Sampling, Resampling, and Warping

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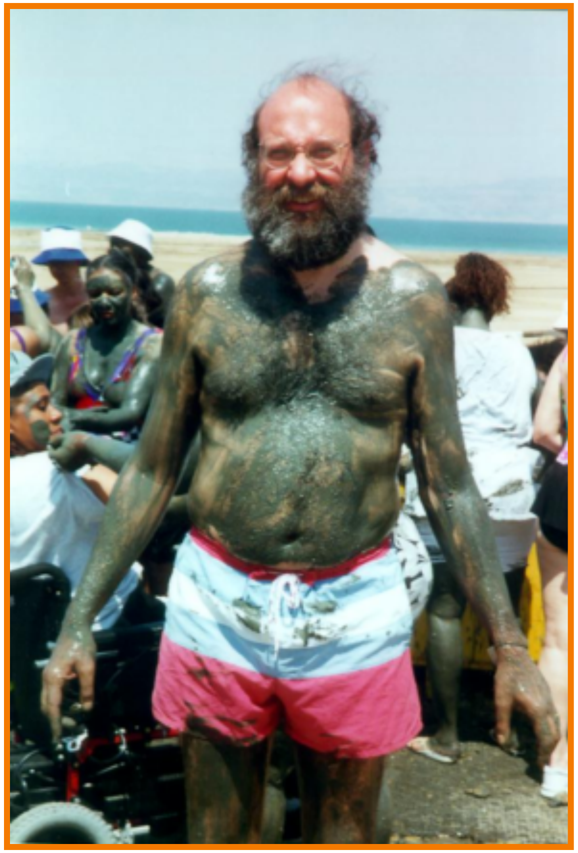
COS 426, Spring 2018

Digital Image Processing

- Changing pixel values
  - Linear: scale, offset, etc.
  - Nonlinear: gamma, saturation, etc.
  - Histogram equalization
- Filtering over neighborhoods
  - Blur & sharpen
  - Detect edges
  - Median
  - Bilateral filter
- Moving image locations
  - Scale
  - Rotate
  - Warp
- Combining images
  - Composite
  - Morph
- Quantization
- Spatial / intensity tradeoff
  - Dithering

Image Warping

- Move pixels of an image



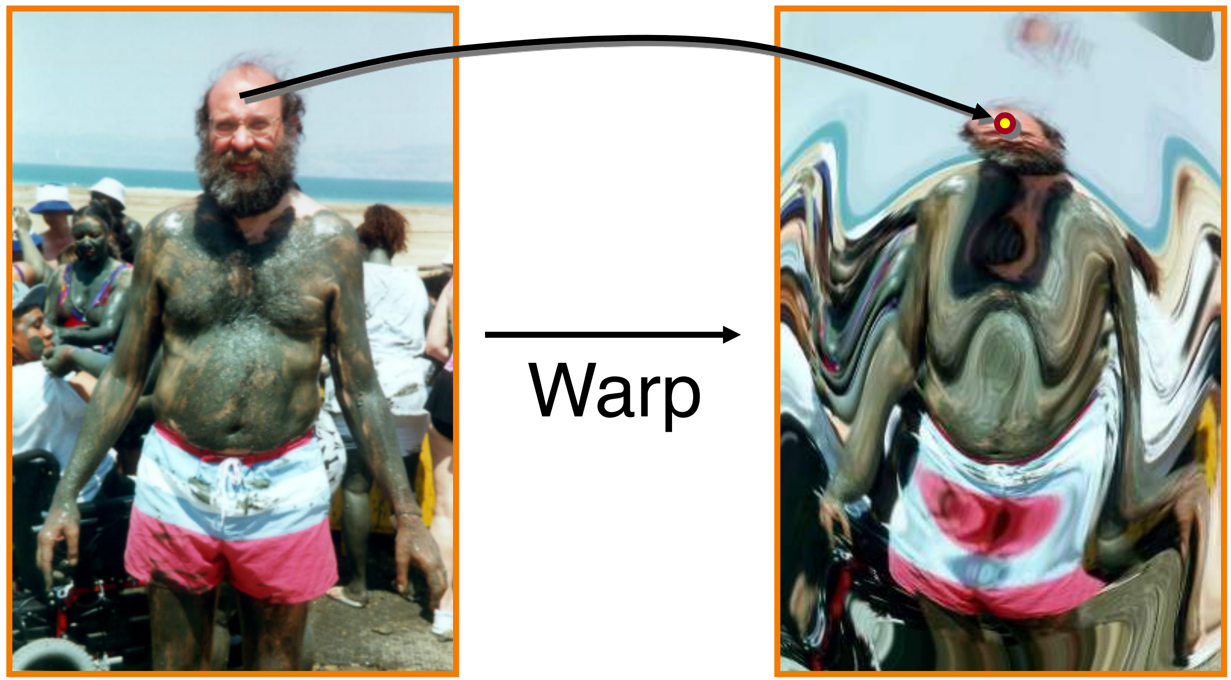
Source image

Warp

Destination image

Image Warping

- Issues:
  - Specifying where every pixel goes (mapping)



Source image Destination image

Image Warping

- Issues:
  - Specifying where every pixel goes (mapping)
  - Computing colors at destination pixels (resampling)

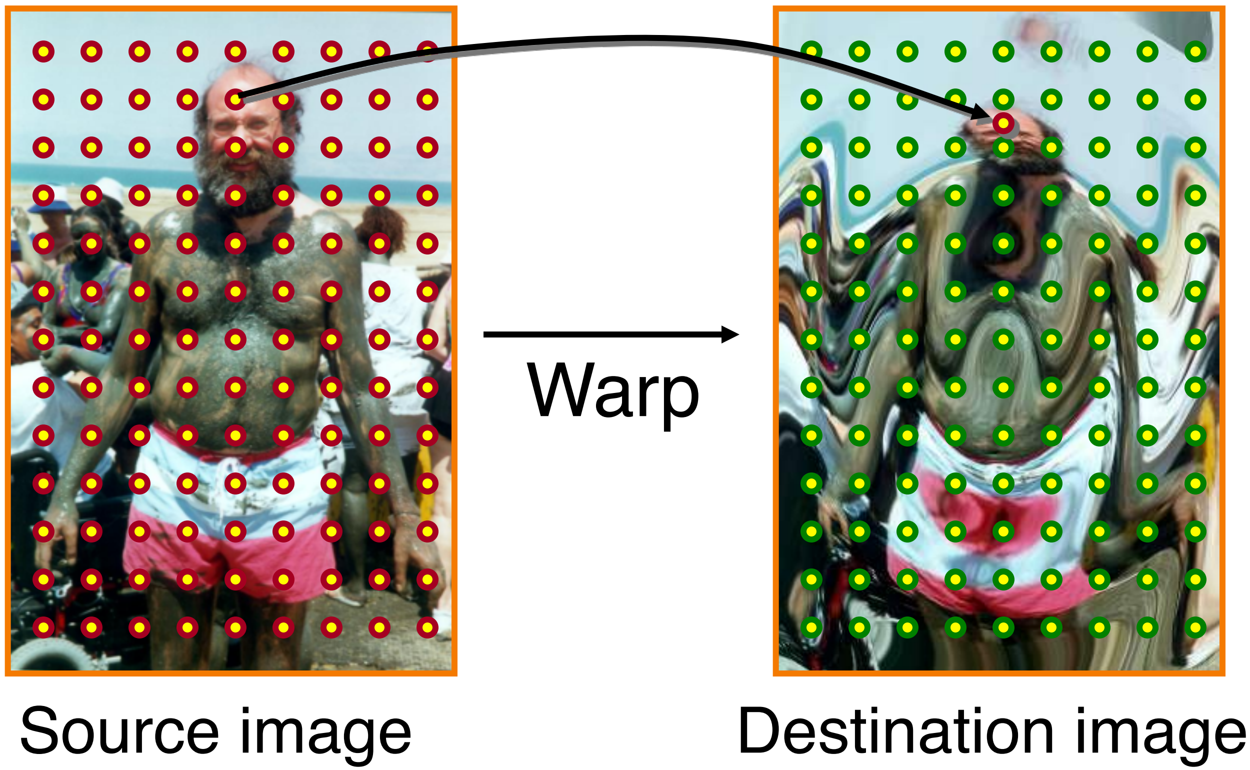


Image Warping

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  - Specifying where every pixel goes (mapping)
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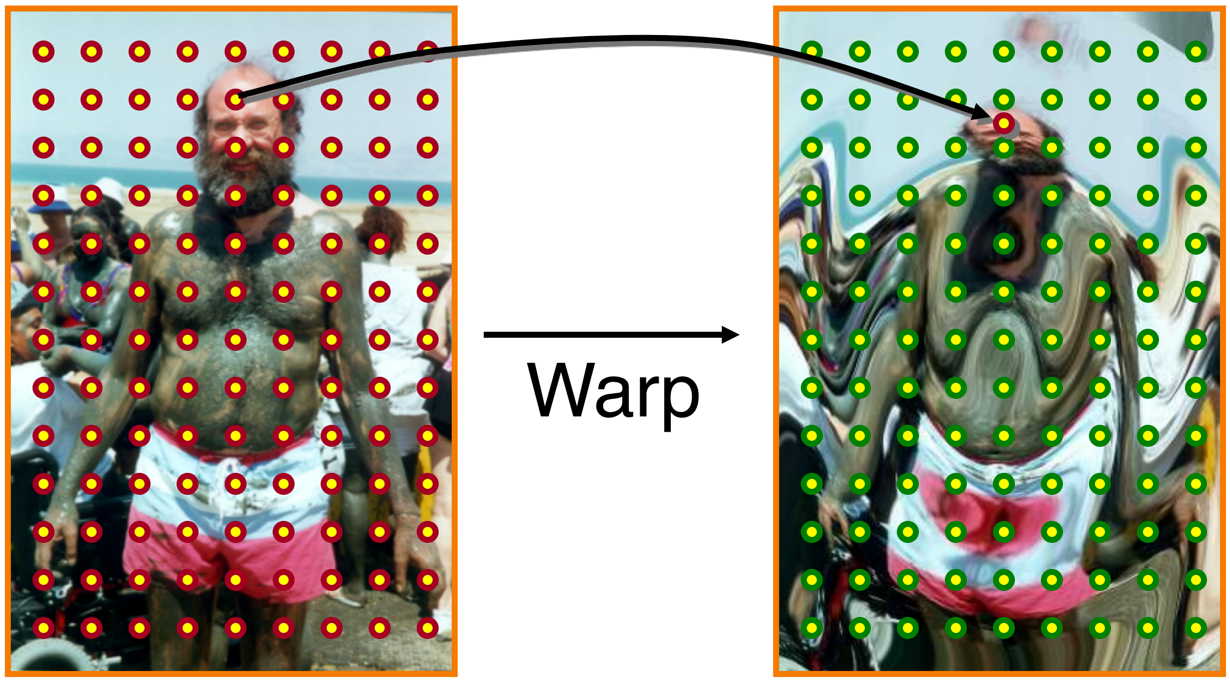
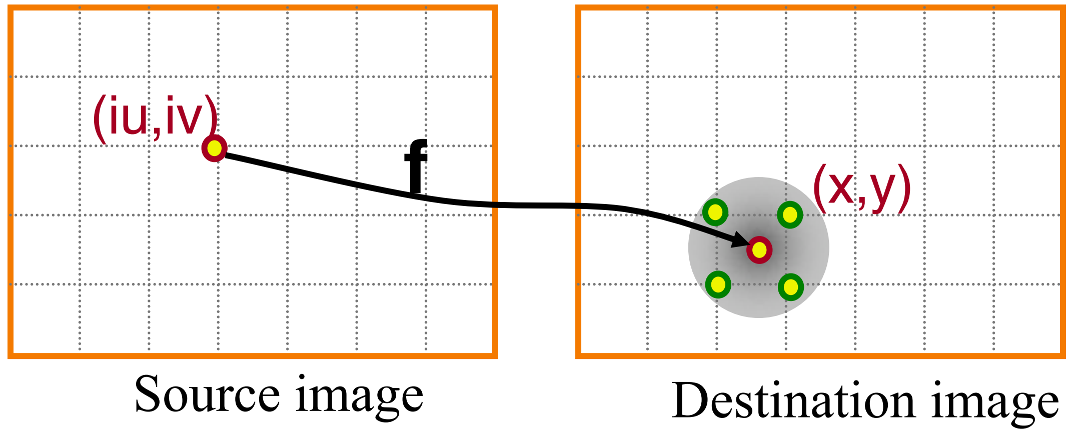


Source image

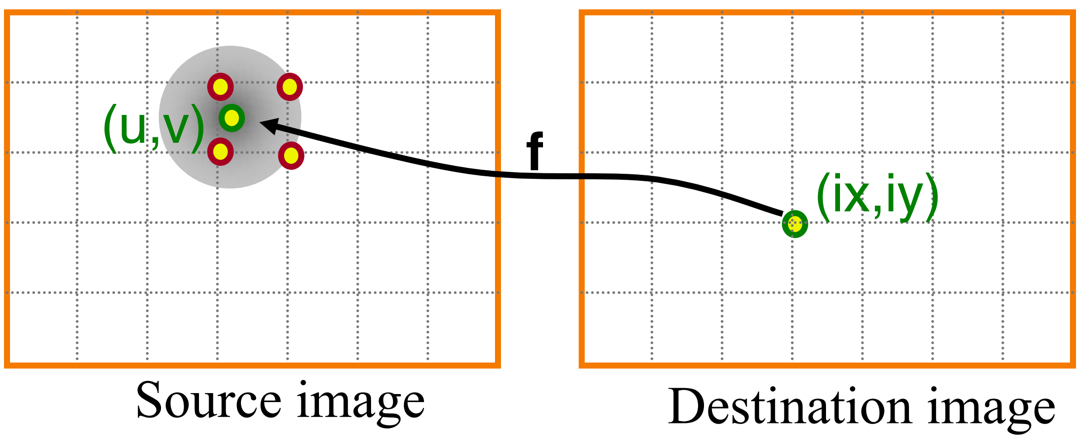
Destination image

Two Options

- Forward mapping

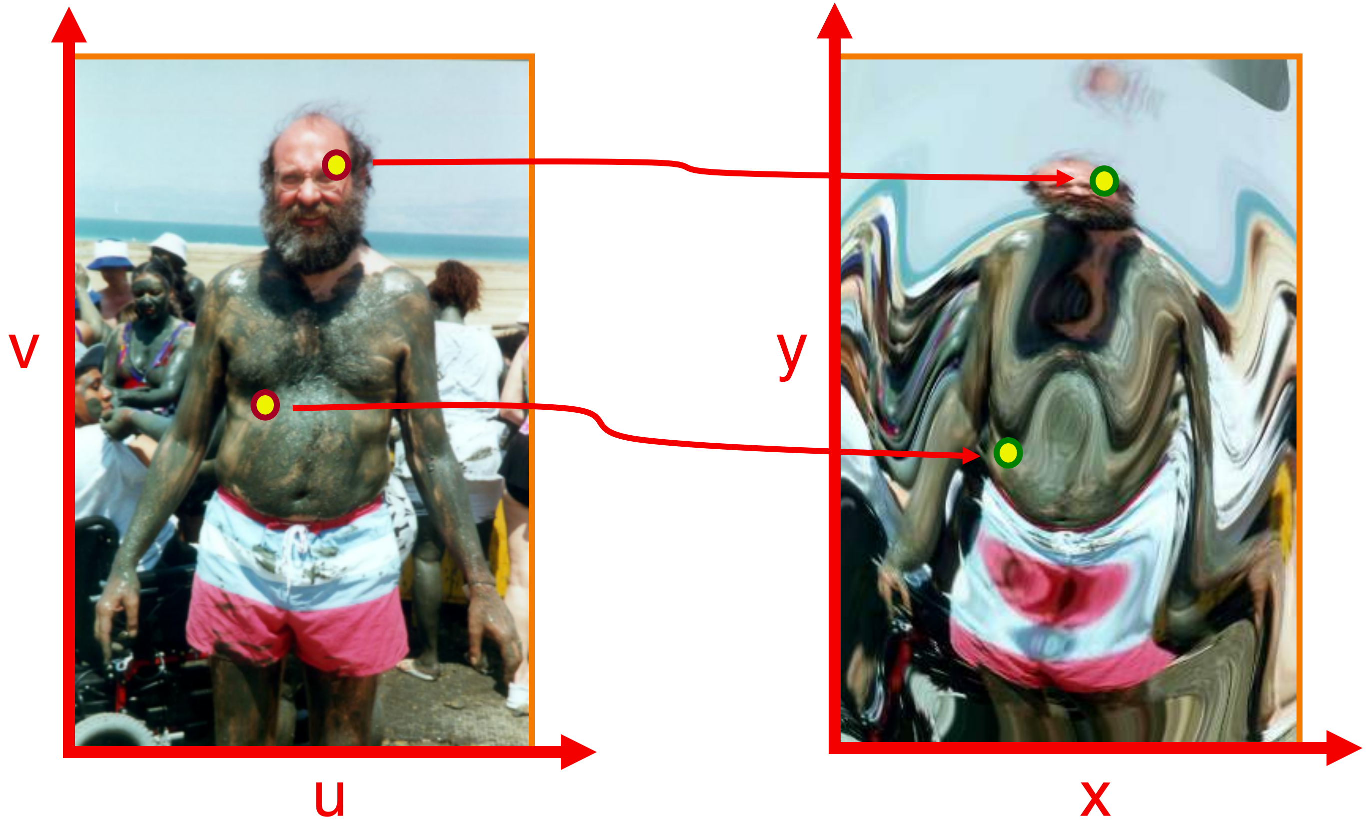


- Reverse mapping



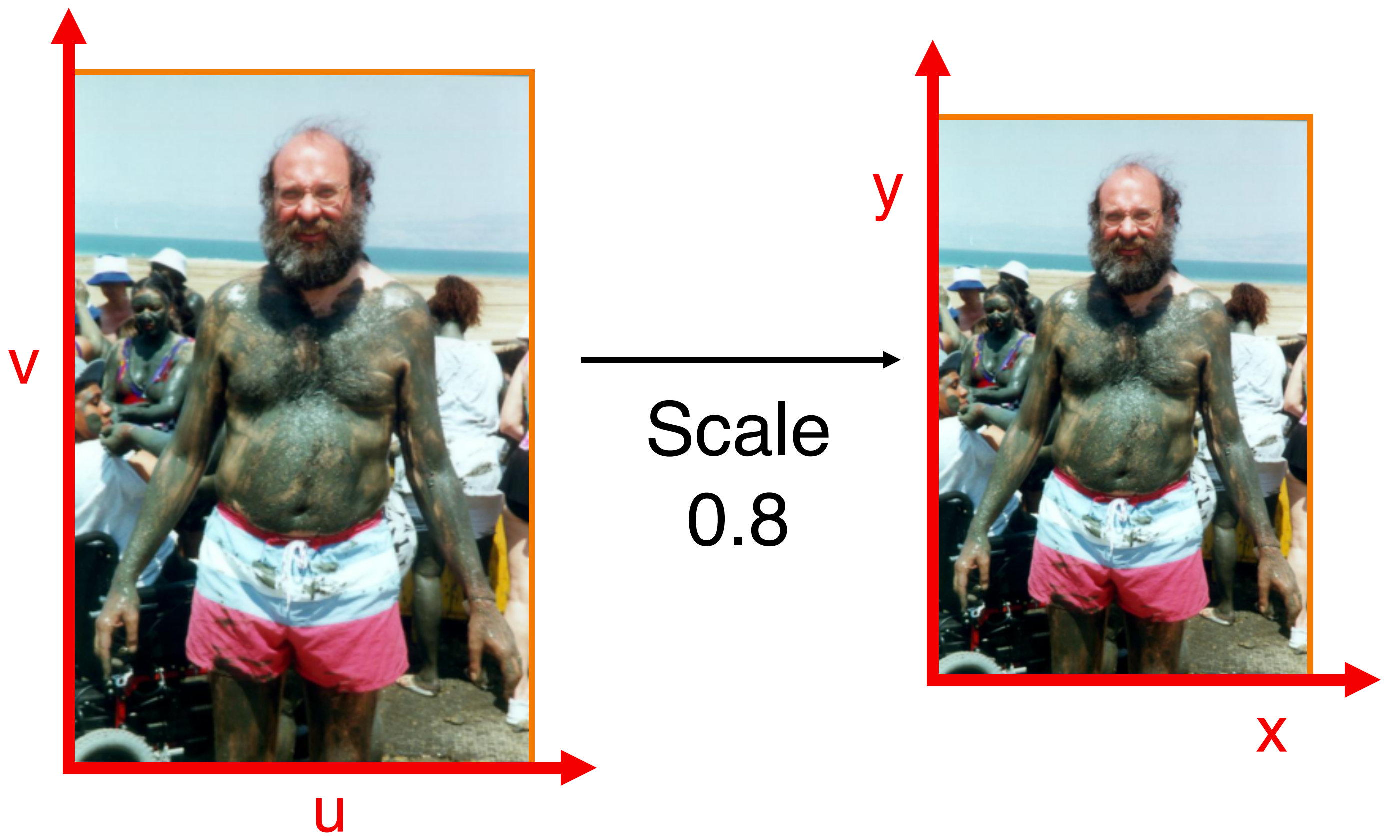
Mapping

- Define transformation
  - Describe the destination (x,y) for every source (u,v) (actually vice-versa, if reverse mapping)



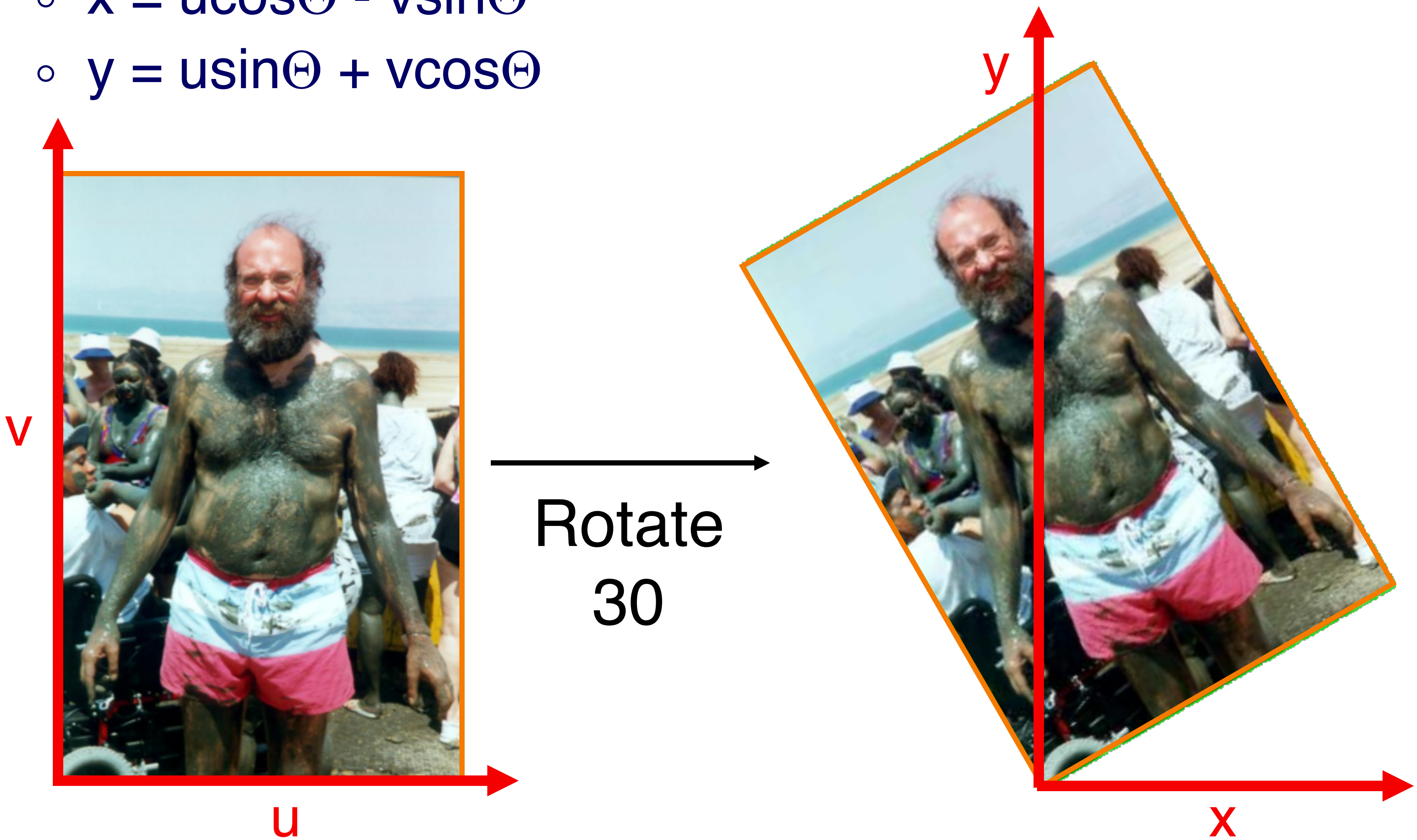
Parametric Mappings

- Scale by *factor*:
  - x = *factor* * u
  - y = *factor* * v



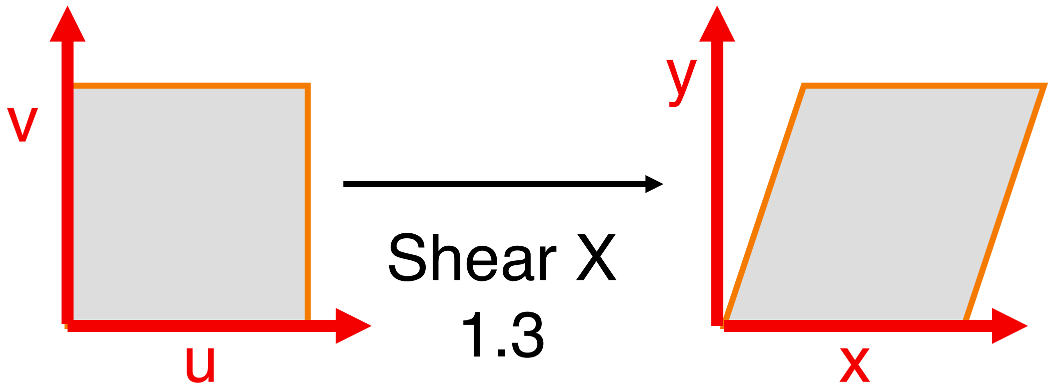
Parametric Mappings

- Rotate by Θ degrees:
  - $x = u\cos\Theta - v\sin\Theta$
  - $y = u\sin\Theta + v\cos\Theta$

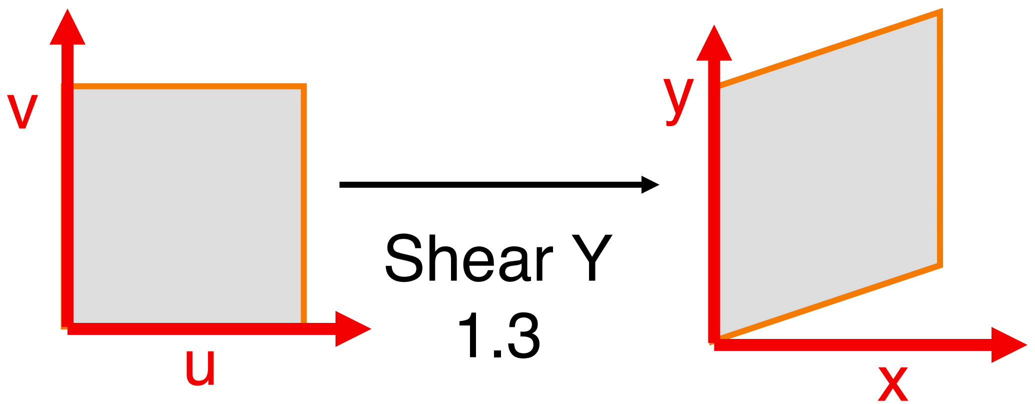


Parametric Mappings

- Shear in X by *factor:*
  - x = u + *factor* * v
  - y = v

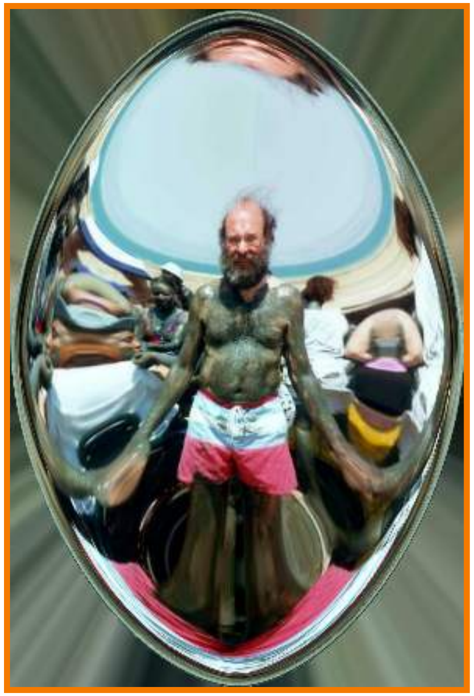


- Shear in Y by *factor:*
  - x = u
  - y = v + *factor* * u



Other Parametric Mappings

- Any function of u and v:
  - $x = f_x(u,v)$
  - $y = f_y(u,v)$



Fish-eye

"Swirl"

"Rain"

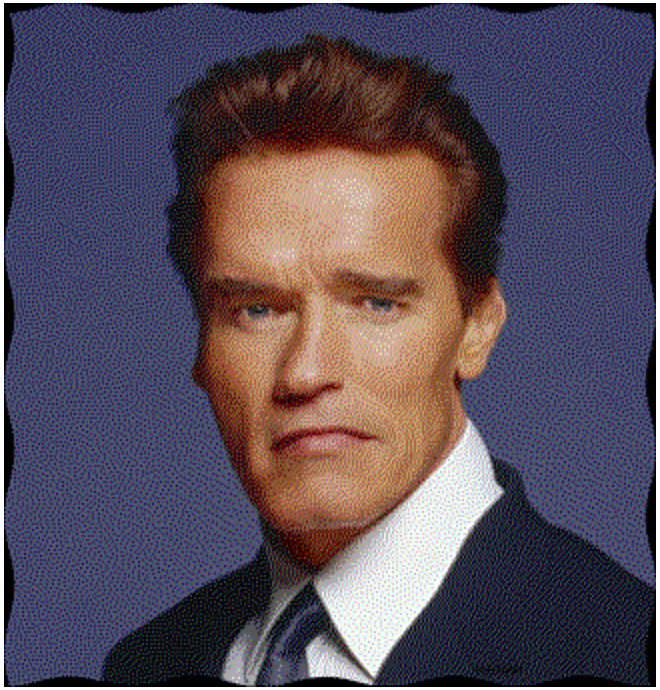
COS426 Examples

Aditya Bhaskara

Wei Xiang

More COS426 Examples

Sid Kapur

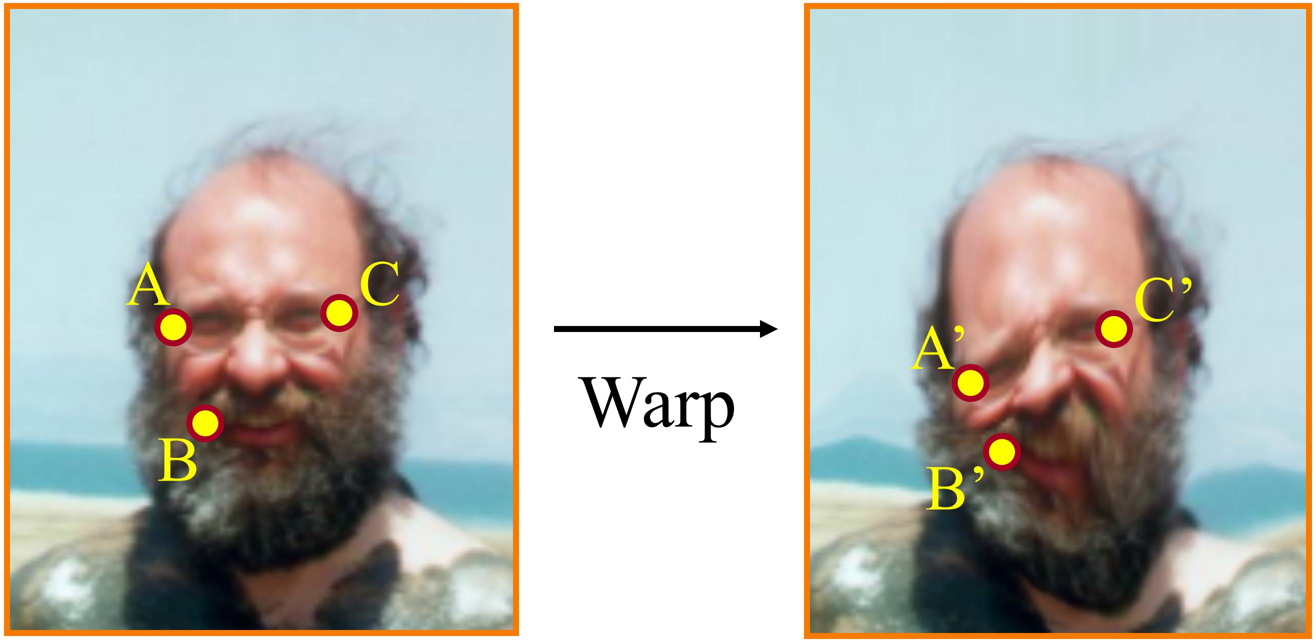


Eirik Bakke

Michael Oranato

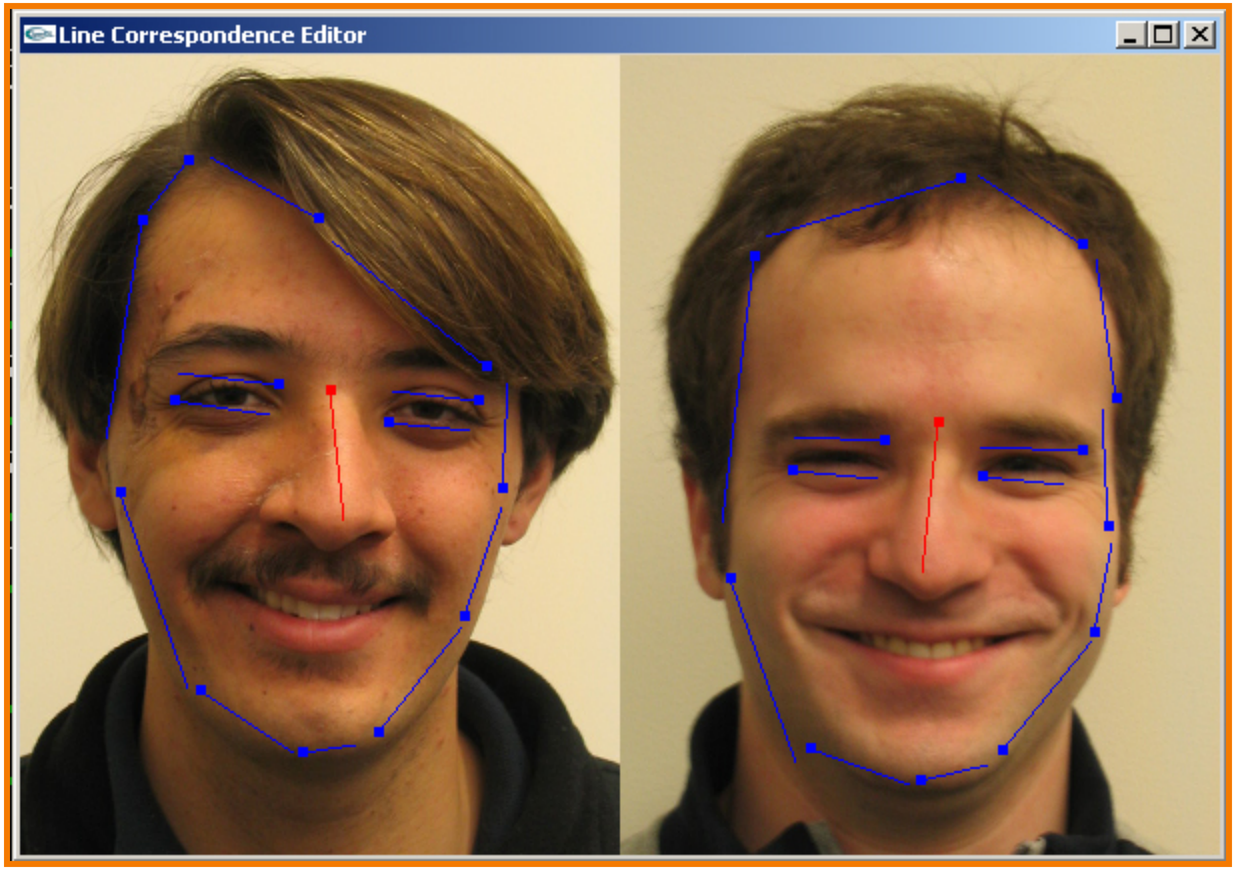
Point Correspondence Mappings

- Mappings implied by correspondences:
  - A ↔ A'
  - B ↔ B'
  - C ↔ C'



Line Correspondence Mappings

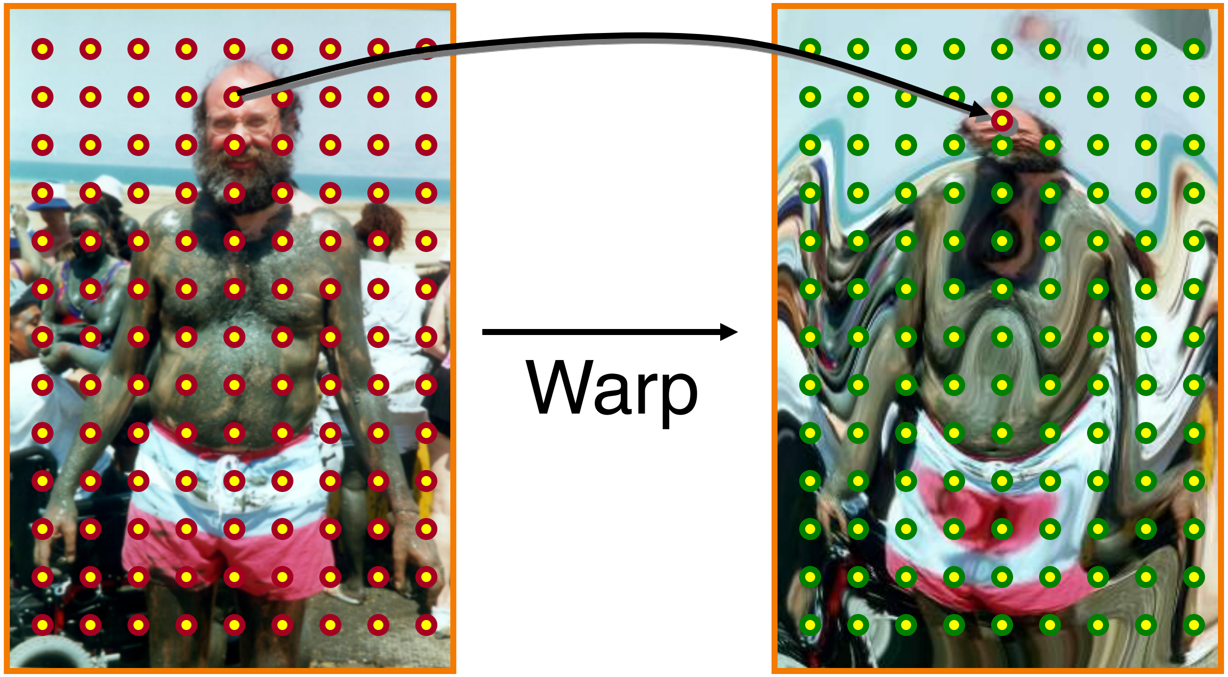
- Beier & Neeley use pairs of lines to specify warp



Beier & Neeley
SIGGRAPH 92

Image Warping

- Issues:
  - Specifying where every pixel goes (mapping)
  - ➢ Computing colors at destination pixels (resampling)



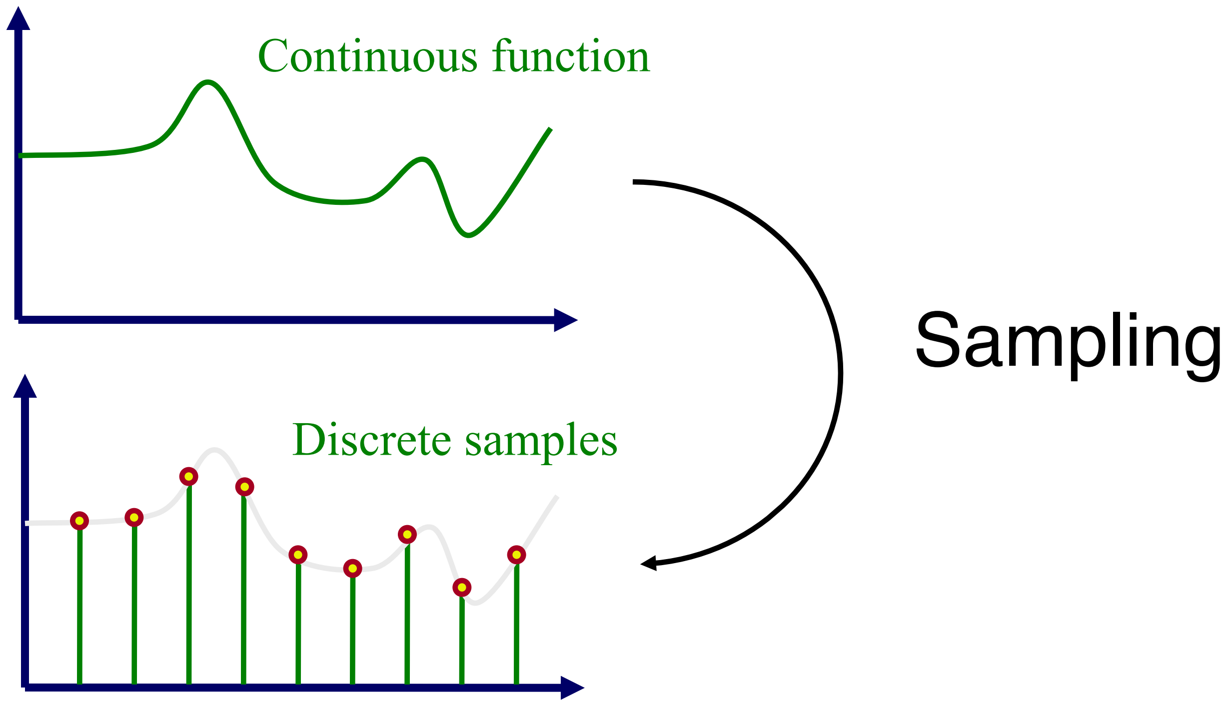
Source image

Destination image

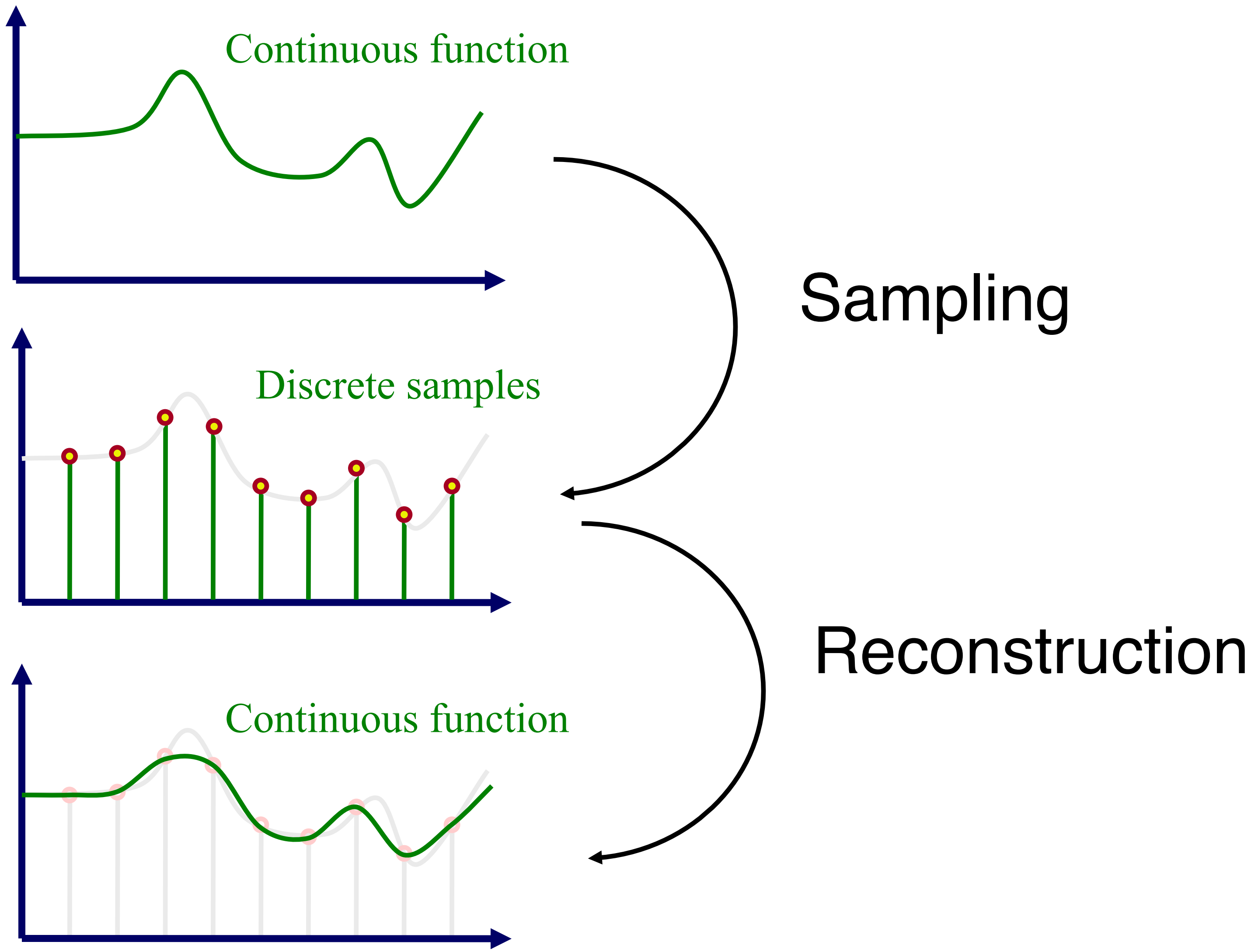
Digital Image Processing

When implementing operations that move pixels, must account for the fact that digital images are sampled versions of continuous ones

Sampling and Reconstruction



Sampling and Reconstruction



Sampling and Reconstruction

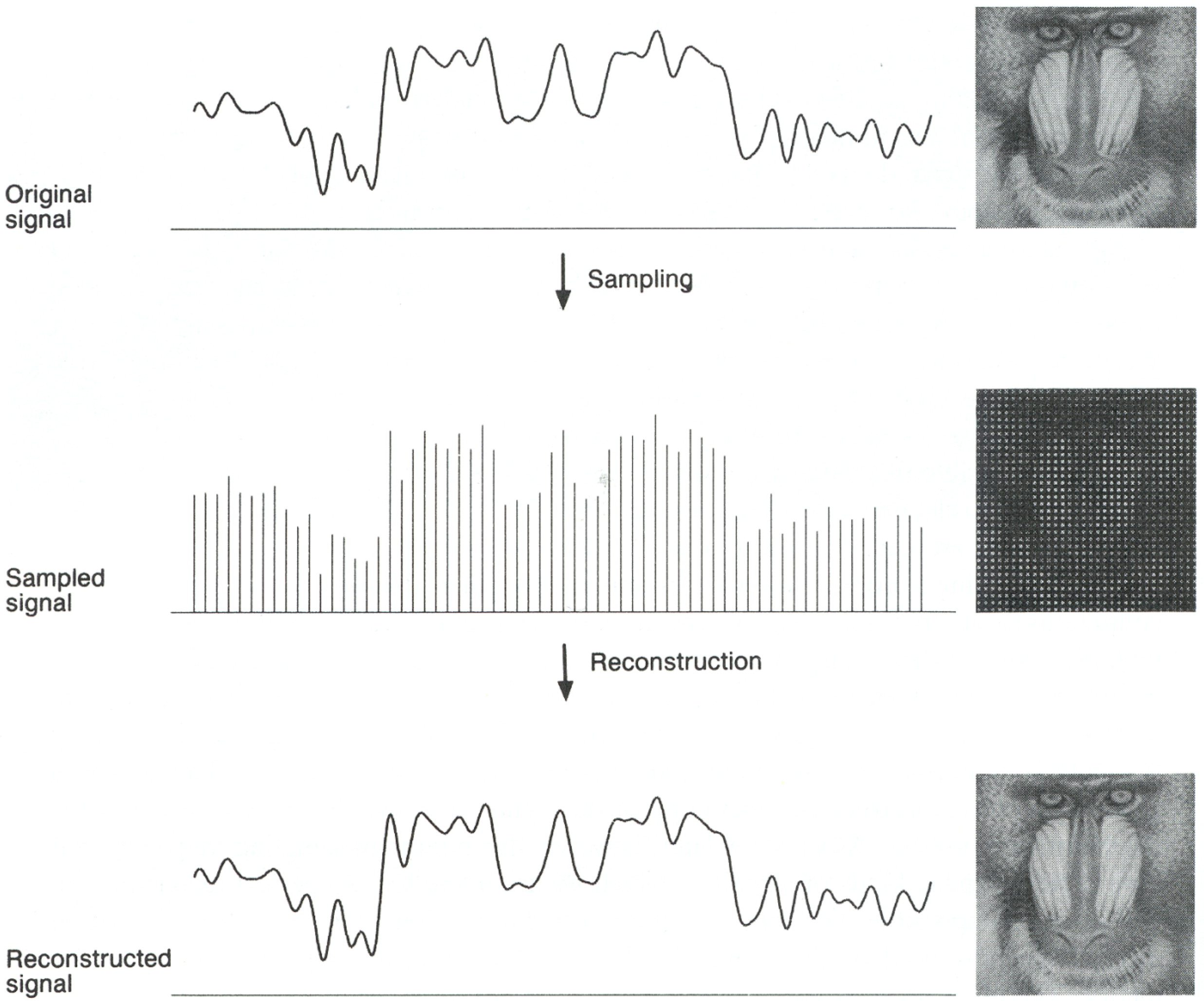


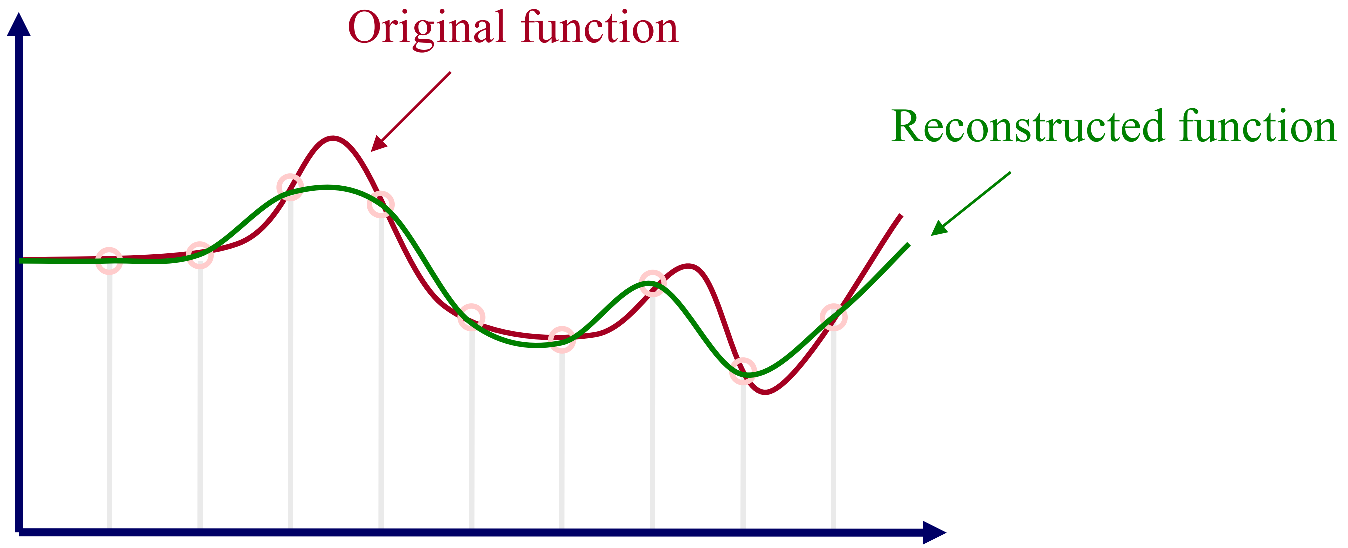
Figure 19.9 FvDFH

Sampling Theory

How many samples are enough?

- How many samples are required to represent a given signal without loss of information?
- What signals can be reconstructed without loss for a given sampling rate?

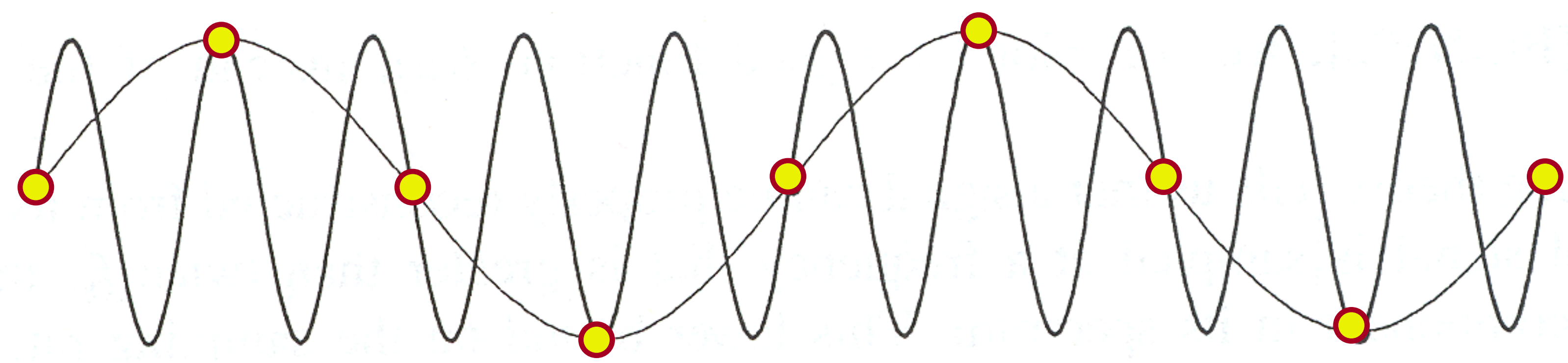
What happens when we use too few samples?



Sampling Theory

What happens when we use too few samples?

- **Aliasing:** high frequencies masquerade as low ones



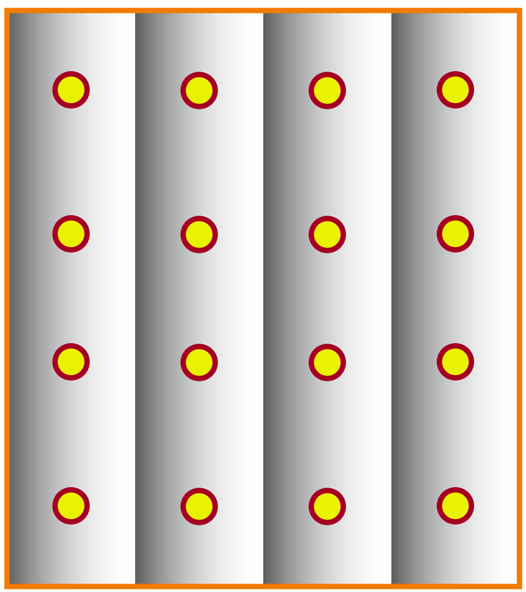
Specifically, in graphics:

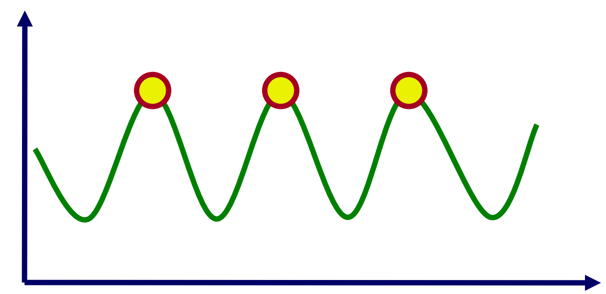
- Spatial aliasing
- Temporal aliasing

Figure 14.17 FvDFH

Spatial Aliasing

Artifacts due to limited spatial resolution





Spatial Aliasing

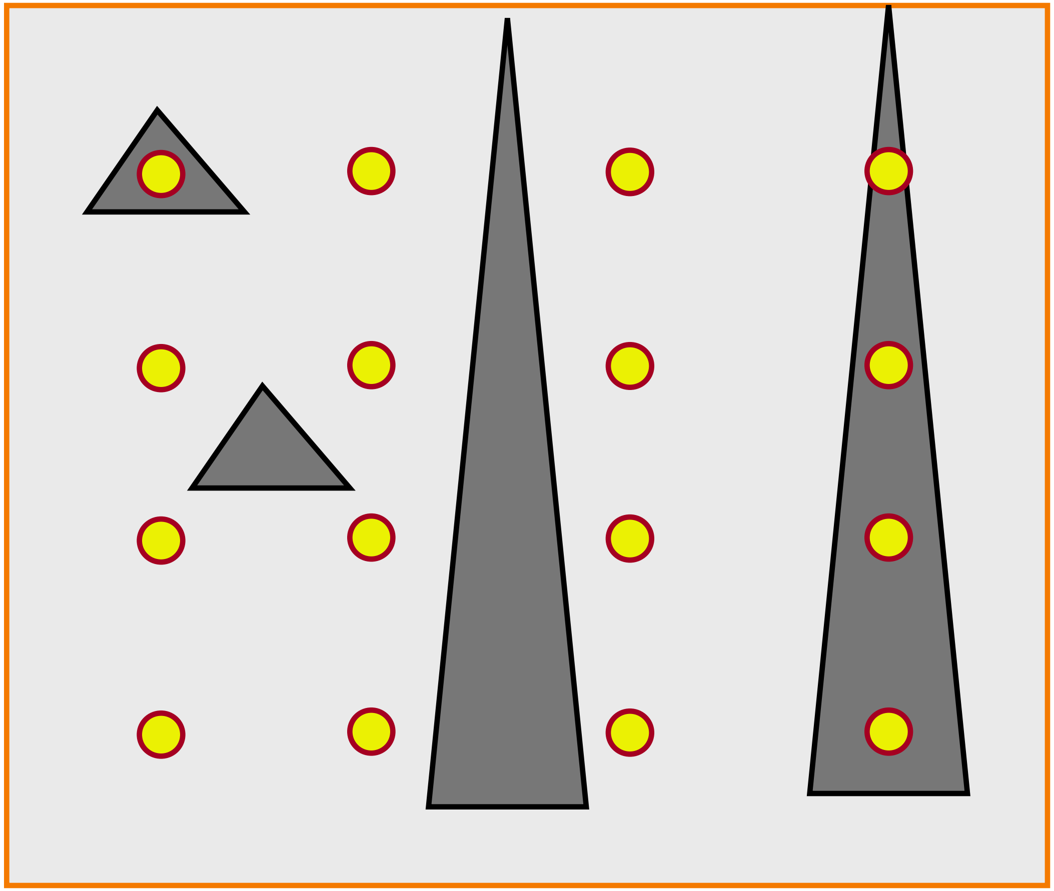
Artifacts due to limited spatial resolution

(Barely) adequate sampling

Inadequate sampling

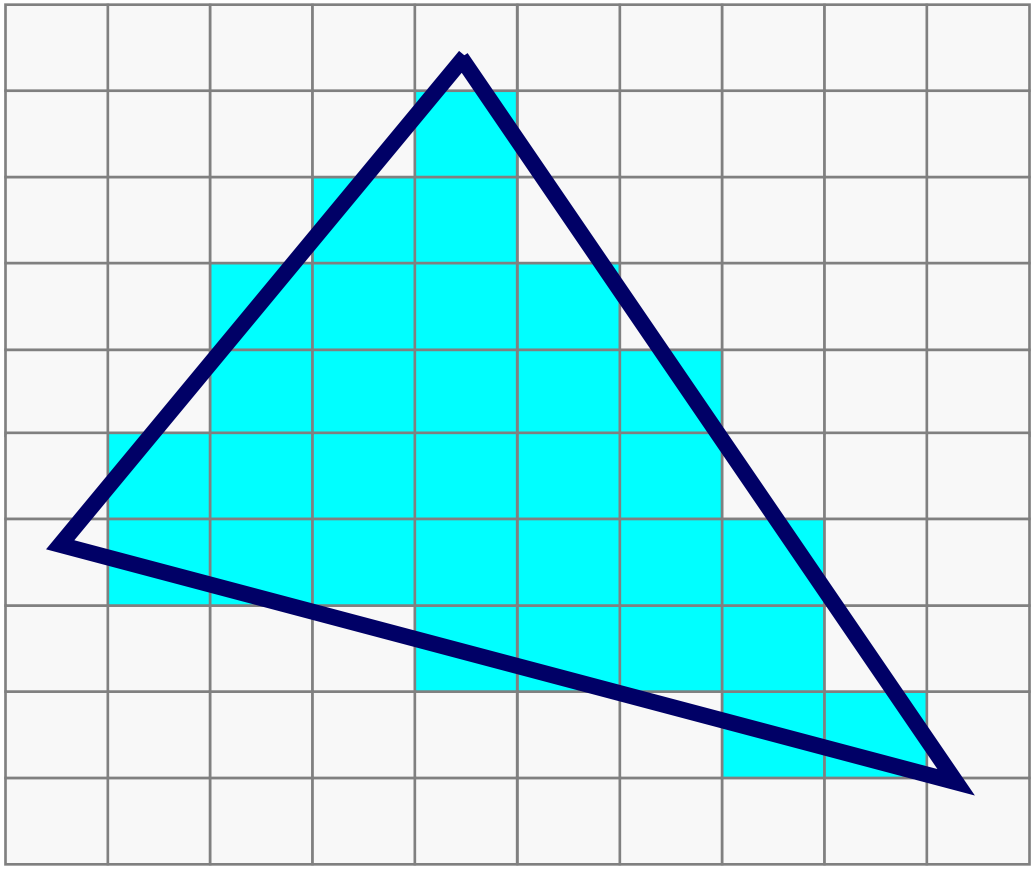
Spatial Aliasing

Artifacts due to limited spatial resolution



Spatial Aliasing

Artifacts due to limited spatial resolution

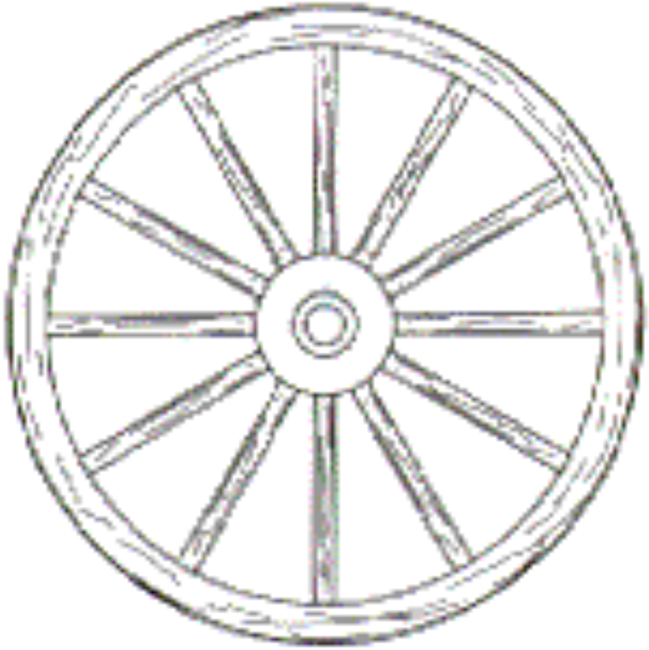


"Jaggies"

Temporal Aliasing

Artifacts due to limited temporal resolution

- Strobing
- Flickering



Temporal Aliasing

Artifacts due to limited temporal resolution

- Strobing
- Flickering

Temporal Aliasing

Artifacts due to limited temporal resolution

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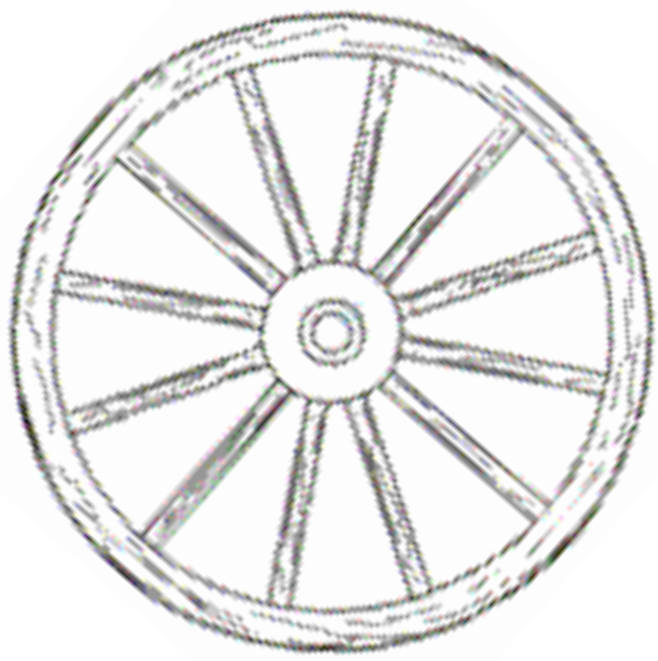
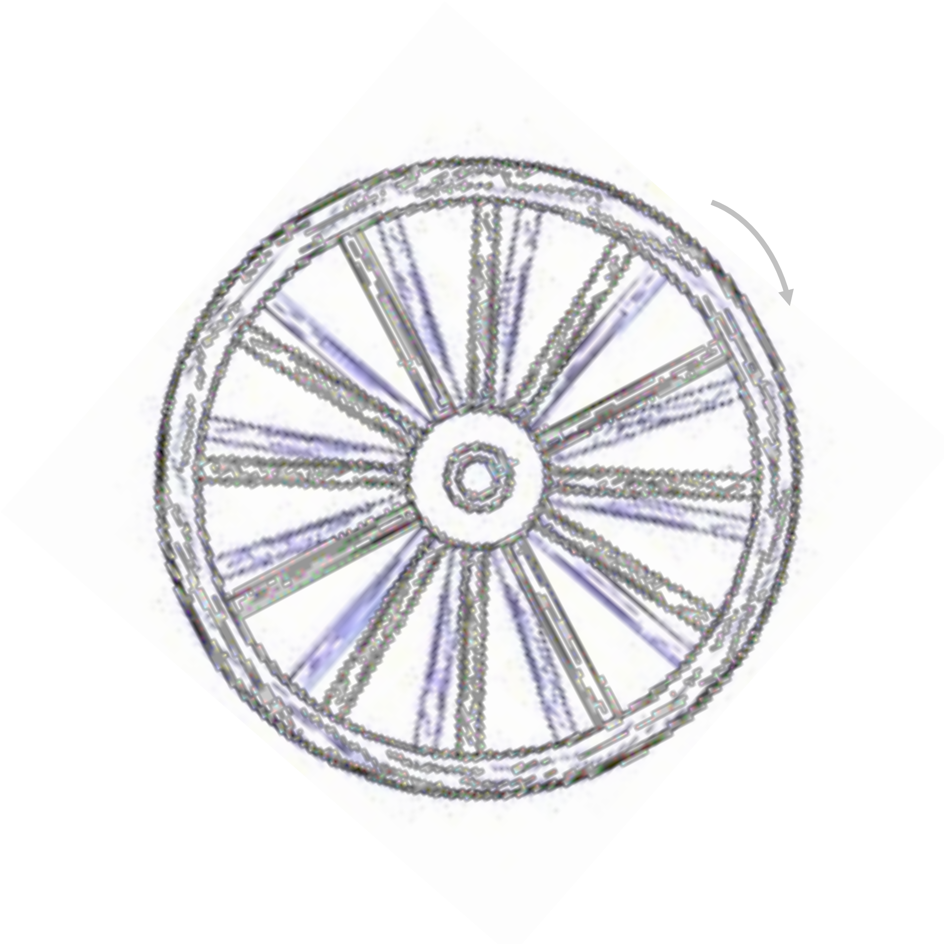
- Flickering



Temporal Aliasing

Artifacts due to limited temporal resolution

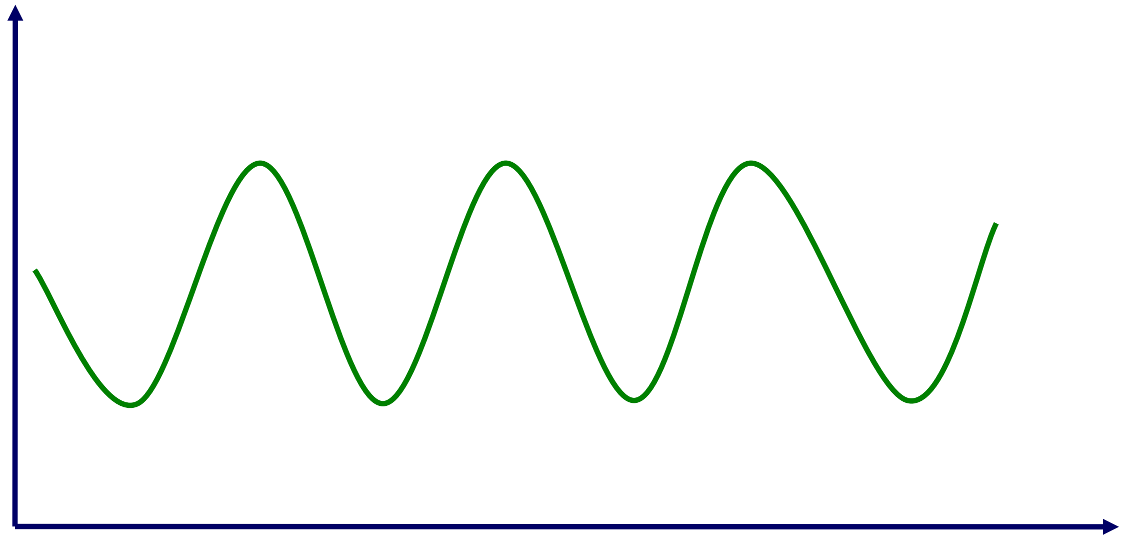
- Strobing
- Flickering



Sampling Theory

How many samples are enough to avoid aliasing?

- How many samples are required to represent a given signal without loss of information?
- What signals can be reconstructed without loss for a given sampling rate?



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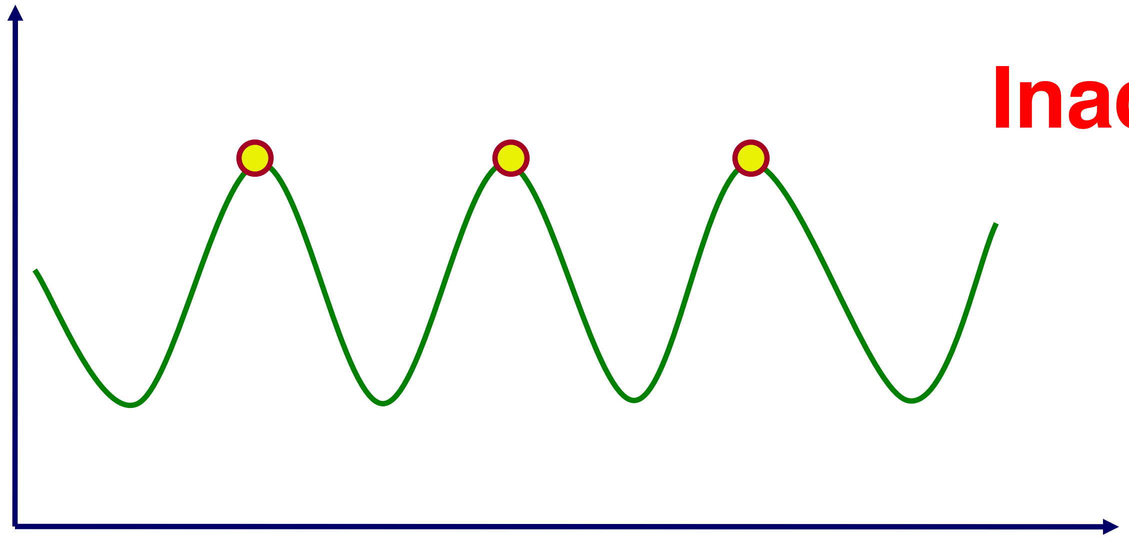

Inadequate

Sampling Theory

How many samples are enough to avoid aliasing?

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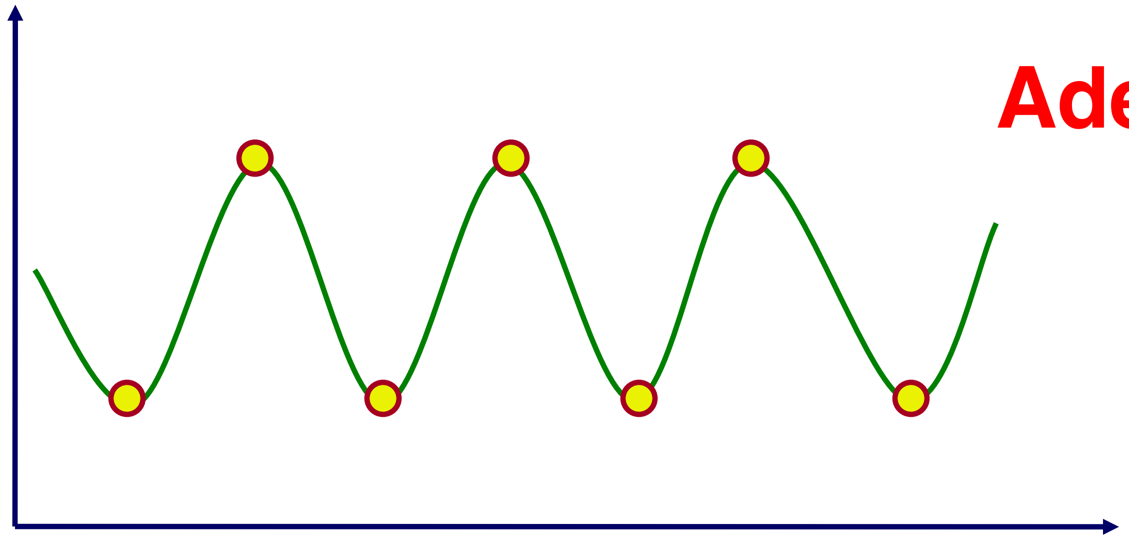

Adequate?

Sampling Theory

How many samples are enough to avoid aliasing?

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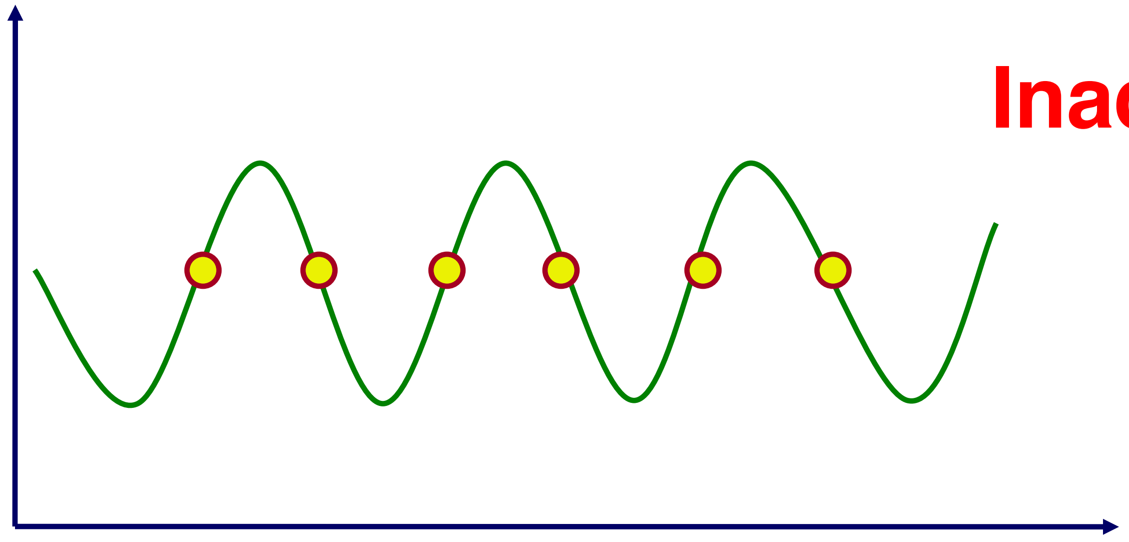

Inadequate

Sampling Theory

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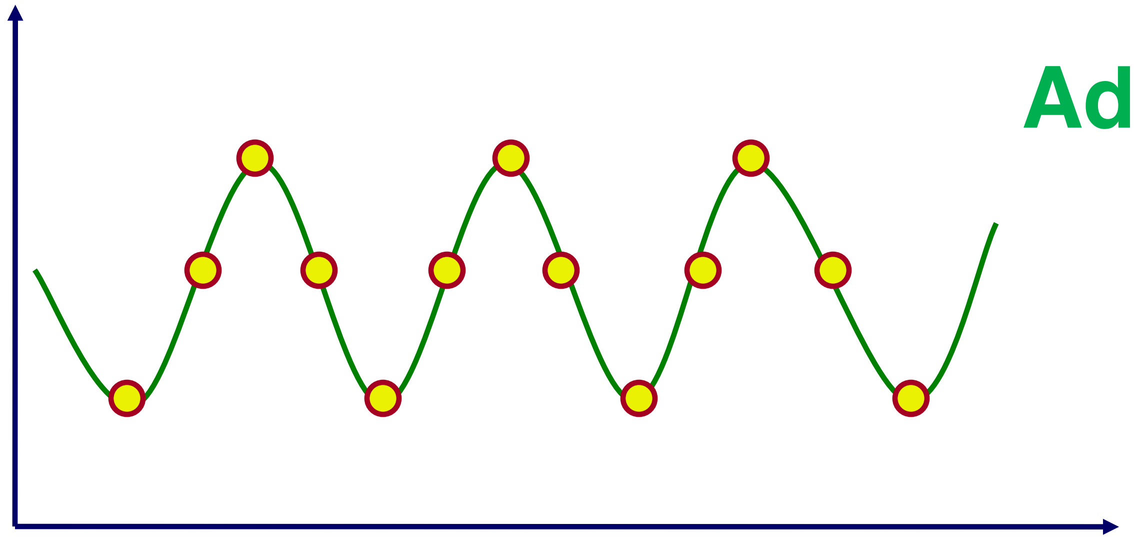

Adequate

Sampling Theory

How many samples are enough to avoid aliasing?

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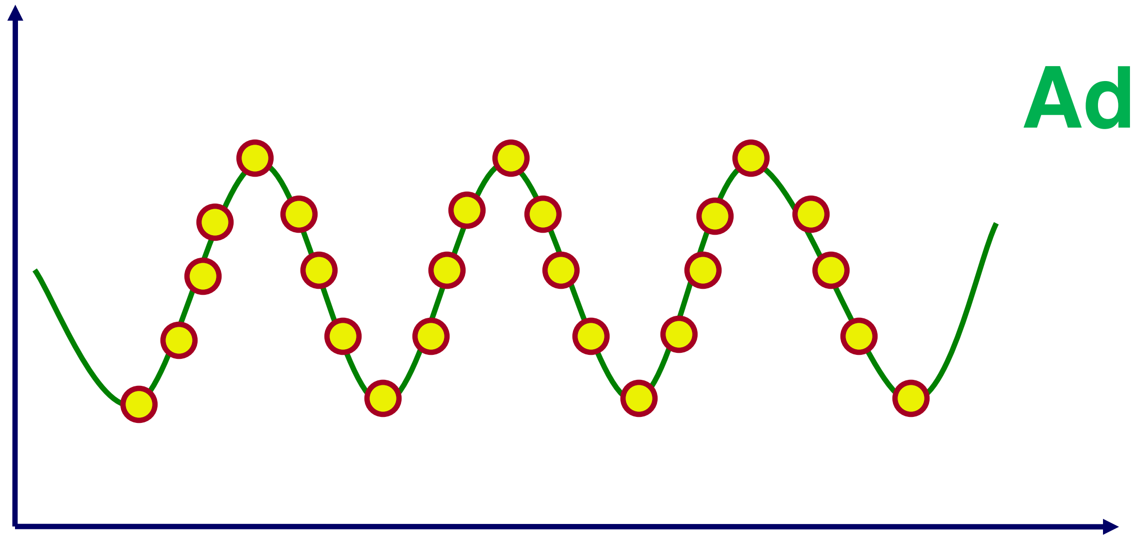
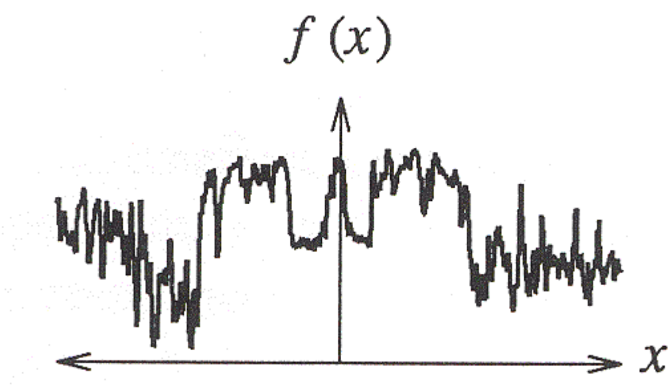


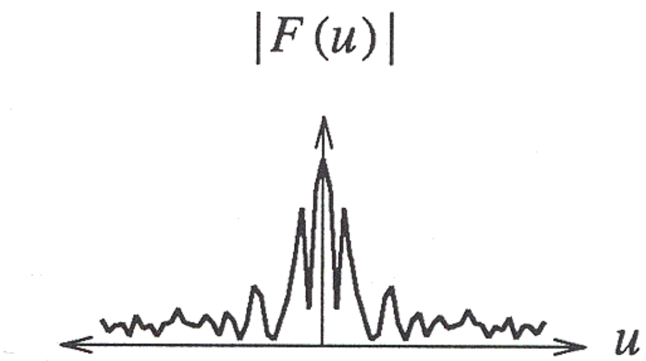
Adequate

Spectral Analysis

- Spatial domain:
  - Function: f(x)
  - Filtering: convolution

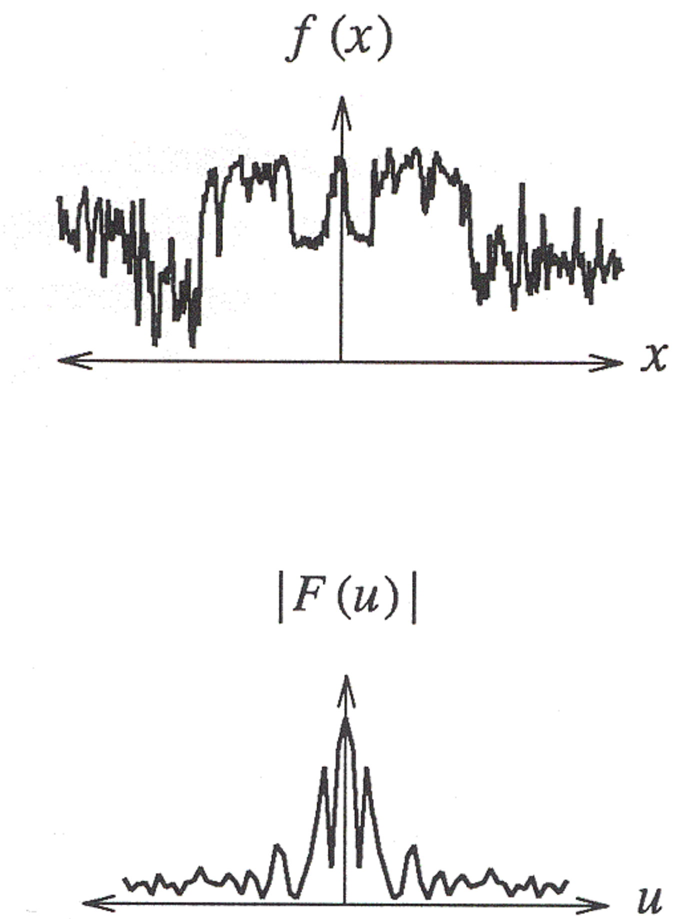
- Frequency domain:
  - Function: F(u)
  - Filtering: multiplication





Any signal can be written as a sum of periodic functions.

Fourier Transform



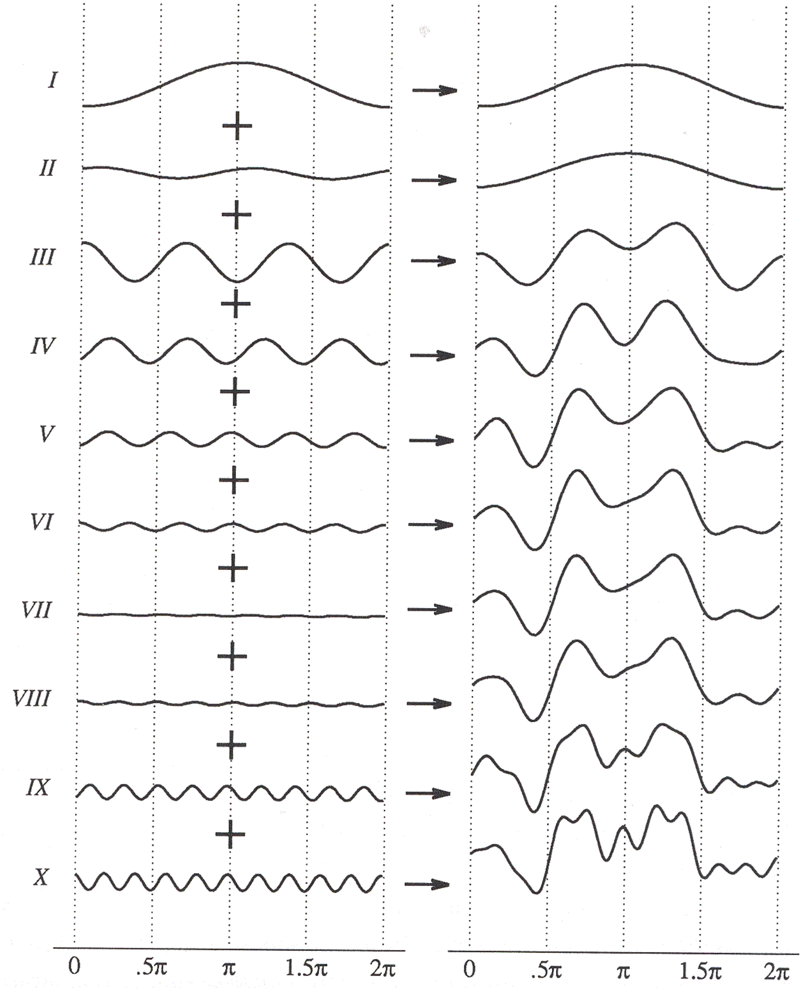


Figure 2.6 Wolberg

Fourier Transform

- Fourier transform:

$$F(u) = \int_{-\infty}^{\infty} f(x) e^{-i2\pi xu} dx$$

- Inverse Fourier transform:

$$f(x) = \int_{-\infty}^{\infty} F(u) e^{+i2\pi ux} du$$

Sampling Theorem

- A signal can be reconstructed from its samples iff it has no content ≥ ½ the sampling frequency
  – Shannon
- The minimum sampling rate for a bandlimited function is called the "Nyquist rate"

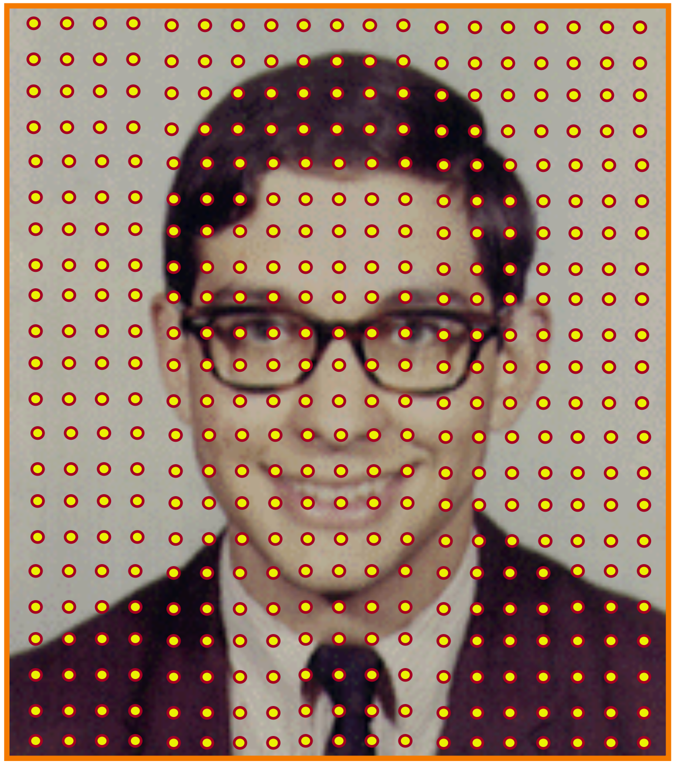
A signal is *bandlimited* if its
highest frequency is bounded.
That frequency is called the bandwidth.

Antialiasing

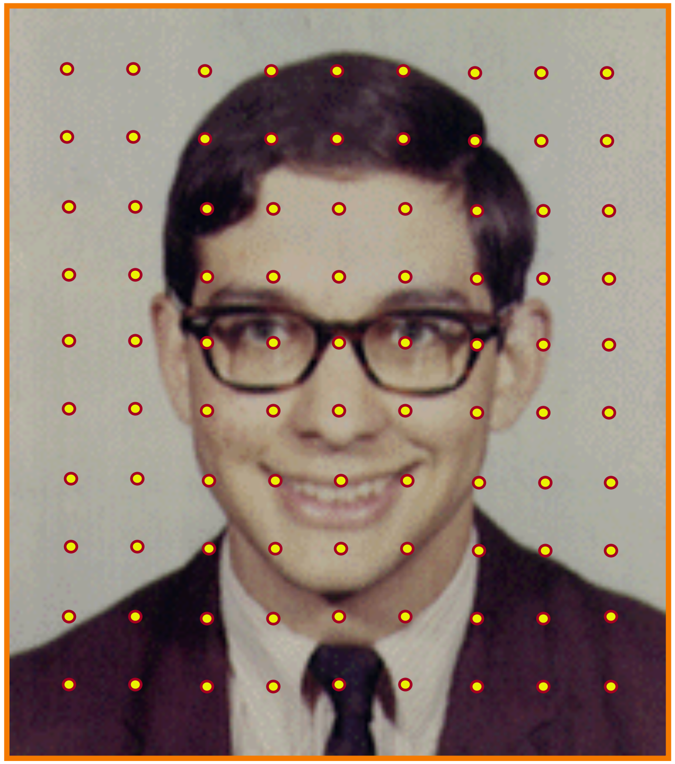
- Sample at higher rate
  - Not always possible
  - Doesn't always solve the problem
- Pre-filter to form bandlimited signal
  - Use low-pass filter to limit signal to < 1/2 sampling rate
  - Trades blurring for aliasing

Image Processing

Consider scaling the image (or, equivalently, reducing resolution)



Original image



1/4 resolution

Image Processing

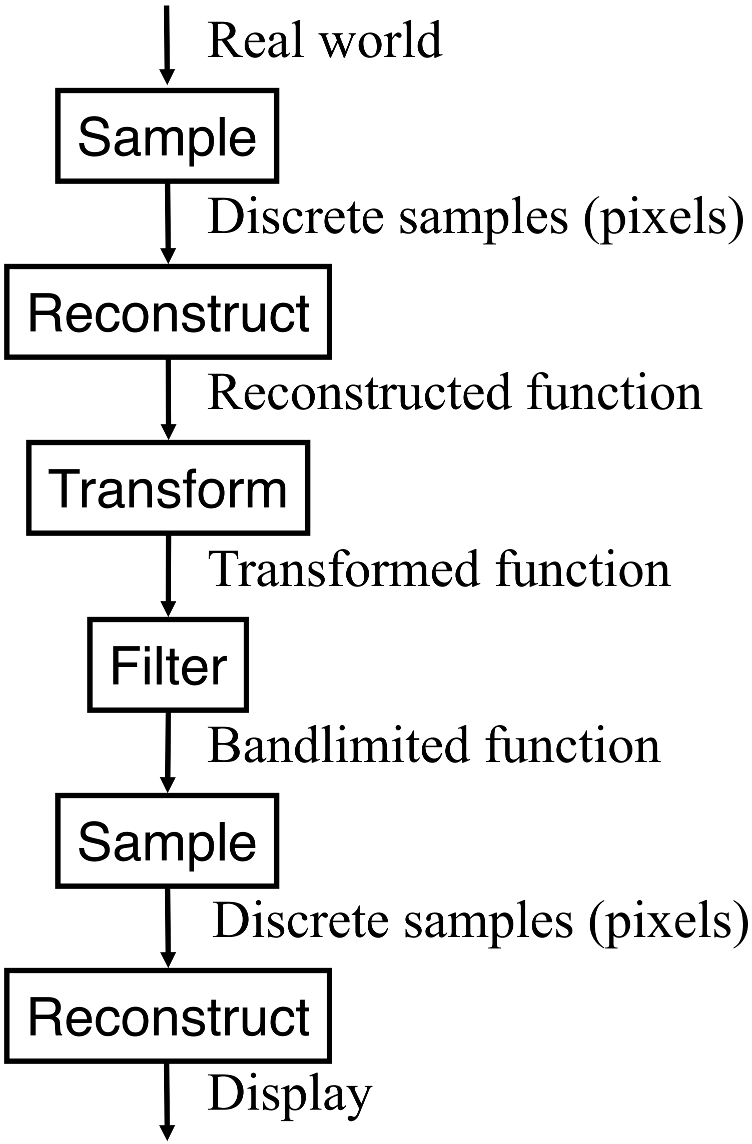


Image Processing

Real world

Sample

Discrete samples (pixels)

Reconstruct

Reconstructed function

Transform

Transformed function

Filter

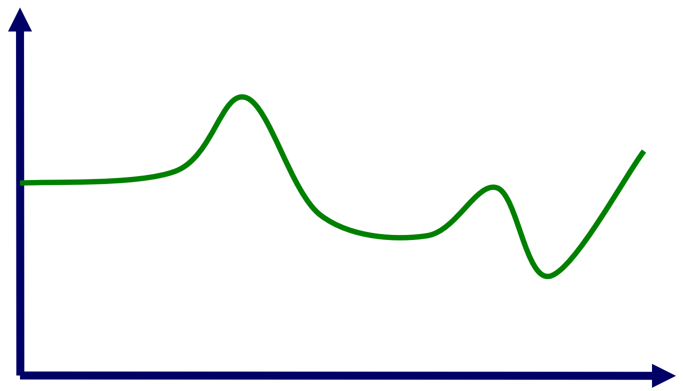
Bandlimited function

Sample

Discrete samples (pixels)

Reconstruct

Display



Continuous Function

Image Processing

Real world

Sample

Discrete samples (pixels)

Reconstruct

Reconstructed function

Transform

Transformed function

Filter

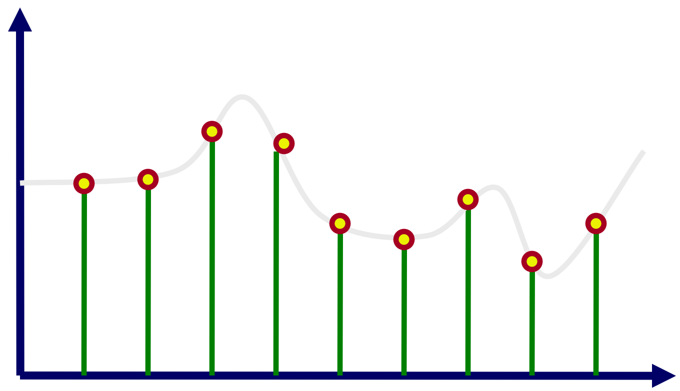
Bandlimited function

Sample

Discrete samples (pixels)

Reconstruct

Display



Discrete Samples

Image Processing

Real world

Sample

Discrete samples (pixels)

Reconstruct

Reconstructed function

Transform

Transformed function

Filter

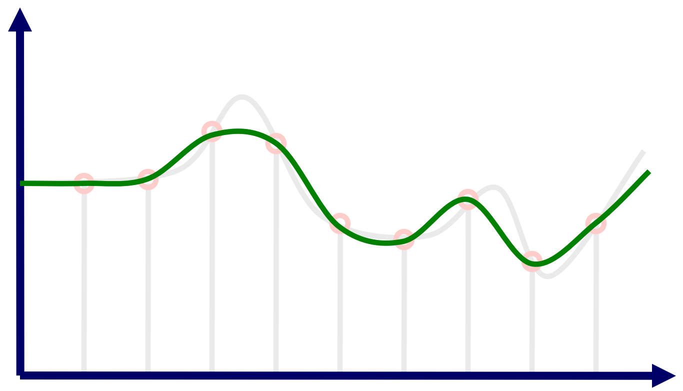
Bandlimited function

Sample

Discrete samples (pixels)

Reconstruct

Display



Reconstructed Function

Image Processing

Real world

Sample

Discrete samples (pixels)

Reconstruct

Reconstructed function

Transform

Transformed function

Filter

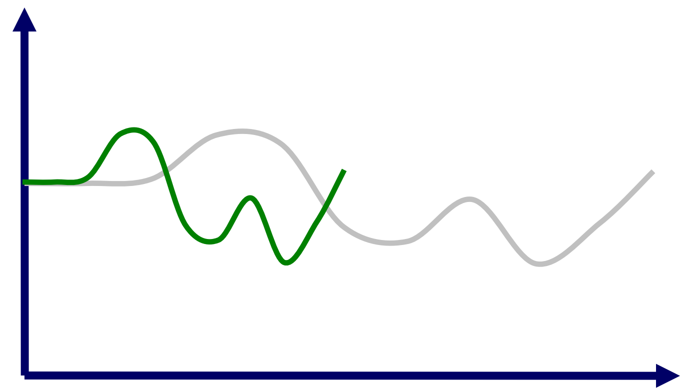
Bandlimited function

Sample

Discrete samples (pixels)

Reconstruct

Display



Transformed Function

Image Processing

Real world

Sample

Discrete samples (pixels)

Reconstruct

Reconstructed function

Transform

Transformed function

Filter

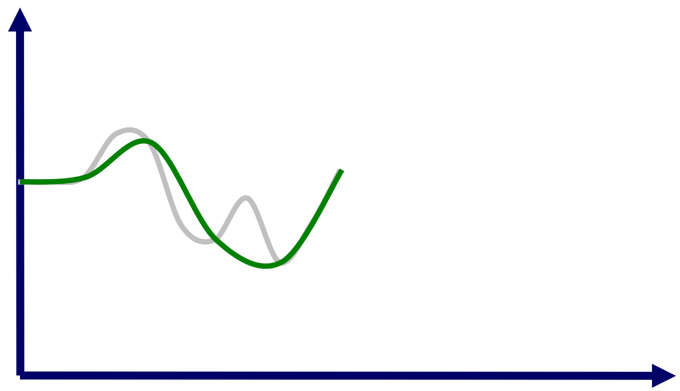
Bandlimited function

Sample

Discrete samples (pixels)

Reconstruct

Display



Bandlimited Function

Image Processing

Real world

Sample

Discrete samples (pixels)

Reconstruct

Reconstructed function

Transform

Transformed function

Filter

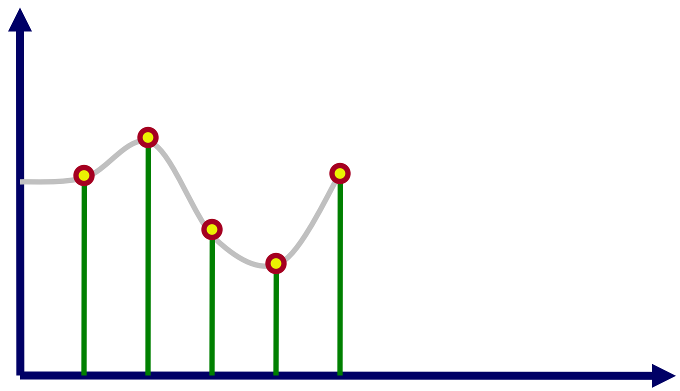
Bandlimited function

Sample

Discrete samples (pixels)

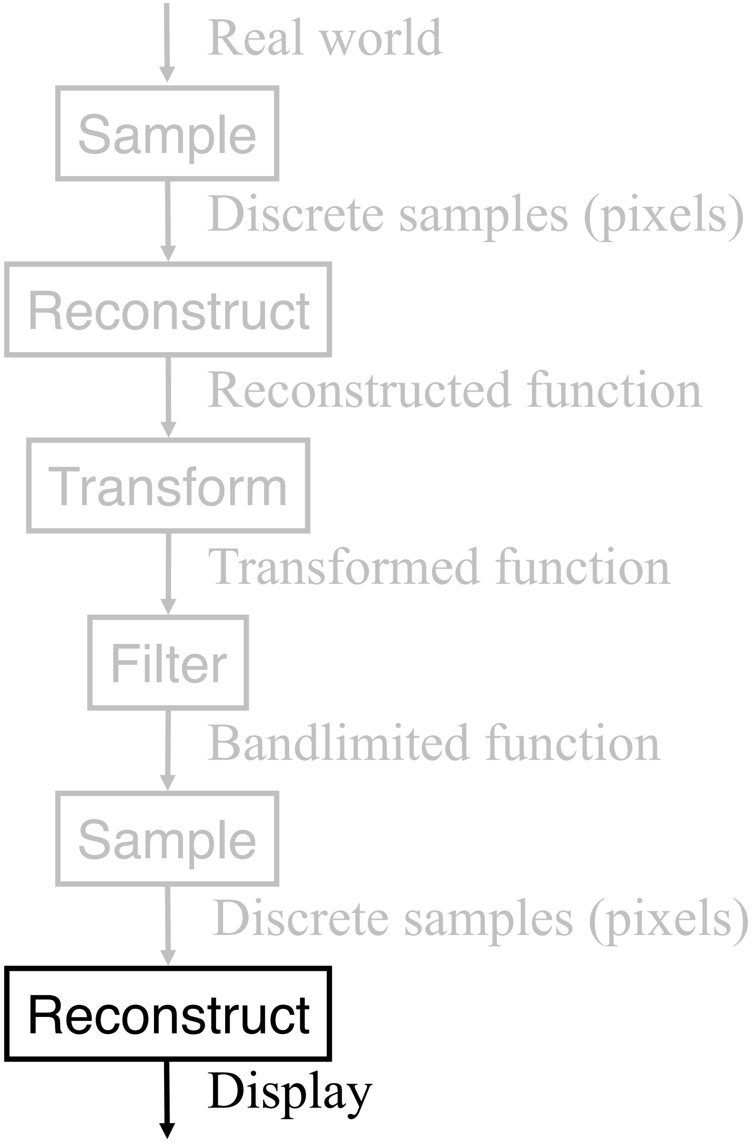
Reconstruct

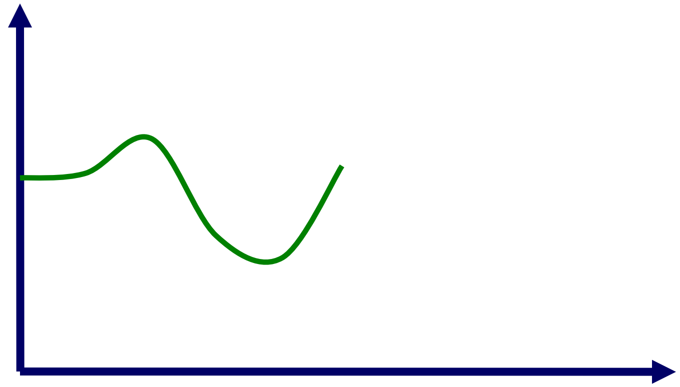
Display



Discrete samples

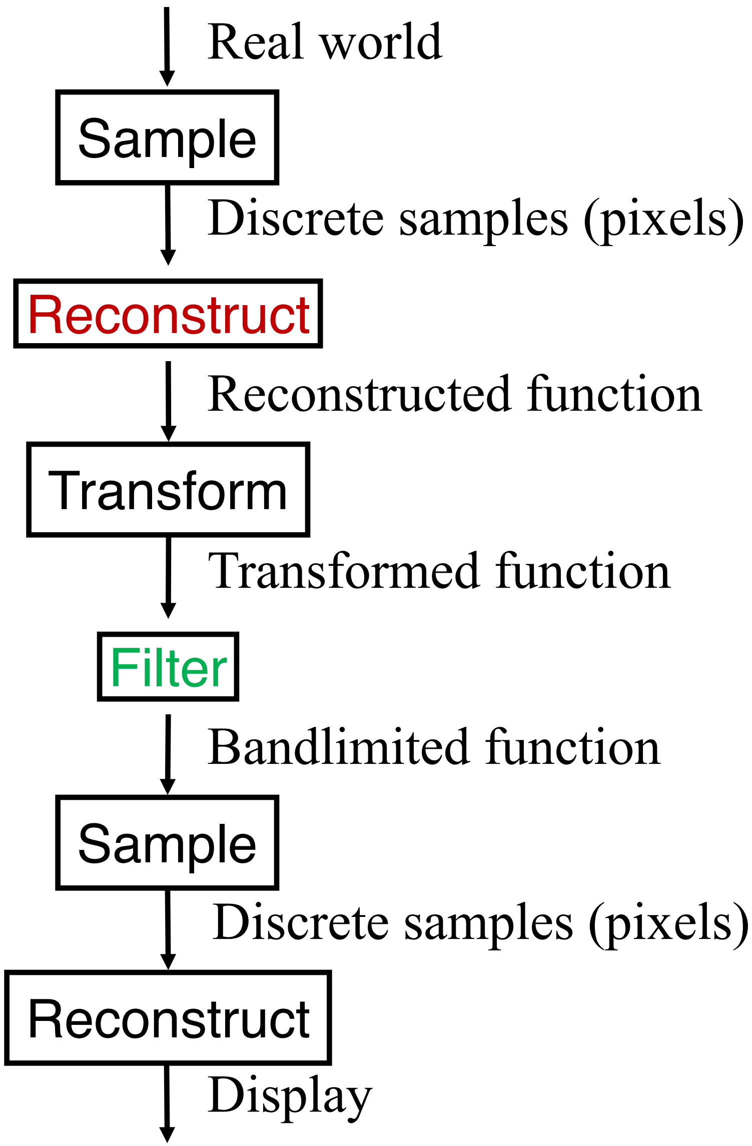
Image Processing





Display

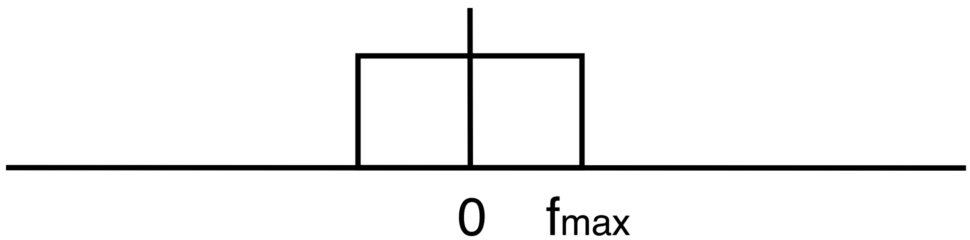
Image Processing



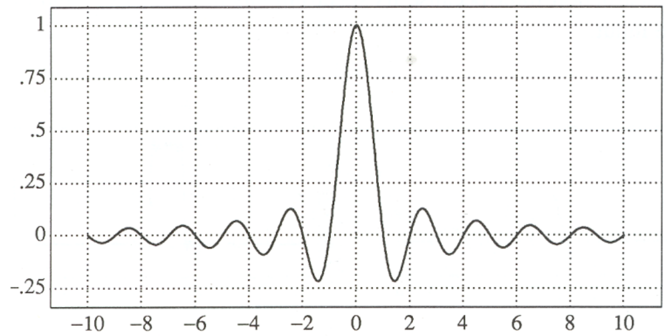
- Ideal resampling requires correct filtering to avoid artifacts
- Reconstruction filter especially important when magnifying
- Bandlimiting filter especially important when minifying

Ideal Image Processing Filter

- Frequency domain



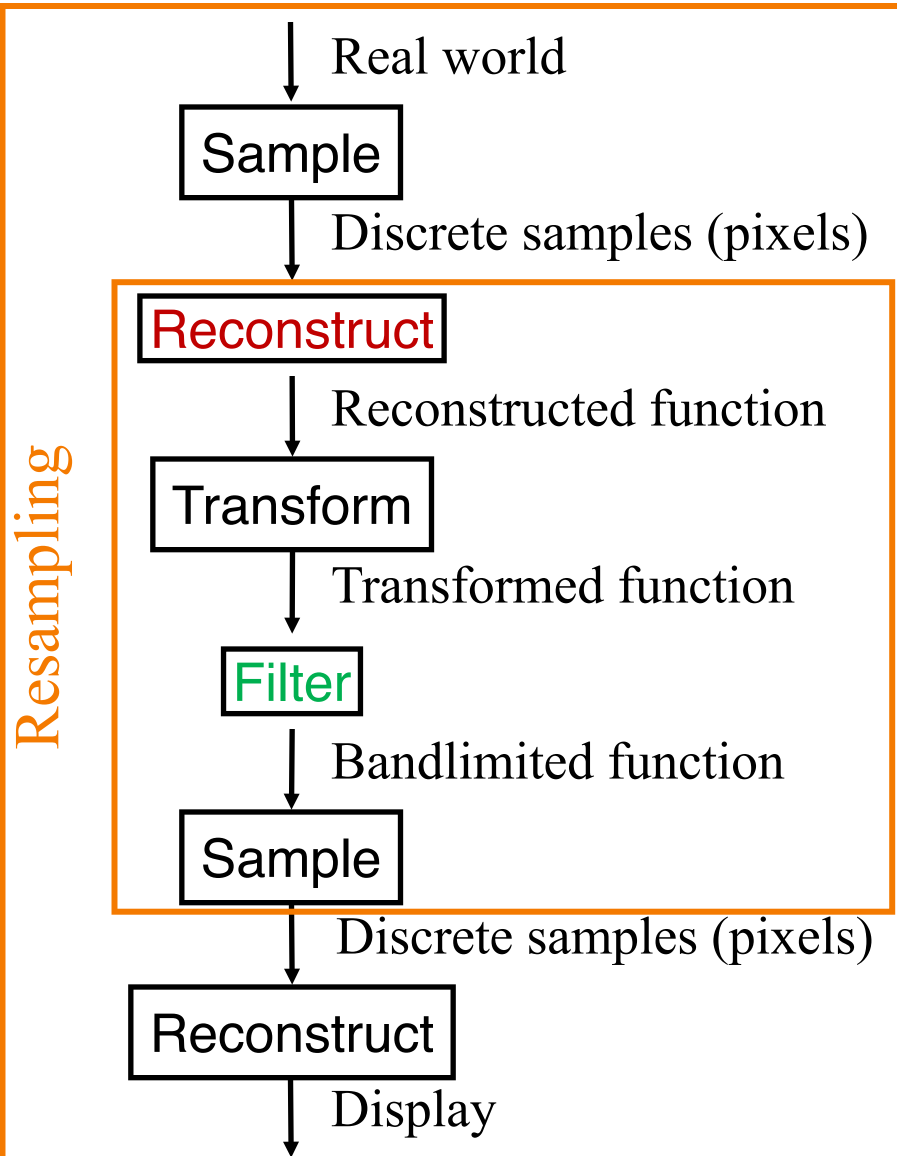
- Spatial domain



$$Sinc(x) = \frac{\sin \pi x}{\pi x}$$

Figure 4.5 Wolberg

Practical Image Processing

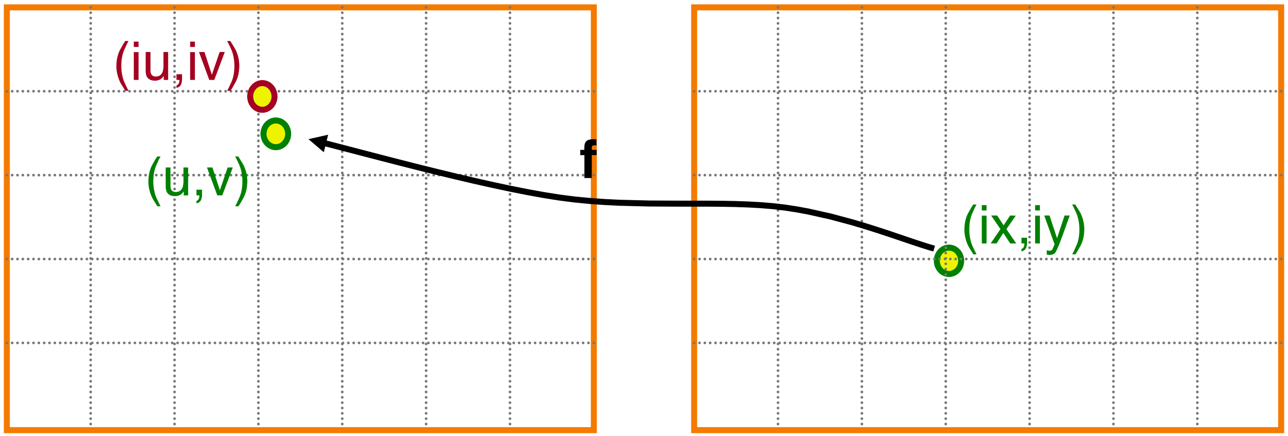


- Resampling: effectively (discrete) convolution to prevent artifacts
- Finite low-pass filters
  - Point sampling (bad)
  - Box filter
  - Triangle filter
  - Gaussian filter

Point Sampling

- Possible (poor) resampling implementation:

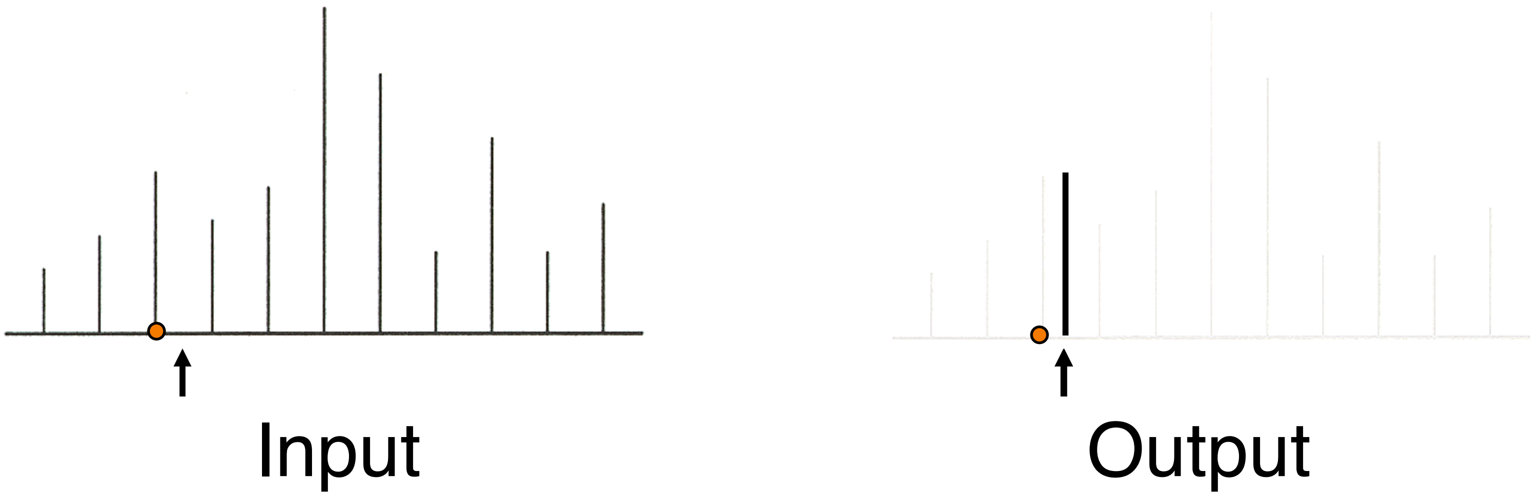
```
float Resample(src, u, v, k, w) {
   int iu = round(u);
   int iv = round(v);
   return src(iu,iv);
}
```



Source image
Destination image

Point Sampling

- Use nearest sample



Point Sampling

Point Sampled: Aliasing!

Correctly Bandlimited

Resampling with Filter

- Output is weighted average of inputs:

```
float Resample(src, u, v, k, w)
{
  float dst = 0;
  float ksum = 0;
  int ulo = u - w; etc.
  for (int iu = ulo; iu < uhi; iu++) {
    for (int iv = vlo; iv < vhi; iv++) {
      dst += k(u,v,iu,iv,w) * src(u,v)
      ksum += k(u,v,iu,iv,w);
    }
  }
  return dst / ksum;
}
```

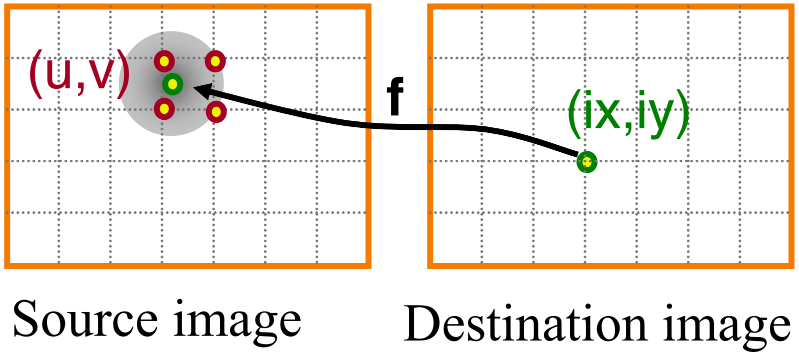
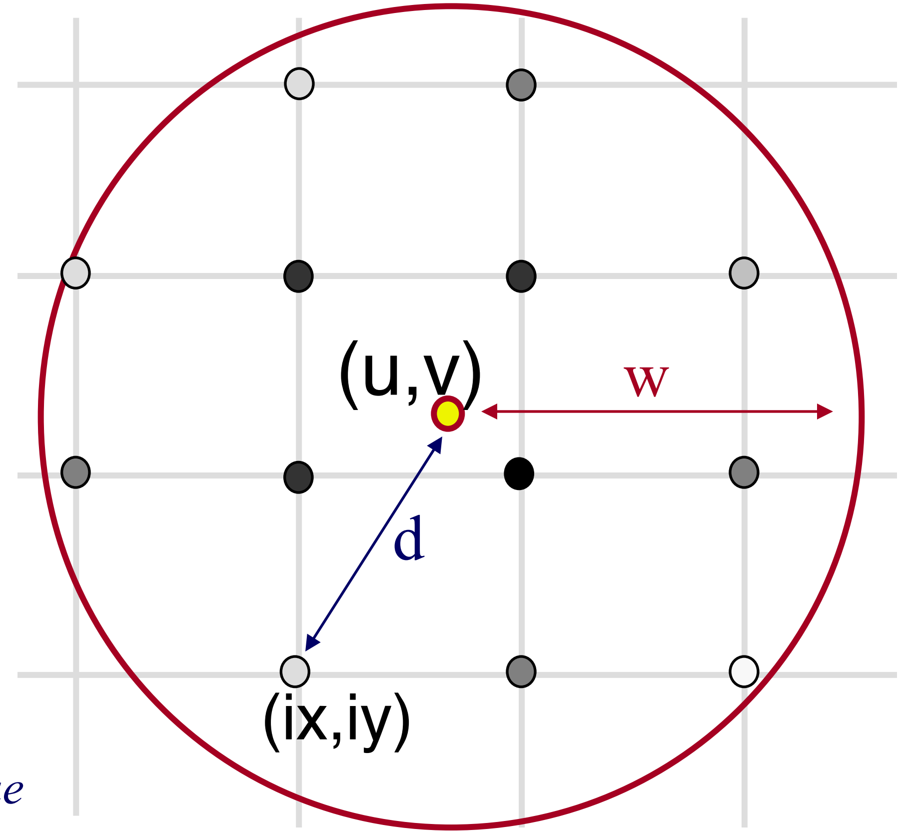


Image Resampling

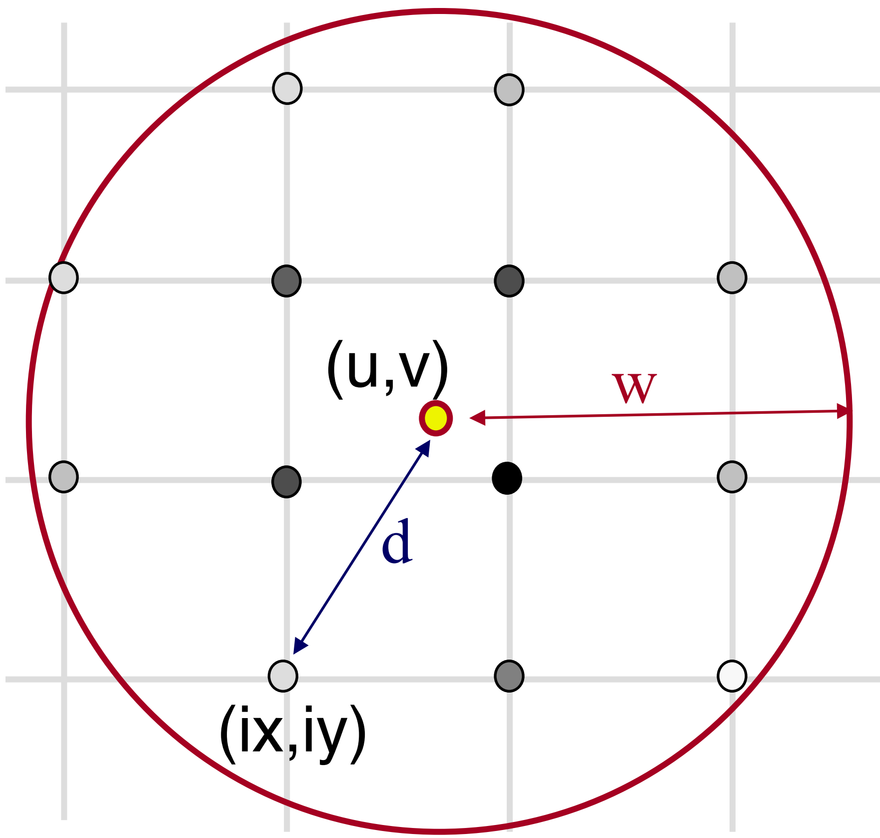
- Compute weighted sum of pixel neighborhood
  - Output is weighted average of input, where weights are normalized values of filter kernel (k)

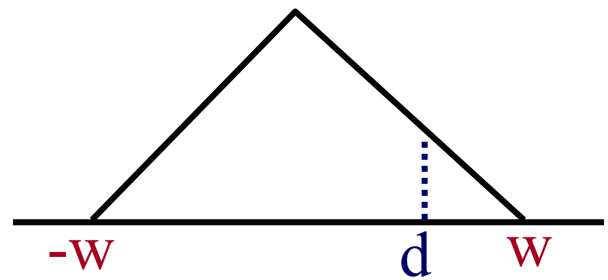


k(ix,iy) represented by gray value

Image Resampling

- For isotropic Triangle and Gaussian filters, k(ix,iy) is function of d and w





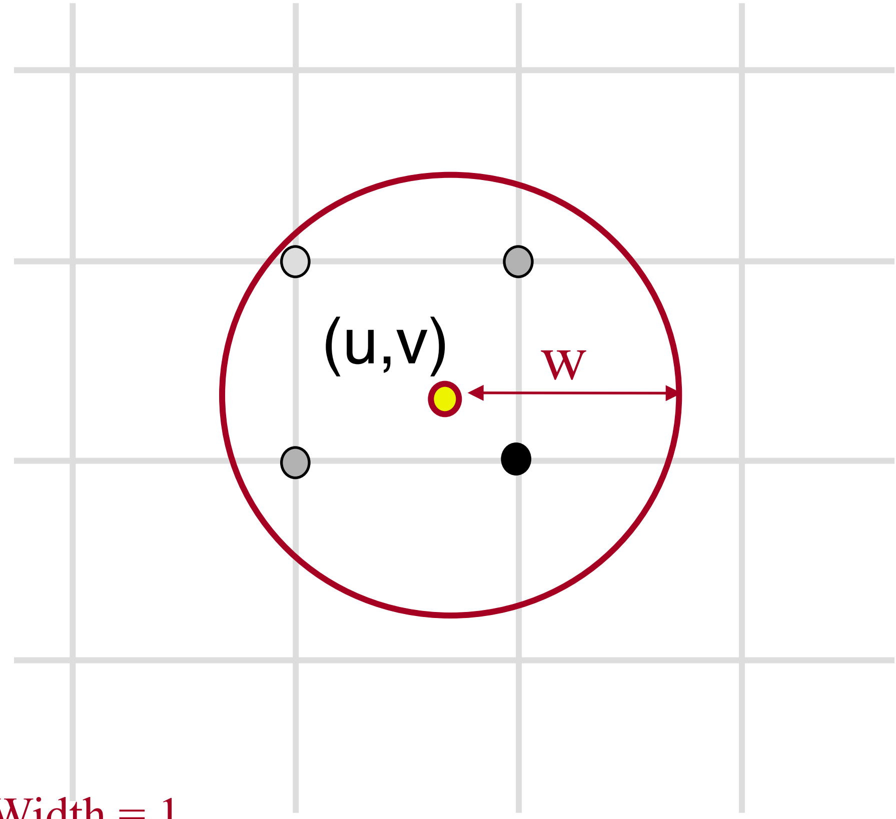
Triangle filter

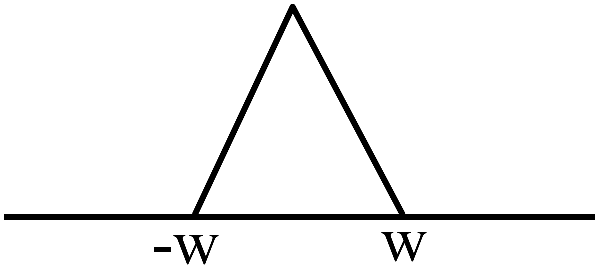
$$k(i,j)=\max(1 - d/w, 0)$$

Filter Width = 2

Image Resampling

- For isotropic Triangle and Gaussian filters, k(ix,iy) is function of d and w
  - Filter width chosen based on scale factor (or blur)





Triangle filter

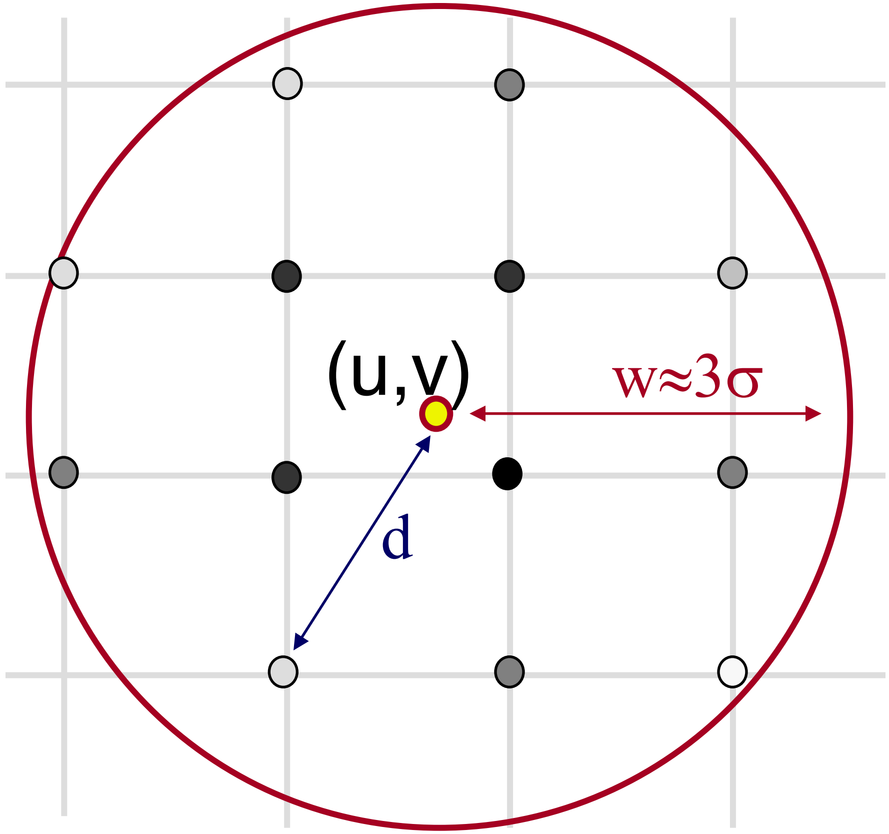
Width of filter affects blurriness

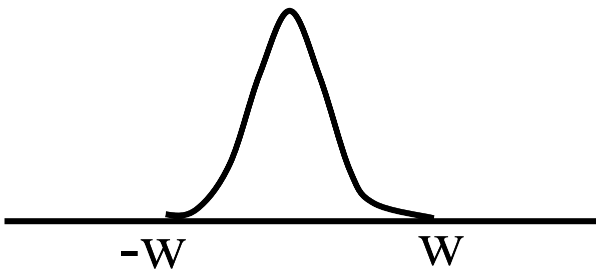
Filter Width = 1

Gaussian Filtering

- Kernel is Gaussian function

$$G(d,\sigma) = e^{-d^2/(2\sigma^2)}$$



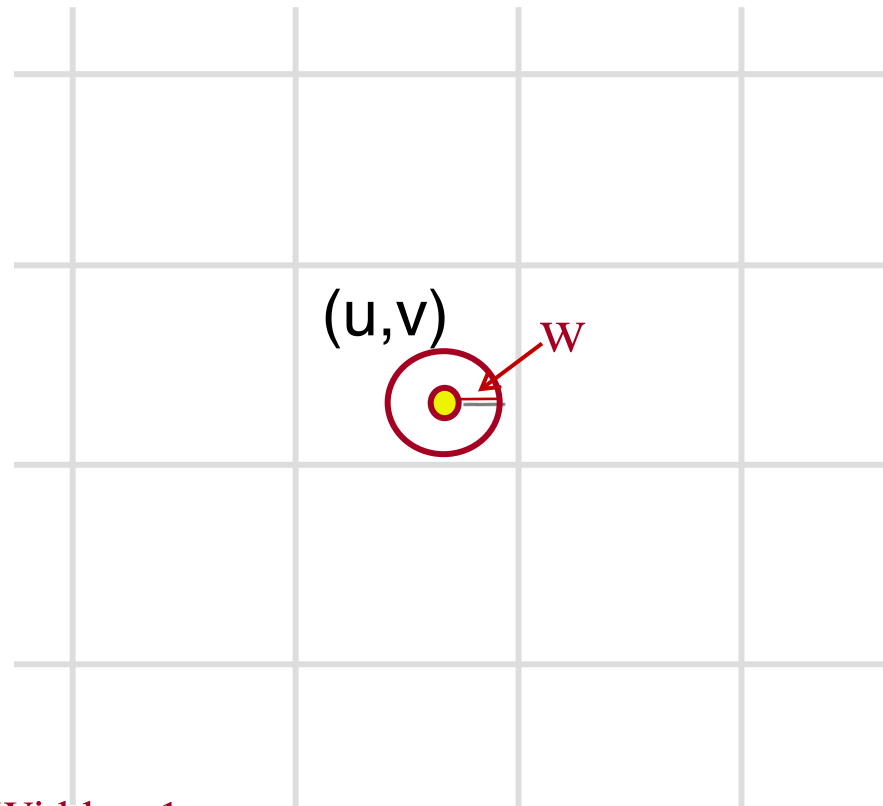


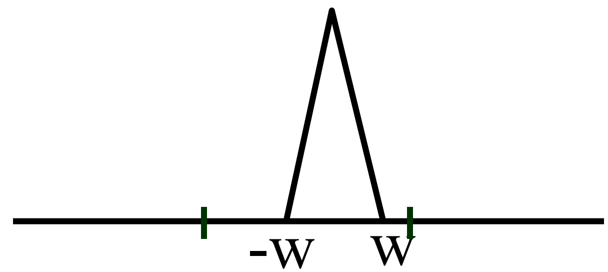
Gaussian Function

- Drops off quickly, but never gets to exactly 0
- In practice: compute out to w ~ 2.5σ or 3σ

Image Resampling

- What if width (w) is smaller than sample spacing?



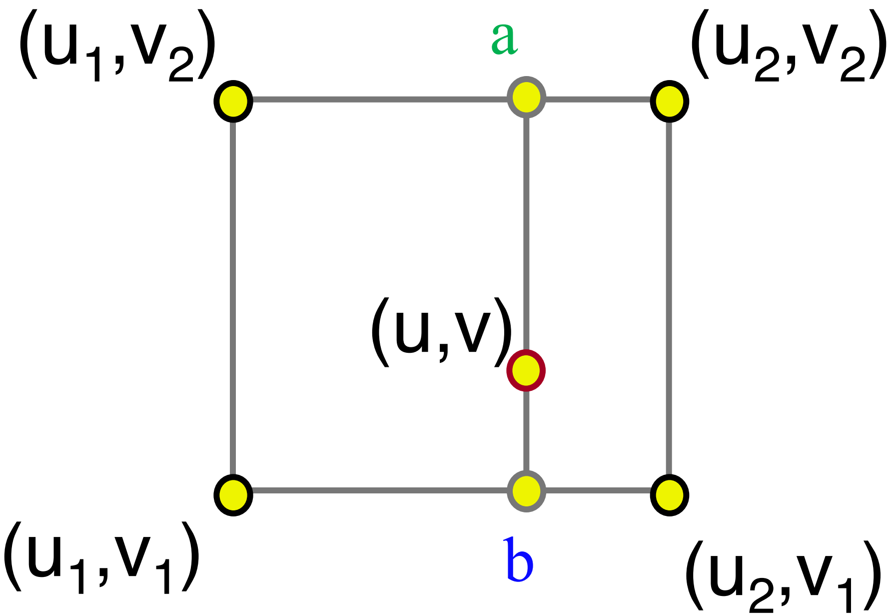


Triangle filter

Filter Width < 1

Image Resampling (with width < 1)

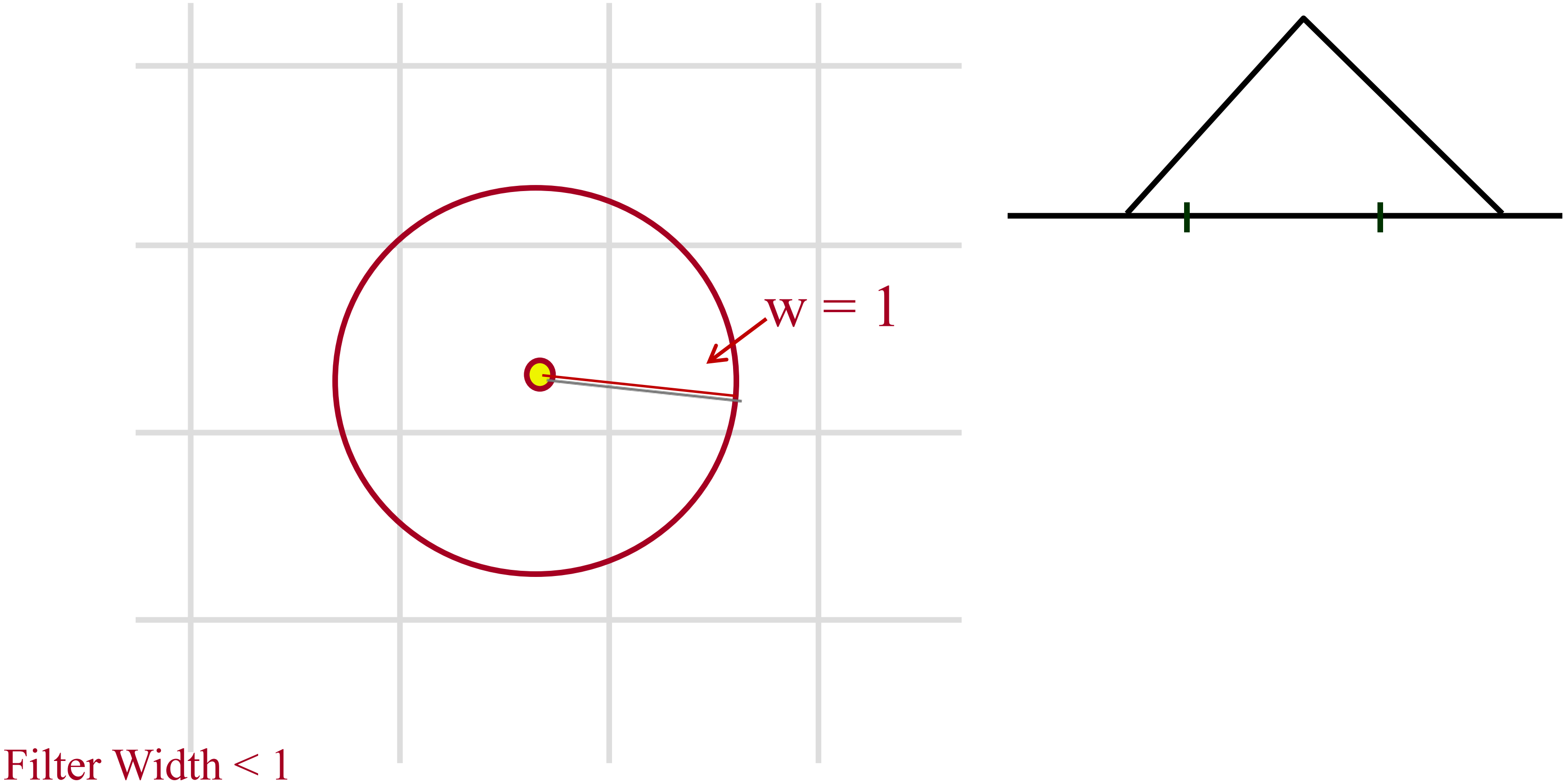
- Reconstruction filter: Bilinearly interpolate four closest pixels
  - a = linear interpolation of src(u_1,v_2) and src(u_2,v_2)
  - b = linear interpolation of src(u_1,v_1) and src(u_2,v_1)
  - dst(x,y) = linear interpolation of "a" and "b"



Filter Width < 1

Image Resampling (with width < 1)

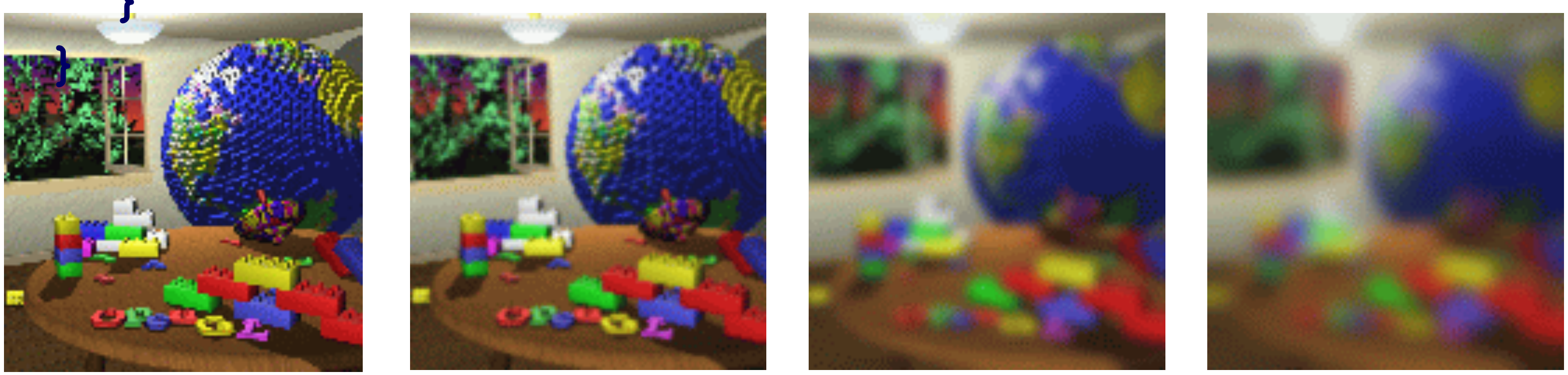
- Alternative: force width to be at least 1



Putting it All Together

- Possible implementation of image blur:

```
Blur(src, dst, sigma) {
  w ≈ 3*sigma;
  for (int ix = 0; ix < xmax; ix++) {
    for (int iy = 0; iy < ymax; iy++) {
      float u = ix;
      float v = iy;
      dst(ix,iy) = Resample(src,u,v,k,w);
    }
  }
}
```

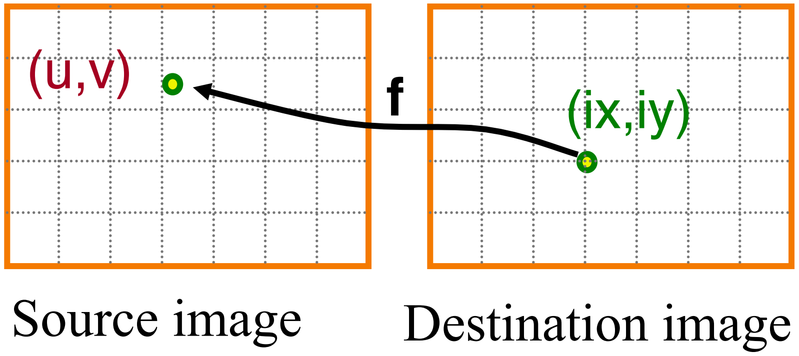


Increasing sigma

Putting it All Together

- Possible implementation of image scale:

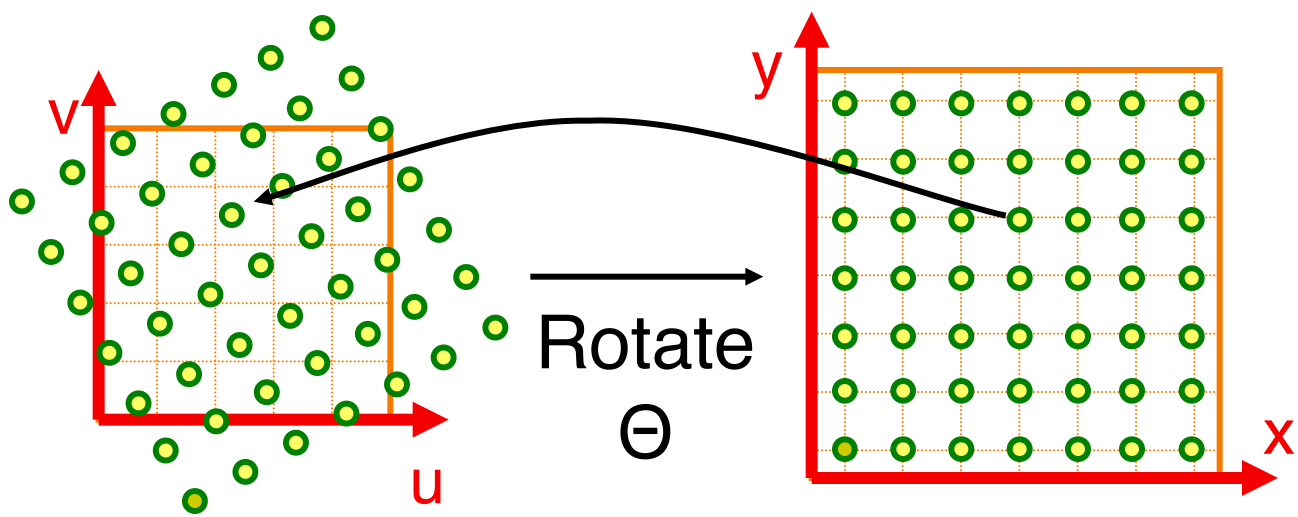
```
Scale(src, dst, sx, sy) {
  w ≈ max(1/sx,1/sy);
  for (int ix = 0; ix < xmax; ix++) {
    for (int iy = 0; iy < ymax; iy++) {
      float u = ix / sx;
      float v = iy / sy;
      dst(ix,iy) = Resample(src,u,v,k,w);
    }
  }
}
```



Putting it All Together

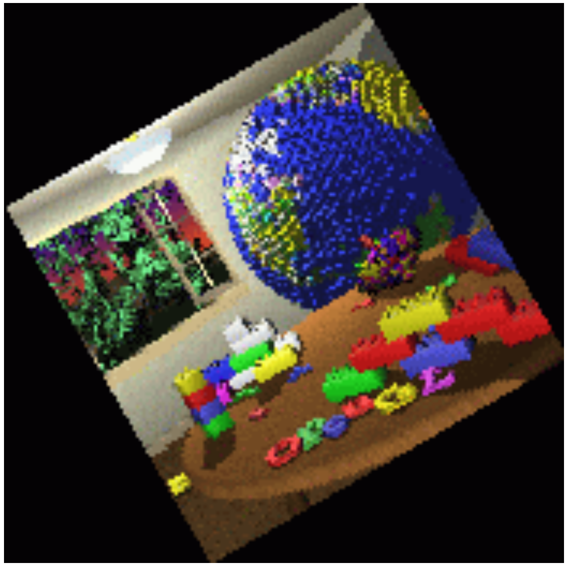
- Possible implementation of image rotation:

```
Rotate(src, dst, Θ) {
  w ≈ 1
  for (int ix = 0; ix < xmax; ix++) {
    for (int iy = 0; iy < ymax; iy++) {
      float u = ix*cos(-Θ) - iy*sin(-Θ);
      float v = ix*sin(-Θ) + iy*cos(-Θ);
      dst(ix,iy) = Resample(src,u,v,k,w);
    }
  }
}
```

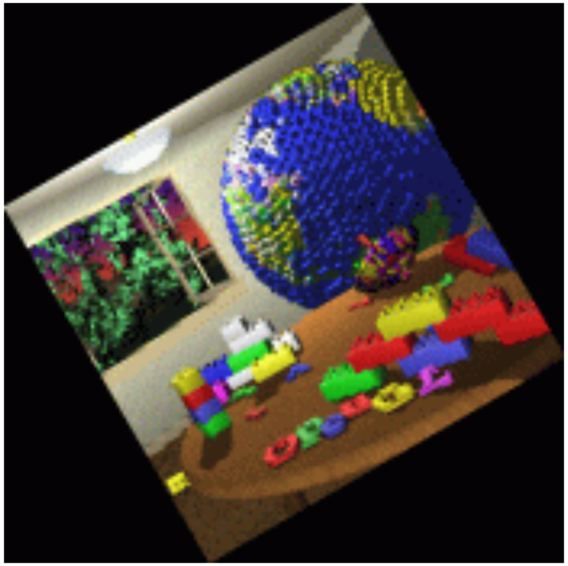


Sampling Method Comparison

- Trade-offs
  - Aliasing versus blurring
  - Computation speed



Point



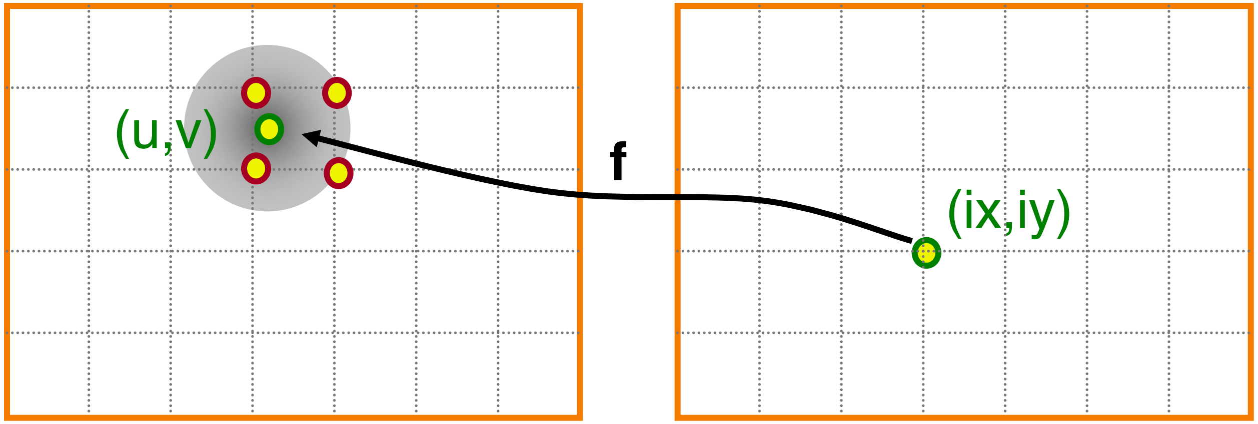
Triangle

Gaussian

Forward vs. Reverse Mapping

- Reverse mapping:

```
Warp(src, dst) {
  for (int ix = 0; ix < xmax; ix++) {
    for (int iy = 0; iy < ymax; iy++) {
      float w ≈ 1 / scale(ix, iy);
      float u = fx^-1(ix,iy);
      float v = fy^-1(ix,iy);
      dst(ix,iy) = Resample(src,u,v,w);
    }
  }
}
```



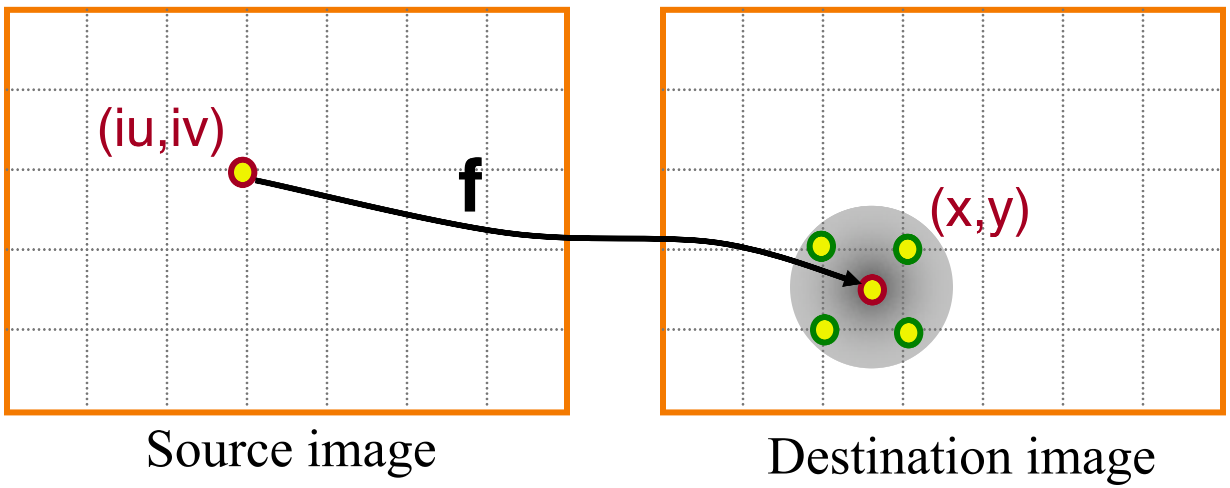
Source image

Destination image

Forward vs. Reverse Mapping

- Forward mapping:

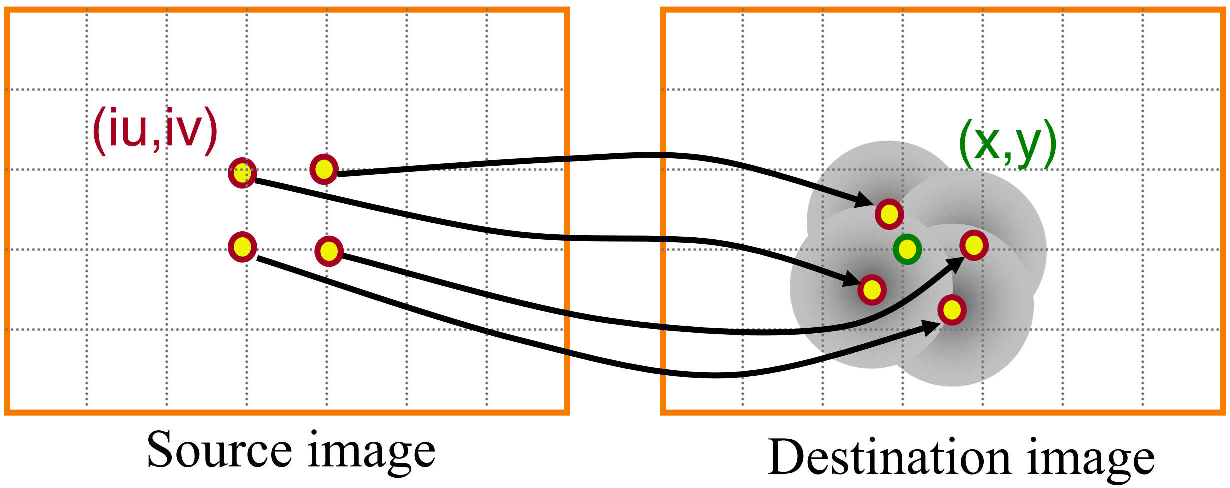
```
Warp(src, dst) {
  for (int iu = 0; iu < umax; iu++) {
    for (int iv = 0; iv < vmax; iv++) {
      float x = fx(iu,iv);
      float y = fy(iu,iv);
      float w ≈ 1 / scale(x, y);
      Splat(src(iu,iv),x,y,k,w);
    }
  }
}
```



Forward vs. Reverse Mapping

- Forward mapping:

```
Warp(src, dst) {
  for (int iu = 0; iu < umax; iu++) {
    for (int iv = 0; iv < vmax; iv++) {
      float x = fx(iu,iv);
      float y = fy(iu,iv);
      float w ≈ 1 / scale(x, y);
      Splat(src(iu,iv),x,y,k,w);
    }
  }
}
```

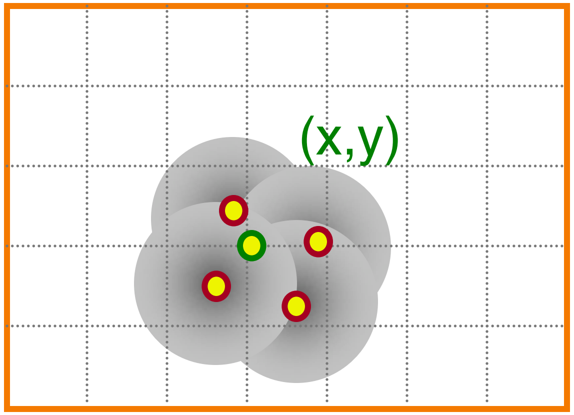


Forward vs. Reverse Mapping

- Forward mapping:

```
for (int iu = 0; iu < umax; iu++) {
  for (int iv = 0; iv < vmax; iv++) {
    float x = fx(iu,iv);
    float y = fy(iu,iv);
    float w ≈ 1 / scale(x, y);
    for (int ix = xlo; ix <= xhi; ix++) {
      for (int iy = ylo; iy <= yhi; iy++) {
        dst(ix,iy) += k(x,y,ix,iy,w) * src(iu,iv);
      }
    }
  }
}
```

Problem?

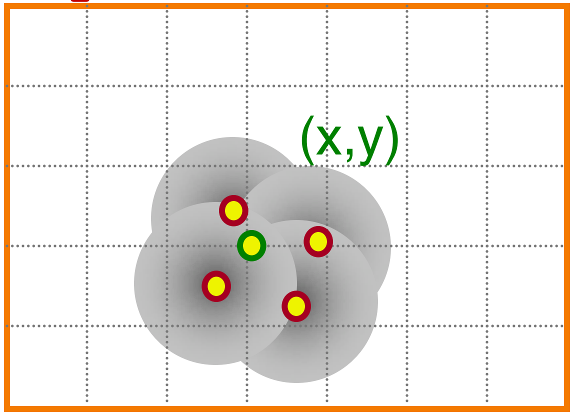


Destination image

Forward vs. Reverse Mapping

- Forward mapping:

```
for (int iu = 0; iu < umax; iu++) {
  for (int iv = 0; iv < vmax; iv++) {
    float x = fx(iu,iv);
    float y = fy(iu,iv);
    float w ≈ 1 / scale(x, y);
    for (int ix = xlo; ix <= xhi; ix++) {
      for (int iy = ylo; iy <= yhi; iy++) {
        dst(ix,iy) += k(x,y,ix,iy,w) * src(iu,iv);
        ksum(ix,iy) += k(x,y,ix,iy,w);
      }
    }
  }
}
for (ix = 0; ix < xmax; ix++)
  for (iy = 0; iy < ymax; iy++)
    dst(ix,iy) /= ksum(ix,iy)
```



Destination image

Forward vs. Reverse Mapping

- Tradeoffs?

Forward vs. Reverse Mapping

- Tradeoffs:
  - Forward mapping:
    - Requires separate buffer to store weights
  - Reverse mapping:
    - Requires inverse of mapping function, random access to original image

Summary

- Mapping
  - Forward vs. reverse
  - Parametric vs. correspondences
- Sampling, reconstruction, resampling
  - Frequency analysis of signal content
  - Filter to avoid undersampling: point, triangle, Gaussian
  - Reduce visual artifacts due to aliasing
    - » Blurring is better than aliasing

Next Time…

- Changing pixel values
  - Linear: scale, offset, etc.
  - Nonlinear: gamma, saturation, etc.
  - Histogram equalization
- Filtering over neighborhoods
  - Blur & sharpen
  - Detect edges
  - Median
  - Bilateral filter
- Moving image locations
  - Scale
  - Rotate
  - Warp
- Combining images
  - Composite
  - Morph
- Quantization
- Spatial / intensity tradeoff
  - Dithering