

Aliasing

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Aliasing arises when a signal is discretely sampled at a rate that is insufficient to capture the changes in the signal. Consider the following contexts in which signals are discretely sampled:

- Retinal images are sampled in space by photoreceptors.
- Film and video are sampled in time by discrete frames.
- Sound is commonly digitally sampled for recording and communications.

Aliasing may arise in all of these situations if sampling is done improperly. Because the effects of aliasing can be rather disastrous, it is important to understand why aliasing occurs, what its consequences are, and how it may be avoided.

1 Nyquist sampling theorem

The Nyquist sampling theorem provides a prescription for the nominal sampling interval required to avoid aliasing. It may be stated simply as follows:

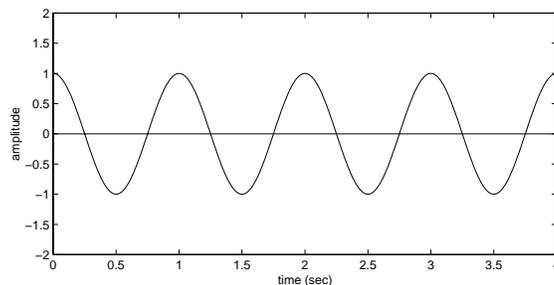
The sampling frequency should be at least twice the highest frequency contained in the signal.

Or in mathematical terms:

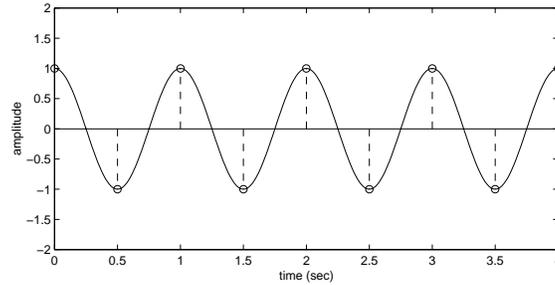
$$f_s \geq 2 f_c \quad (1)$$

where f_s is the sampling frequency (how often samples are taken per unit of time or space), and f_c is the highest frequency contained in the signal.

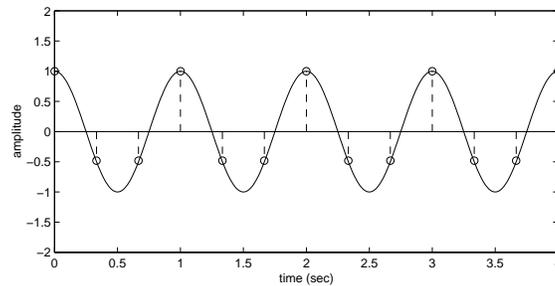
That this is so is really quite intuitive. Consider for example a signal composed of a single sinewave at a frequency of 1 Hz:



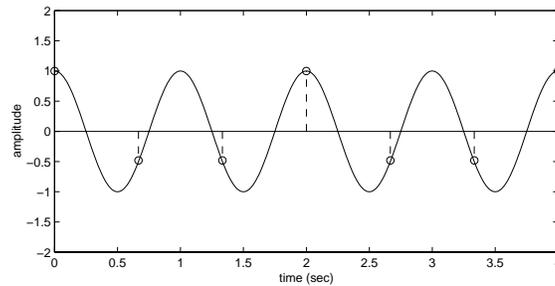
If we sample this waveform at 2 Hz (as dictated by the Nyquist theorem), that is sufficient to capture each peak and trough of the signal:



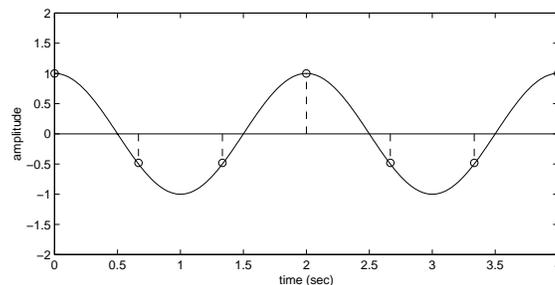
If we sample at a frequency higher than this, for example 3 Hz, then there are *more* than enough samples to capture the variations in the signal:



If however we sample at a frequency lower than 2 Hz, for example at 1.5 Hz, then there are now not enough samples to capture all the peaks and troughs in the signal:

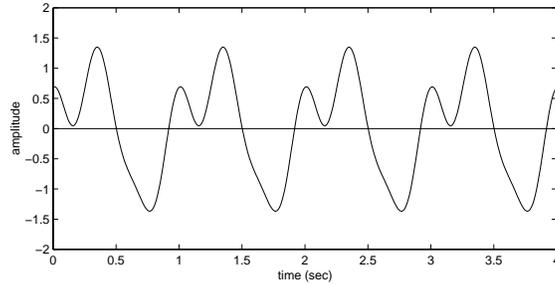


Note here that we are not only losing information, but we are getting the *wrong* information about the signal. The person receiving these samples, without any previous knowledge of the original signal, may well be misled into thinking that the signal has quite a different form:

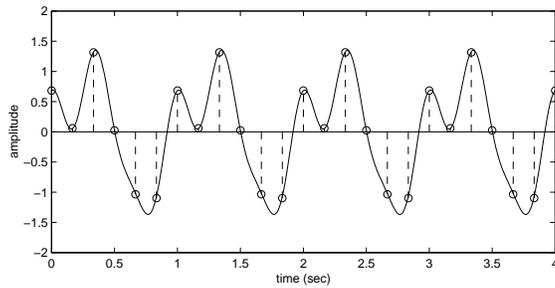


From this example, we can see the reason for the term *aliasing*. That is, the signal now takes on a different “persona,” or a false presentation, due to being sampled at an insufficiently high frequency.

Now we are ready to think about the sampling of a complex signal composed of many frequency components. By Fourier’s theorem, we know that any continuous signal may be decomposed in terms of a sum of sines and cosines at different frequencies. For example, the following waveform was composed by adding together sinewaves at frequencies of 1 Hz, 2 Hz, and 3 Hz:



According to the Nyquist sampling theorem, the signal must be sampled at twice the highest frequency contained in the signal. In this case, we have $f_c=3$ Hz, and so the Nyquist theorem tells us that the sampling frequency, f_s , must be at least 6 Hz. And sure enough, this appears to be sufficient:



Thus, when a signal contains not just one but many different frequencies added together, the minimal sampling rate needed to avoid aliasing is just twice whatever the highest frequency is, irrespective of how many other frequency components there are.

2 Effects of aliasing

2.1 The wagon wheel effect

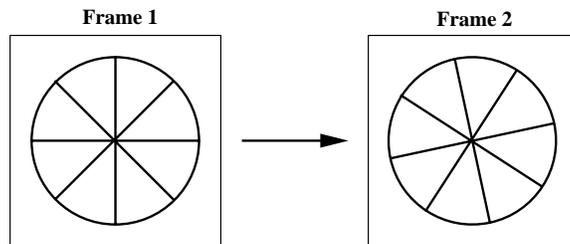
One common situation in which aliasing occurs is in film. This is because continuously varying images are being discretely sampled at a rate of 24 frames/sec. The Nyquist sampling theorem tells us that aliasing will occur if at any point in the image plane there are frequency components, or light-dark transitions, that occur faster than $f_s/2$, which in this case is 12 frames/sec. But in many situations the light-dark transitions may be occurring faster than this—such as a wagon wheel or propeller rotating at high speed.

Consider a wagon wheel with eight spokes turning at a rate of 3 revolutions per second (or 180 rpm) in the clockwise direction. In this case, the wagon wheel will move by one spoke per frame, since

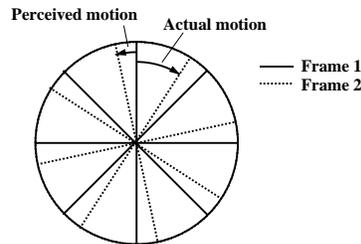
$$\frac{(3 \text{ revolutions/sec}) \times (8 \text{ spokes/revolution})}{(24 \text{ frames/sec})} = 1 \text{ spoke/frame} \quad (2)$$

Thus, the wagon wheel will appear to stand still! One rarely sees this happen though, because you have to be awfully lucky for the wheel to be turning at just this speed.

Consider though what would happen if the wagon wheel were turning at a rate just below this, say at 2.5 revolutions/sec. Now the wheel will move by 83% of the spoke spacing each frame. So, comparing two adjacent frames of the film we would see the following:

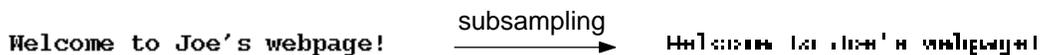


The problem the brain faces in viewing these frames in rapid succession is that there are two interpretations. One interpretation is that the wheel has moved by 83% of the spoke interval in the clockwise direction. The alternative interpretation is that it has moved by 17% of the spoke interval in the counter-clockwise direction. It turns out that the brain prefers the latter interpretation, and so as a result you perceive the wheel moving backwards (counter-clockwise) at a slower speed than it is actually moving:



2.2 Text images

Another place aliasing is often seen is in images of text. This happens for example when people put an image on their web page that contains some text characters that has been improperly subsampled. Here is a simple example:



The image on the right contains half as many samples in each dimension (horizontal and vertical), and as a result it is now for all practical purposes illegible. In the next section we will see a simple way to avoid such disastrous effects of aliasing.

3 How to avoid aliasing

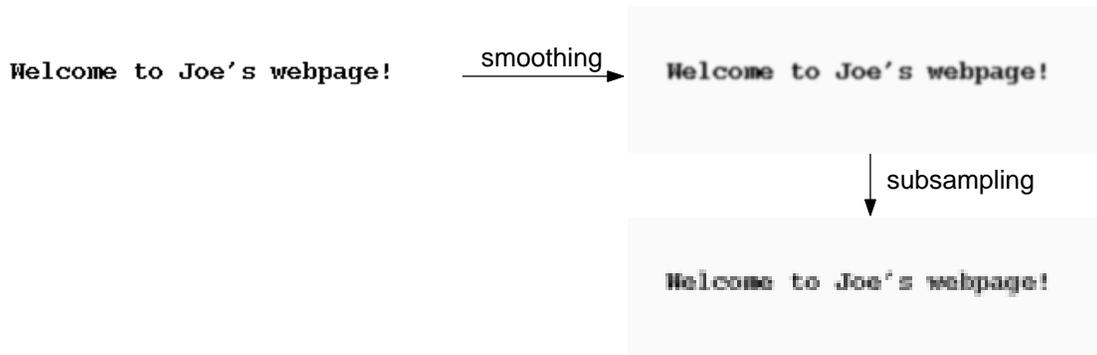
The simple way to avoid aliasing of course is to always have enough samples to capture the spatial or temporal variations in a signal. But there are many situations in which the sample rate may be limited—for example, film is limited to 24 frames/sec, or you may be designing a web page where the images need to be a minimum number of pixels so they download quickly. So, if the sampling rate f_s is fixed, then there is only one other parameter we can play with, and that is the highest frequency contained in the signal, f_c . This can be done by lowpass filtering, or in layman’s terms, “smoothing.”

The easiest way to smooth a signal is to blur it. In film this is easily done by adjusting the lens so the image is a bit out of focus. The more out of focus it is, the lower the spatial-frequency content of the image becomes. And as the spatial-frequency content is lowered, the temporal frequency content is also lowered. Blurring is often used by film directors to eliminate aliasing from a scene. For example, when filming a shot where a car is racing by, it would be a mistake to keep the car in focus. The image of the car is usually purposefully de-focussed so as to eliminate the jerky appearance of motion, which breaks the illusion of reality in film. Oftentimes though, directors will not bother to eliminate aliasing in the form of backwards moving wheels, or they may even prefer to keep it for “visual effect.”

Blurring may be easily accomplished on digital images by filtering (or convolving) with a mask that averages together the values of several neighboring pixels. Such an operation is usually provided in image manipulation programs such as Photoshop.

3.1 Text images revisited

Now that we know how to avoid aliasing, we can form a properly subsampled version of our text image by blurring the image a bit before subsampling:

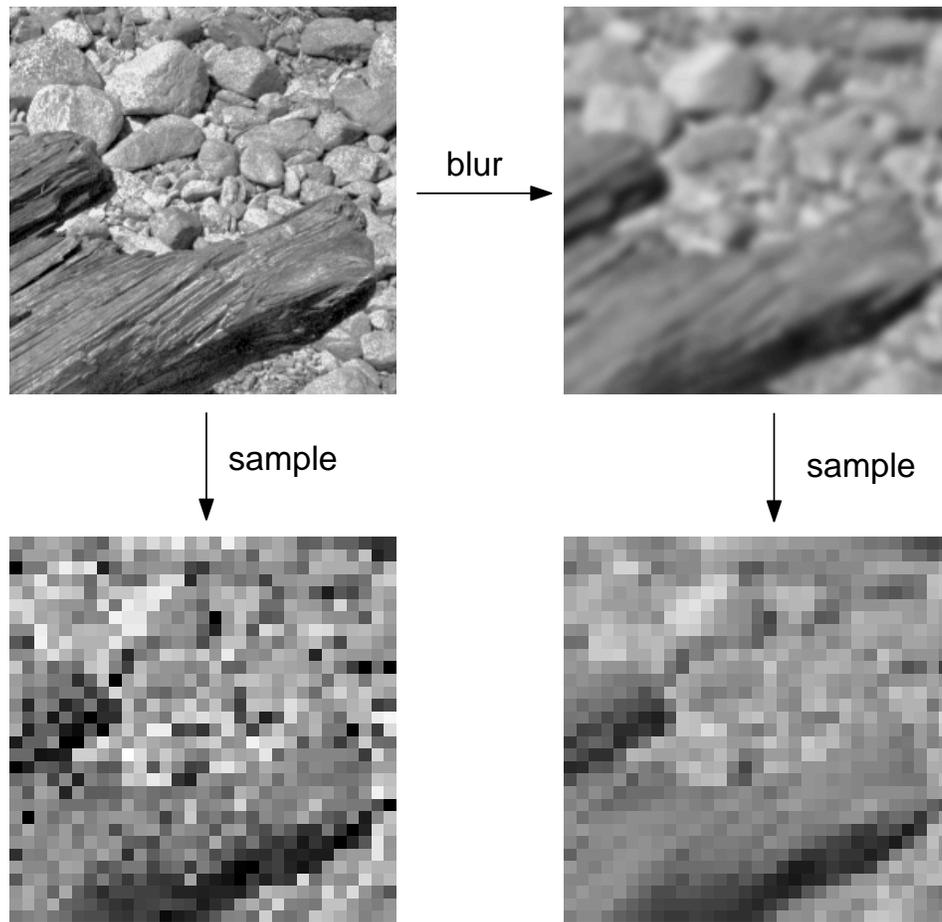


Note that the subsampled text image is now at least legible, yet it has the same number of pixels as the image in the previous example. This technique is also known

as “anti-aliasing” in the computer graphics community.

3.2 Natural images

The avoidance of aliasing is also important in the design of the eye. For example, in the human eye the lens filters out any spatial variations finer than 60 cycles/degree. The Nyquist theorem tells us in this case that the photoreceptor spacing should be at least 120 sample/degree, which is exactly what we have! Consider though what the sampled retinal image might look like if this were not the case:



At upper left is shown a natural image that may typically fall on the retina, and below this is shown the result of subsampling the image without properly blurring beforehand. Compare this to the sampled version after blurring (lower right). It is not difficult to see why nature has gone to the effort it has to match photoreceptor spacing with the point-spread function (spatial-frequency cutoff) of the lens.