EE4C5 Digital Signal Processing

Lecture 5 – Decimation and Interpolation

This lecture

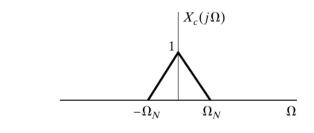
- Based on Chapter 4 of O&S
- All images from O&S book unless otherwise stated

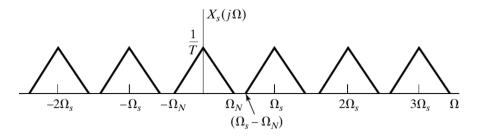
Recall...

• The impulse train approximation $x_s(t)$ of the original continuous time signal $x_c(t)$ has (continuous) Fourier Transform:

•
$$X_S(j\Omega) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_C(j(\Omega - k\Omega_S))$$

- With T the sampling period, Ω is the continuous-time frequency in radians/s, and Ω_s (= $2\pi f_s$) is the angular sampling frequency in radians/s
- Saw that $X_s(j\Omega)$ comprises copies of $X_c(j\Omega)$ shifted by $k\Omega_s$ and scaled by $\frac{1}{T}$





Pay attention – discrete or continuous FT

- Can show that (how?)
 - $X_s(j\Omega) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\Omega nt}$
- The DTFT gives
 - $X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n}$
- So:
 - $X_S(j\Omega) = X(e^{j\omega})\Big|_{\omega=\Omega T} = X(e^{j\Omega T})$
- Thus the DTFT $X(e^{j\omega})$ is a frequency scaled version of $X_s(j\Omega)$, with the scaling specified by $\omega=\Omega T$
- The above relates the CTFT to the DTFT

Useful for later...

• Combining we get:

•
$$X(e^{j\Omega T}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c(j(\Omega - k\Omega_s))$$

Or equivalently

•
$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right)$$

Resampling a discrete-time sequence

- A continuous time signal $x_c(t)$ can be represented by the discrete sequence x[n] where
 - $x[n] = x_c(nT), -\infty < n < \infty$
- Often wish to change the sampling rate of x[n] such that
 - $x_1[n] = x_c(nT_1), T_1 \neq T$
- This is resampling
- Could we reconstruct original $x_c(t)$ and use new period T_1 ?
 - Not practical
 - Restrict to discrete time operations

- $x_d[n] = x[nM] = x_c(nMT)$
- With $X_c(j\Omega)$ bandlimited to $\pm\Omega_N$ then $x_d[n]$ will be exact representation if $\pi/T_d \geq \Omega_N$ or equivalently $\pi/T_M \geq \Omega_N$
- Can avoid aliasing if:
 - Original sampling rate is a least M times the Nyquist rate

• Bandwidth of the sequence is first reduced by factor M by discrete-time filtering

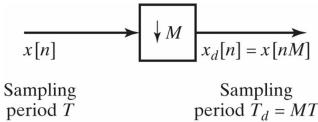


Figure 4.19 Representation of a compressor or discrete-time sampler.

• Using
$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right)$$
 (slide #5)

- DTFT of $x_d[n] = x[nM] = x_c(nMT)$
 - $X_d(e^{j\omega}) = \frac{1}{T_d} \sum_{r=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{T_d} \frac{2\pi r}{T_d} \right) \right)$
- But with $T_d = MT$ can rewrite:
 - $X_d(e^{j\omega}) = \frac{1}{MT} \sum_{r=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{MT} \frac{2\pi r}{MT} \right) \right)$

• Using
$$X(e^{j\omega}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{T} - \frac{2\pi k}{T} \right) \right)$$
 (slide #5)

• DTFT of
$$x_d[n] = x[nM] = x_c(nMT)$$

•
$$X_d(e^{j\omega}) = \frac{1}{T_d} \sum_{r=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{T_d} - \frac{2\pi r}{T_d} \right) \right)$$

• But with $T_d = MT$ can rewrite:

•
$$X_d(e^{j\omega}) = \frac{1}{MT} \sum_{r=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{MT} - \frac{2\pi r}{MT} \right) \right)$$

• The summation index r is related as

•
$$r = i + kM$$

• k and i integers, $-\infty < k < \infty$ and $0 \le i \le M-1$

Want to relate back

•
$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} \left\{ \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j \left(\frac{\omega}{MT} - \frac{2\pi k}{T} - \frac{2\pi i}{MT} \right) \right) \right\}$$

Recognising that

•
$$X(e^{j(\omega-2\pi i/M)}) = \frac{1}{T} \sum_{k=-\infty}^{\infty} X_c \left(j\left(\frac{\omega-2\pi i}{MT} - \frac{2\pi k}{T}\right) \right)$$

Can arrive at:

•
$$X_d(e^{j\omega}) = \frac{1}{M} \sum_{i=0}^{M-1} X(e^{j(\omega - 2\pi i/M)})$$

- Expresses FT of $x_d[n]$ in terms of FT of x[n]
- $X_d(e^{j\omega})$ consists of copies of $X(e^{j\omega})$ scaled by $\frac{1}{M}$, and frequency scaled by $\frac{1}{M}$ and shifted by $2\pi i$.

Avoid aliasing...

- $X_d(e^{j\omega})$ periodic with period 2π (like all DTFT!!)
- Can avoid aliasing if $X(e^{j\omega})$ is bandlimited, i.e.
- $X(e^{j\omega}) = 0$ $\omega_N \le |\omega| \le \pi$
- And $^{2\pi}/_{M} \geq 2\omega_{N}$
- Example shown opposite has M=2, at the limit to avoid aliasing here

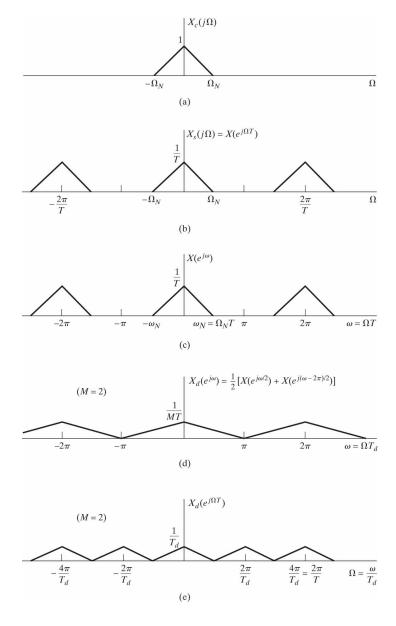


Figure 4.20 Frequency-domain illustration of downsampling.

Aliasing with M=3

- Will get aliasing with M=3
- Prefilter to avoid this and bandlimit the signal

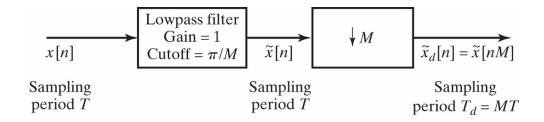


Figure 4.22 General system for sampling rate reduction by M.

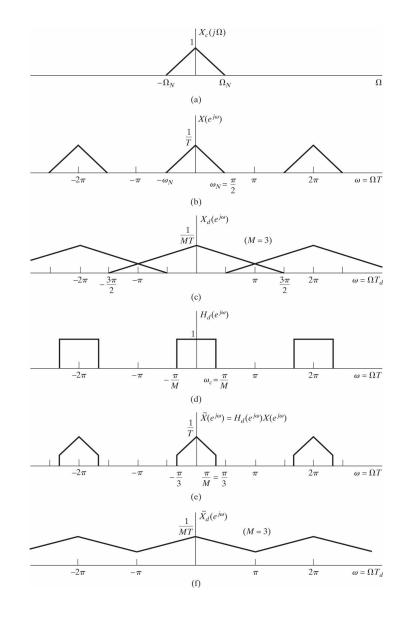


Figure 4.21 (a)–(c) Downsampling with aliasing. (d)–(f) Downsampling with prefiltering to avoid aliasing.

Increase sampling rate by integer factor

- Increase sampling rate of x[n] by integer factor L
- Want
 - $x_i[n] = x_c(nT_i)$, $T_i = T/L$
- Where you can take the samples from
 - $x[n] = x_c(nT)$
- This is upsampling
 - $x_i[n] = x[n/L] = x_c(n^T/L), \quad n = 0, \pm L, \pm 2L, ...$

System

- Note distinction between $x_i[n]$ and $\chi_{\rho}|n|$
- $x_e[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-kL]$

•
$$x_e[n] = \begin{cases} x[^n/_L], & n = 0, \pm L, \pm 2L, \dots \\ 0, & otherwise \end{cases}$$

- Expander
- LPF the reconstructs the sequence

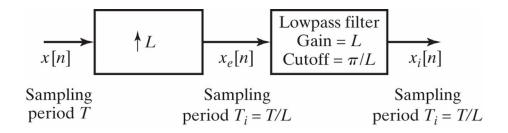
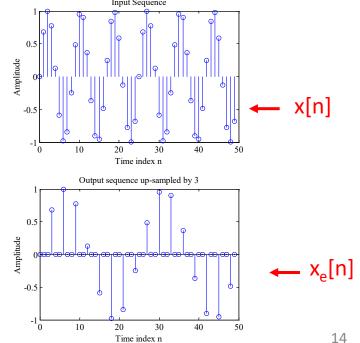


Figure 4.23 General system for sampling rate increase by L.



Increase sampling rate by integer factor

• Frequency domain interpretation, can epxpress the FT as:

•
$$X_e(e^{j\omega}) = \sum_{n=-\infty}^{\infty} (\sum_{k=-\infty}^{\infty} x[k]\delta[n-kL]) e^{-j\omega n}$$

$$= \sum_{k=-\infty}^{\infty} x[k] e^{-j\omega Lk} = X_e(e^{j\omega L})$$

- So FT of output of expander is a frequency scaled version of FT of the input, $\omega \to \omega L$ with ω now normalised by
- $\omega = \Omega T_i$

Get the desired signal

- $X_i(e^{j\omega})$ can be obtained from $X_e(e^{j\omega})$ by correcting the amplitude scaling from $^1/_T$ to $^1/_{T_i}$ and removing all frequency-scaled images of $X_c(j\Omega)$ except at integer multiples of 2π
- This requires a lowpass filter, gain of L and cut-off frequency π/L

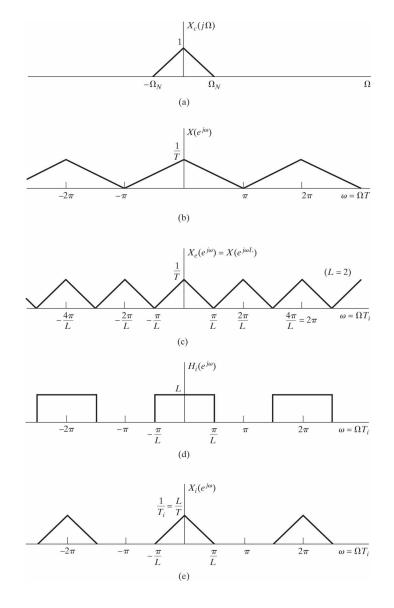


Figure 4.24 Frequency-domain illustration of interpolation.

Interpolation

- Can view this as interpolation in the time domain
- Impulse response of the LPF is

•
$$h_i[n] = \frac{\sin(\pi n/L)}{\pi n/L}$$

Can see that:

•
$$x_i[n] = \sum_{k=-\infty}^{\infty} x[k] \frac{\sin(\pi(n-kL)/L)}{\pi(n-kL)/L}$$

- Ideal case $h_i[0]=1$ and $h_i[n]=0$ $n=\pm L,\pm 2L,...$
- Can show (you should try!) this gives:
 - $x_i[n] = x[n/L] = x_c[nT/L] = x_c(nT_i)$

Practical Interpolation

- We looked at the ideal LPF
- Can only be approximated in reality
- Linear interpolation can be useful
 - Simplicity/accuracy tradeoff

Change sample rate by non-integer factor

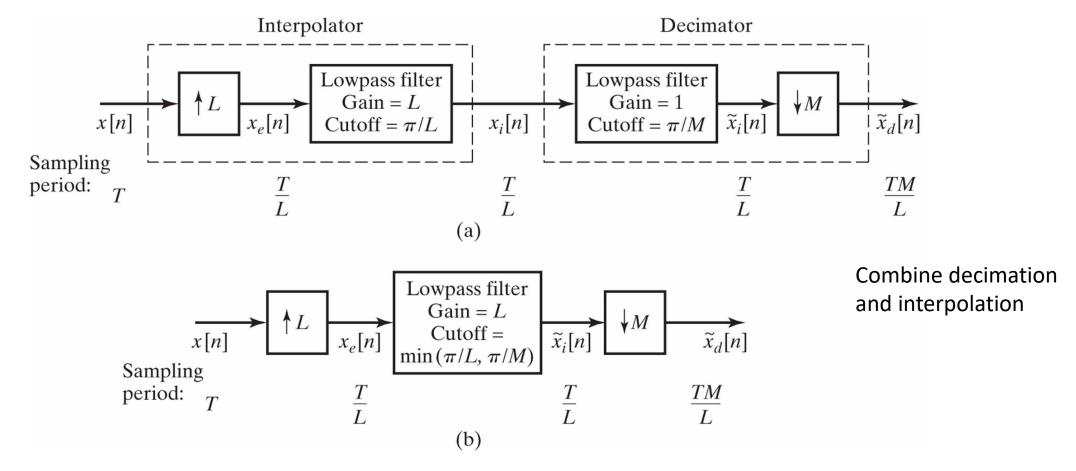


Figure 4.29 (a) System for changing the sampling rate by a noninteger factor. (b) Simplified system in which the decimation and interpolation filters are combined.

Required Reading & other material

• Oppenheim & Schafer, Chapter 4, particularly 4.6