

Intro to Sampling Theory

Sampling Theory

- The world is continuous
- Like it or not, images are discrete.
 - We work using a discrete array of pixels
 - We use discrete values for color
 - We use discrete arrays and subdivisions for specifying textures and surfaces
- Process of going from continuous to discrete is called sampling.

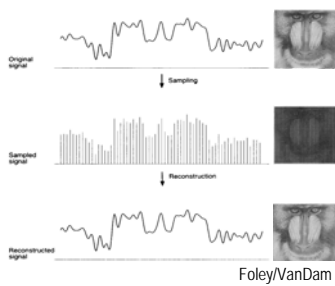
Sampling Theory

- Signal - function that conveys information
 - Audio signal (1D - function of time)
 - Image (2D - function of space)
- Continuous vs. Discrete
 - Continuous - defined for all values in range
 - Discrete - defined for a set of discrete points in range.

Sampling Theory

- Point Sampling
 - start with continuous signal
 - calculate values of signal at discrete, evenly spaced points (sampling)
 - convert back to continuous signal for display or output (reconstruction)

Sampling Theory



Sampling Theory

- Sampling can be described as creating a set of values representing a function evaluated at evenly spaced samples

$$f_n = f(i\Delta) \quad i = 0, 1, 2, \dots, n$$

Δ = interval between samples = range / n.

Sampling Theory

- Sampling Rate = number of samples per unit

$$f = \frac{1}{\Delta}$$

- Example -- CD Audio
 - sampling rate of 44,100 samples/sec
 - $\Delta = 1$ sample every 2.26×10^{-5} seconds

Issues:

- Important features of a scene may be missed
- If view changes slightly or objects move slightly, objects may move in and out of visibility.
- To fix, sample at a higher rate, but how high does it need to be?

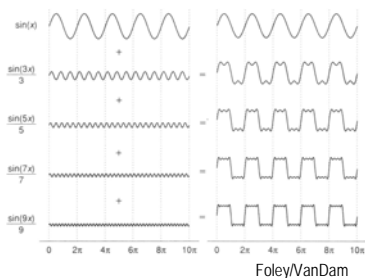
Sampling Theory

- Rich mathematical foundation for sampling theory
- Hope to give an “intuitive” notion of these mathematical concepts

Sampling Theory

- Spatial vs frequency domains
 - Most well behaved functions can be described as a sum of sin waves (possibly offset) at various frequencies
 - *Frequency spectrum* - a function by the contribution (and offset) at each frequency is describing the function in the frequency domain
 - Higher frequencies equate to greater detail

Sampling Theory



Sampling Theory

- Nyquist Theorem
 - A signal can be properly reconstructed if the signal is sampled at a frequency (rate) that is greater than *twice* the highest frequency component of the signal.
 - Said another way, if you have a signal with highest frequency component of f_h , you need at least $2f_h$ samples to represent this signal accurately.

Sampling Theory

- Example -- CD Audio
 - sampling rate of 44,100 samples/sec
 - $\Delta = 1$ sample every 2.26×10^{-5} seconds
- Using Nyquist Theorem
 - CDs can accurately reproduce sounds with frequencies as high as 22,050 Hz.

Sampling Theory

- Aliasing
 - Failure to follow the Nyquist Theorem results in *aliasing*.
 - Aliasing is when high frequency components of a signal appear as low frequency due to inadequate sampling.
- In CG:
 - Jaggies (edges)
 - Textures
 - Missed objects

Sampling Theory

- Aliasing - example



High frequencies masquerading as low frequencies

Sampling Theory

- Annoying Audio Applet
 - <http://ptolemy.eecs.berkeley.edu/eecs20/week13/aliasing.html>

Anti-Aliasing

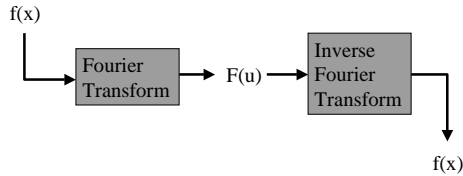
- What to do in an aliasing situation
 - Increase your sampling rate (supersampling)
 - Decrease the frequency range of your signal (Filtering)
- How do we determine the contribution of each frequency on our signal?

Fourier analysis

- Given $f(x)$ we can generate a function $F(u)$ which indicates how much contribution each frequency u has on the function f .
- $F(u)$ is the Fourier Transform
- Fourier Transform has an inverse

Sampling Theory

- Fourier Transforms



Sampling Theory

- The Fourier transform is defined as:

$$F(u) = \int_{-\infty}^{\infty} f(t)e^{-i2\pi ut} dt$$

Note: the Fourier Transform is defined in the complex plane

Sampling Theory

- The Inverse Fourier transform is defined as:

$$f(t) = \int_{-\infty}^{\infty} F(u)e^{i2\pi ut} du$$

Sampling Theory

- How do we calculate the Fourier Transform?
 - Use Mathematics
 - For discrete functions, use the Fast Fourier Transform algorithm (FFT)
- Can filter the transform to remove offending high frequencies - partial solution to anti-aliasing

Anti-aliasing -- Filtering

- Removes high component frequencies from a signal.
- Removing high frequencies results in removing detail from the signal.
- Can be done in the frequency or spatial domain

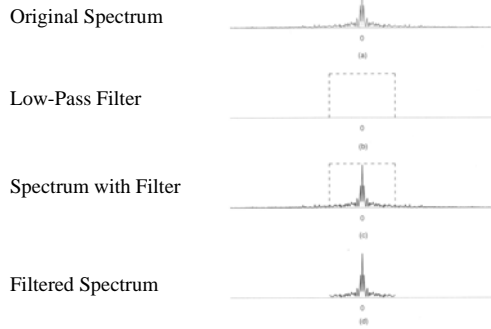
Getting rid of High Frequencies

- Filtering -- Frequency domain
 - Place function into frequency domain $F(u)$
 - Simple multiplication with *box filter* $S(u)$, aka *pulse function*, band(width) limiting or low-pass filter.

$$S(u) = \begin{cases} 1, & \text{when } -k \leq u \leq k \\ 0, & \text{elsewhere} \end{cases}$$

- Suppress all frequency components above some specified cut-off point k

Filtering – Frequency Domain



Foley/VanDam(631)

Getting Rid of High Frequencies

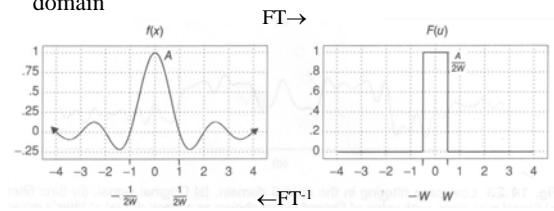
- Filtering -- Spatial Domain
 - Convolution (* operator) - equivalent to multiplying two Fourier transforms

$$h(x) = f(x) * g(x) = \int_{-\infty}^{\infty} f(\tau)g(x - \tau)d\tau$$

Taking a weighted average of the neighborhood around each point of f , weighted by g (the **convolution** or **filter kernel**) centered at that point.

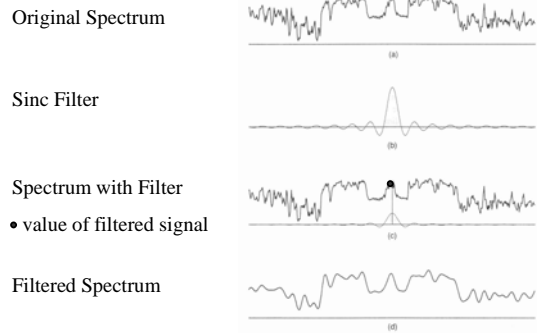
Convolution *sinc* Function

- Convoluting with a *sinc* function in the spatial domain is the same as using a box filter in the frequency domain



Foley/VanDam (634)

Filtering using Convolution



Foley/VanDam (633)

Convolution

- Joy of Convolution applet

<http://www.jhu.edu/~Esignals/convolve/index.html>

Sampling Theory

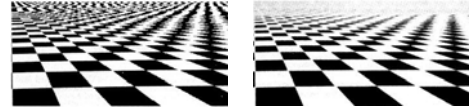
- Anti-aliasing -- Filtering
 - Removes high component frequencies from a signal.
 - Removing high frequencies results in removing detail from the signal.
 - Can be done in the frequency or spatial domain

Sampling Theory

- 2D Sampling
 - Images are examples of sampling in 2-dimensions.
 - 2D Fourier Transforms provides strength of signals at frequencies in the horizontal and vertical directions

Sampling Theory

- 2D Aliasing



aliased image

anti-aliased image

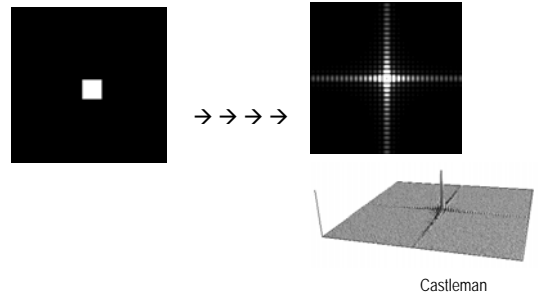
Foley/VanDam

Sampling Theory

- 2D Fourier Transform

$$F(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-i2\pi(ux+vy)} dx dy$$

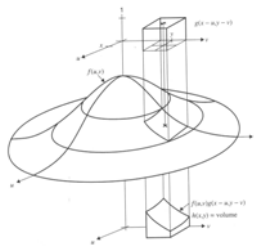
Sampling Theory



Castleman

Sampling Theory

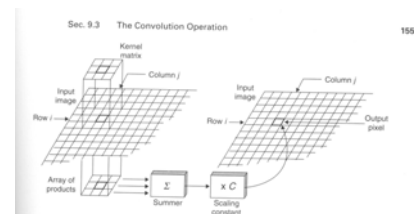
- Filtering - Convolution in 2D



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Sampling Theory

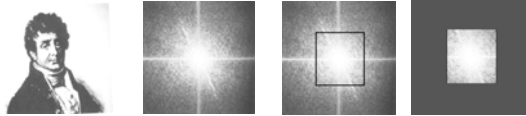
- Filtering – Convolution with images



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Sampling Theory

- Filtering – Convolution in frequency domain



Image

2D FFT

Filter out
high
frequencies

Filtered
2D FFT

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Anti-Aliasing – (Unweighted) Area Sampling

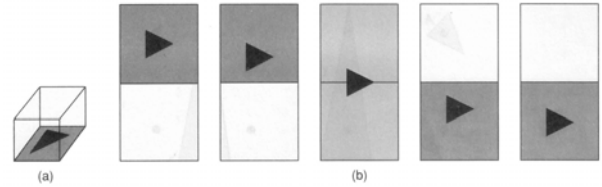


Fig. 14.11 Unweighted area sampling. (a) All points in the pixel are weighted equally. (b) Changes in computed intensities as an object moves between pixels.

Foley/VanDam (622)

Anti-Aliasing - Weighted Area Sampling

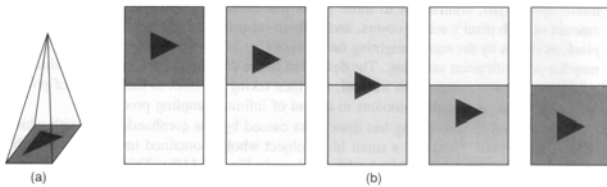


Fig. 14.12 Weighted area sampling. (a) Points in the pixel are weighted differently. (b) Changes in computed intensities as an object moves between pixels.

Foley/VanDam (622)

Anti-Aliasing - Weighted with Overlap

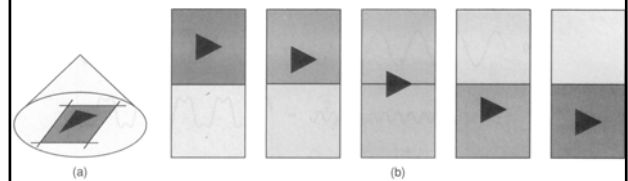
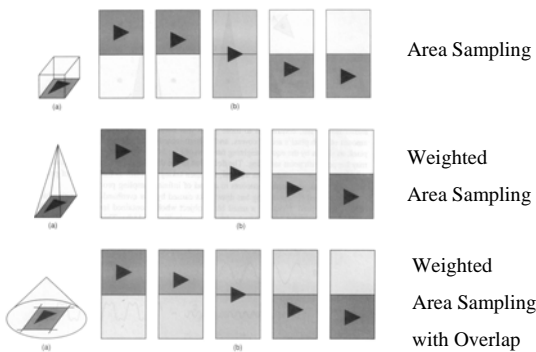


Fig. 14.13 Weighted area sampling with overlap. (a) Typical weighting function. (b) Changes in computed intensities as an object moves between pixels.

Foley/VanDam (622)



Area Sampling

Weighted
Area Sampling

Weighted
Area Sampling
with Overlap

Foley/VanDam (622/3)

Other Anti-aliasing Methods

- **Pre-filtering** - filtering at object precision before calculating pixel's sample
- **Post-filtering** - supersampling (as we've seen)
- **Adaptive supersampling** - sampling rate is varied, applied only when needed (changes, edges, small items)
- **Stochastic supersampling** - places samples at stochastically determined positions rather than regular grid

Anti-Aliasing

- Applet
<http://www.nbb.cornell.edu/neurobio/land/OldStudentProjects/cs490-96to97/anson/AntiAliasingApplet/index.html>

Sampling Theory

- Summary
 - Digital images are discrete with finite resolution...the world is not.
 - Spatial vs. Frequency domain
 - Nyquist Theorem
 - Convolution and Filtering
 - 2D Convolution & Filtering
- Questions?

Sampling Theory

- Further Reading
 - Foley/VanDam – Chapter 14
 - *Digital Image Processing* by Kenneth Castleman
 - Glassner, Unit II (Book 1)

Remember

- Class Web Site:
 - <http://www.cs.rit.edu/~jmg/cgII>
- Any questions?