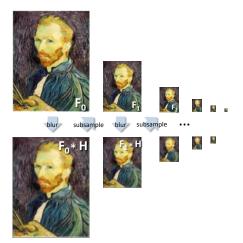
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[Adopted from: R. Urtasun]

Gaussian pre-filtering

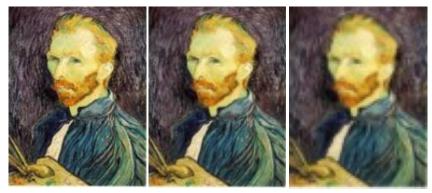
• Solution: Blur the image via Gaussian, then subsample. Very simple!



[Source: N. Snavely]

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Subsampling with Gaussian pre-filtering



Gaussian 1/2

G 1/4

G 1/8

[Source: S. Seitz]

Compare with ...



1/2

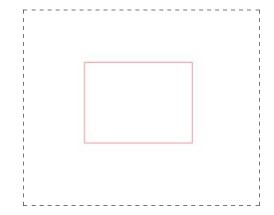
1/4 (2x zoom)

1/8 (4x zoom)

[Source: S. Seitz]

Where is the Rectangle?

• My image



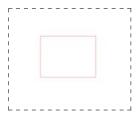
Where is the Rectangle?

My imageLet's blur



Where is the Rectangle?

- 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



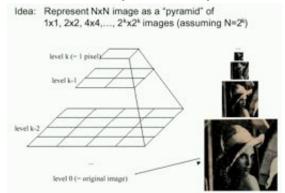
Where is the Chicken?

- 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]



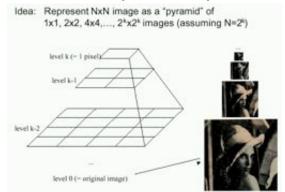
• How much space does a Gaussian pyramid take compared to original image?

[Source: S. Seitz]

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Gaussian Pyramids [Burt and Adelson, 1983]

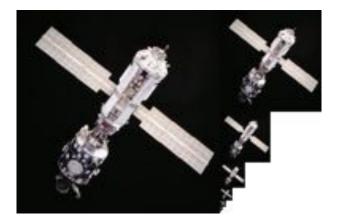
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• How much space does a Gaussian pyramid take compared to original image? [Source: S. Seitz]

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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

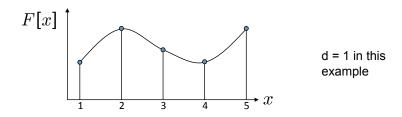
• This image is too small, how can we make it 10 times as big?



• Simplest approach: repeat each row and column 10 times



[Source: N. Snavely, R. Urtasun]



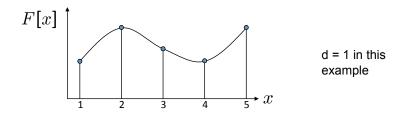
Recall how a digital image is formed

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

• 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

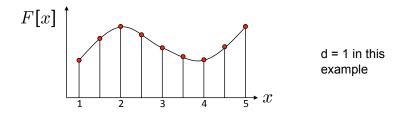
[Source: N. Snavely, S. Seitz]



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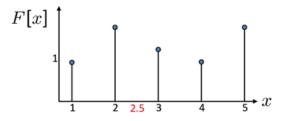
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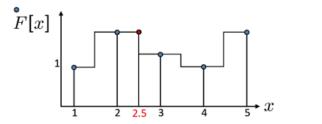
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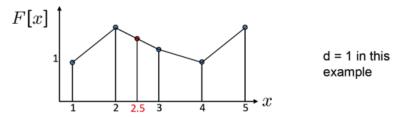


d = 1 in this example



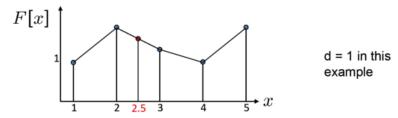
d = 1 in this example

• Guess an approximation: for example nearest-neighbor



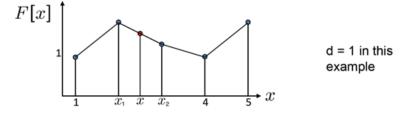
- Guess an approximation: for example nearest-neighbor
- Guess an approximation: for example linear

```
[Source: N. Snavely, S. Seitz]
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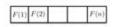
- Guess an approximation: for example nearest-neighbor
- Guess an approximation: for example linear
- More complex approximations: cubic, B-splines

Linear Interpolation

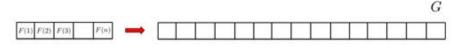


• Linear interpolation:

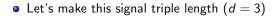
$$G(x) = \frac{x_2 - x}{x_2 - x_1}F(x_1) + \frac{x - x_1}{x_2 - x_1}F(x_2)$$



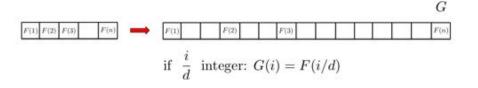
• Let's make this signal triple length



Make a vector G with d times the size of F



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

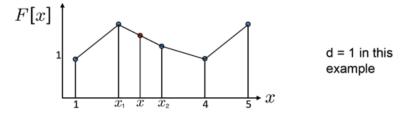
 $F(1) \xrightarrow{F(2)} F(3) \xrightarrow{F(n)} \longrightarrow F(1) \xrightarrow{F(2)} F(2) \xrightarrow{F(3)} F(3) \xrightarrow{F(n)} F(n)$ if $\frac{i}{d}$ integer: G(i) = F(i/d)otherwise: $G(i) = \frac{x_2 - x}{x_2 - x_1} F(x_1) + \frac{x - x_1}{x_2 - x_1} F(x_2)$ x = i/dwhere $x_1 = \lfloor i/d \rfloor$ $x_2 = \lceil i/d \rceil$

- 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

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G

Linear Interpolation via Convolution



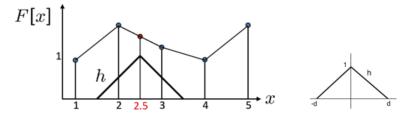
• Linear interpolation:

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

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

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

Linear Interpolation via Convolution



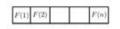
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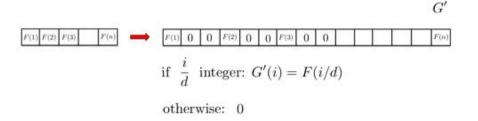
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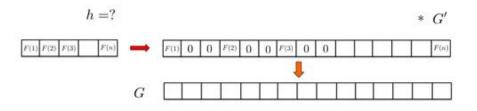
(Kind of looks like convolution: $G(x) = \sum_t h(t)F(x-t)$))



• Let's make this signal triple length



• 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')?

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

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

$$G \xrightarrow{F^{(1)}} F^{(1)} F^{(1)} F^{(1)} F^{(1)} F^{(2)} F^{(1)} F^{(1)} F^{(2)} F^{(2)} F^{(2)} F^{(2)} F^{(2)} F^{(2)} F^{(3)} F^$$

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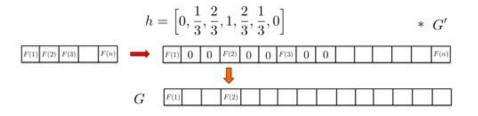
$$F^{(1)}F^{(2)}F^{(3)} \xrightarrow{F^{(n)}} \xrightarrow{F^{(n)}} F^{(n)} \xrightarrow{F^{(1)} 0 \ 0 \ F^{(2)} \ 0 \ 0 \ F^{(3)} \ 0 \ 0 \ F^{(3)} \ 0 \ 0 \ F^{(n)} \xrightarrow{F^{(n)}} \xrightarrow{F^{(n)}} \xrightarrow{F^{(n)}} \xrightarrow{\frac{1}{3}} G \xrightarrow{F^{(1)} 1} \xrightarrow{F^{(1)}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

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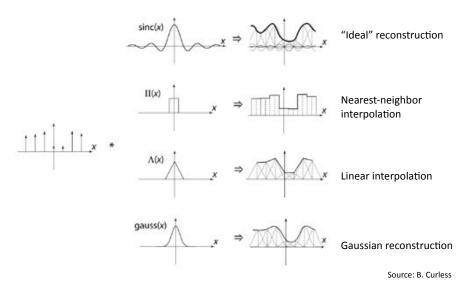
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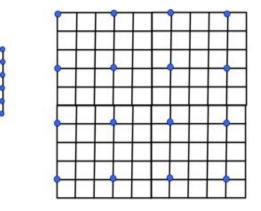
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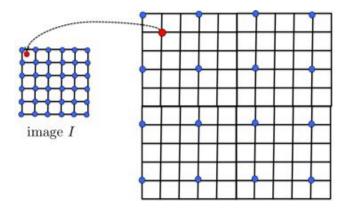
Interpolation via Convolution (1D)



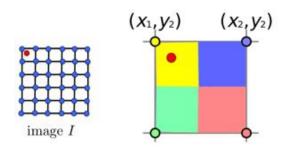


- Let's make this image triple size
- Copy image in every third pixel. What about the remaining pixels in G?

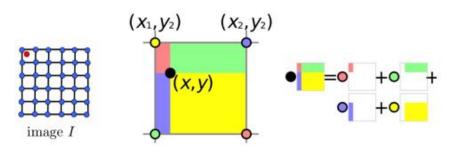
image I



- 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)

h(x,y)

performs linear interpolation

(tent function) performs bilinear interpolation

42.

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

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• And filter for nearest neighbor interpolation?

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

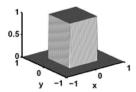
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performs linear interpolation

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(tent function) performs bilinear interpolation

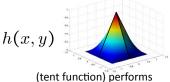
• And filter for nearest neighbor interpolation?



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

h(x)

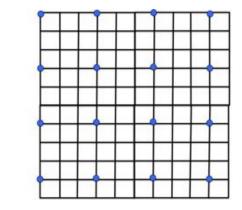
performs linear interpolation



bilinear interpolation

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

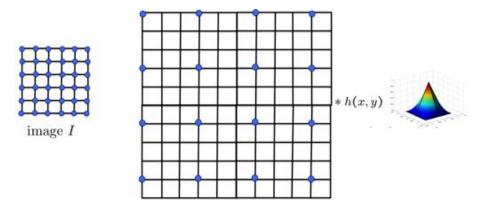
Image Interpolation via Convolution (2D)



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

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]

Deep Learning for Image Superresolution

• You can use DL to increase image resolution!



https://www.youtube.com/watch?v=pZXFXtfd-Ak

Pic credit:

 $\tt https://medium.com/beyondminds/an-introduction-to-super-resolution-using-deep-learning-f60aff9a499dises and the second secon$

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CSC420: Intro to Image Understanding

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/

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CSC420: Intro to Image Understanding

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...

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