

# Multirate Signal Processing

## Lecture Notes

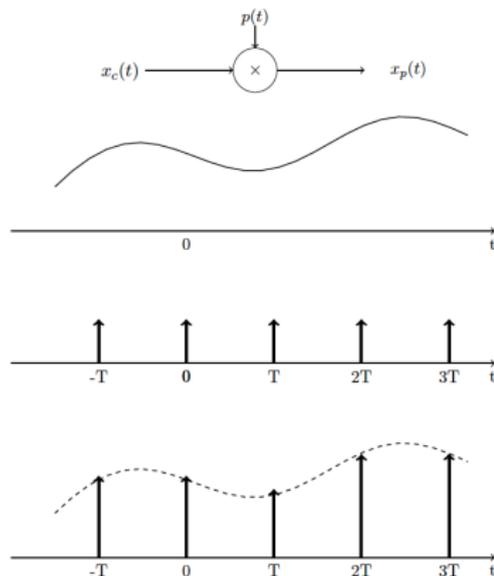
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Some concepts and illustrations in this lecture are adapted from the textbooks,

**Signals and Systems**, 2nd Edition by Alan Oppenheim, Alan Willisky and H. Nawab, *Prentice Hall*.

**Applied Digital Signal Processing**, Dimitris G. Manolakis & K. Ingle, *Cambridge*.

# Nyquist Sampling Theorem



$$x_p(t) = \sum_{n=-\infty}^{\infty} x_c(nT)\delta(t - nT)$$

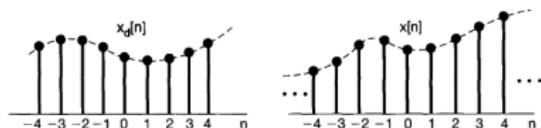
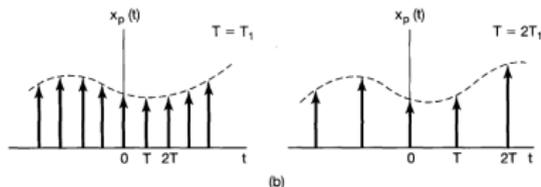
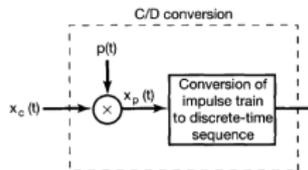
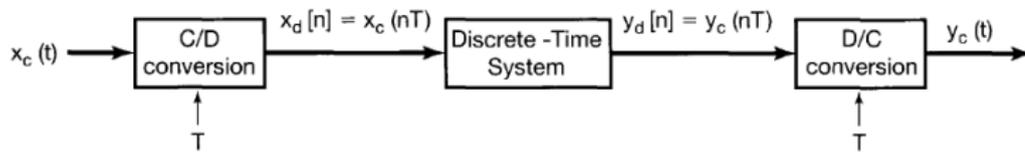
$$x_p(t) = x_c(t)p(t)$$

$$X_p(\Omega) = \frac{1}{2\pi} X_c(\Omega) * P(\Omega)$$

$$P(\Omega) = \frac{2\pi}{T} \sum_{n=-\infty}^{\infty} \delta(\Omega - n\frac{2\pi}{T})$$

$$X_p(\Omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(\Omega - n\frac{2\pi}{T})$$

# Analog to Digital Conversion



$$x_p(t) = \sum_{n=-\infty}^{\infty} x_c(nT)\delta(t-nT) \longleftrightarrow X_p(\Omega) = \sum_{n=-\infty}^{\infty} x_c(nT)e^{-j\Omega nT}$$

Discrete Fourier Transform of  $X_d[n]$  is

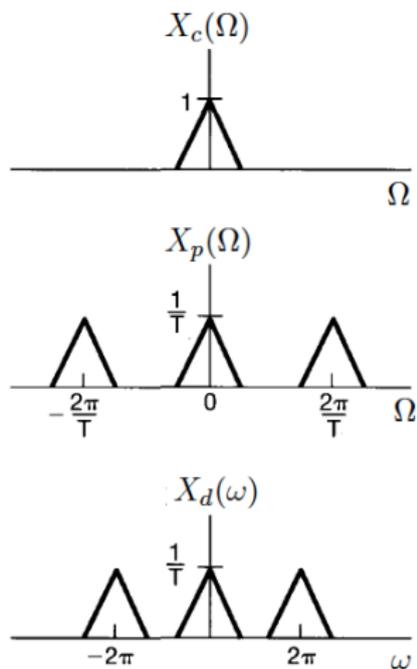
$$X_d(\omega) = \sum_{n=-\infty}^{\infty} x_d[n]e^{-j\omega n} = \sum_{n=-\infty}^{\infty} x_c[nT]e^{-j\omega n}$$

Comparing  $X_p(\Omega)$  and  $X_d(\omega)$ ,

$$X_d(\omega) = X_p(\Omega) \Big|_{\Omega = \frac{\omega}{T}}$$

$$X_p(\Omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(\Omega - n\frac{2\pi}{T}) \text{ and } X_d(\omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(\omega - 2\pi n)$$





When  $x_c(t)$  is converted into  $x_d[n]$ , the time axis is scaled by  $1/T$  which leads to a scaling of frequency axis by  $T$ .

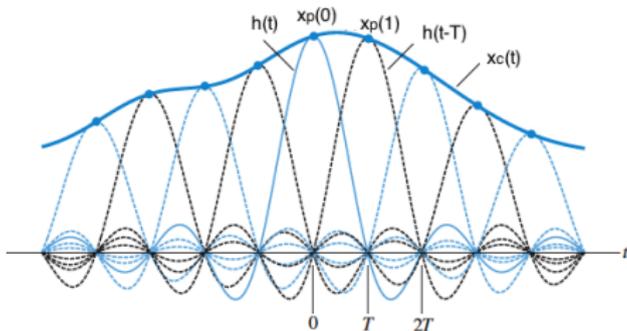
# Reconstruction of $x_c(t)$ from $x_p(t)$

$x_c(t)$  can be recovered from  $x_p(t)$  by a LP filter  $H(\Omega)$  with cut-off frequency  $\Omega_s = \frac{2\pi}{T}$  as  $H(\Omega) = \begin{cases} T & -\Omega_s/2 \leq \Omega \leq \Omega_s/2 \\ 0 & \text{otherwise} \end{cases}$

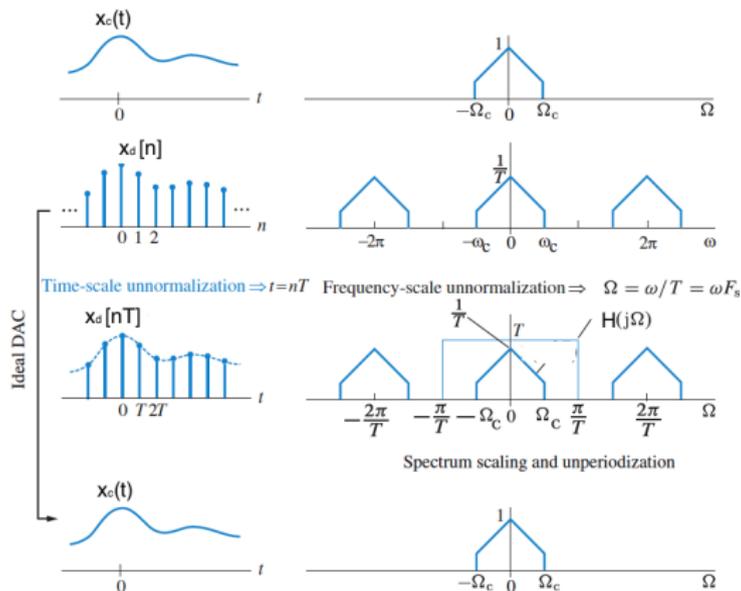
$$h(t) = \frac{\sin(\pi t/T)}{\pi t/T} = \text{sinc}\left(\frac{t}{T}\right)$$

$$x_c(t) = x_p(t) * h(t)$$

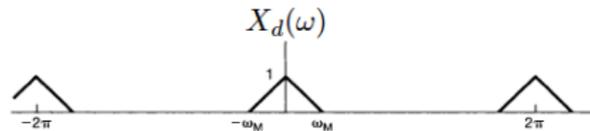
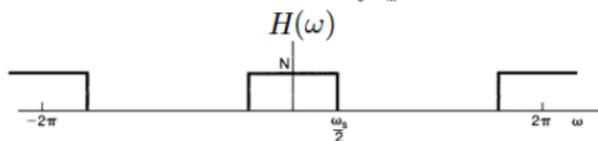
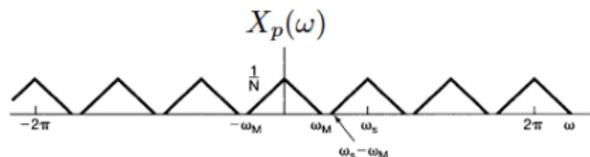
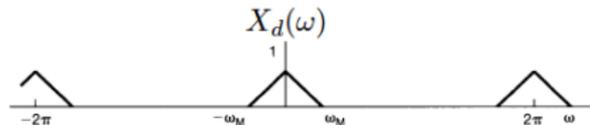
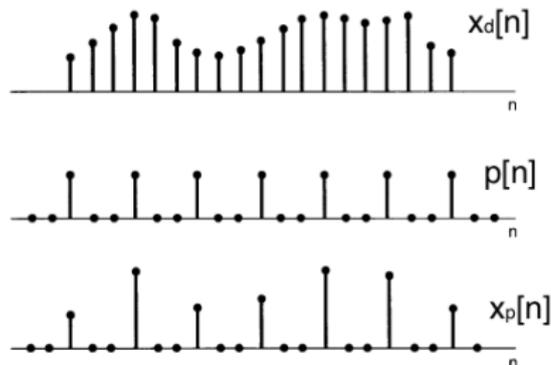
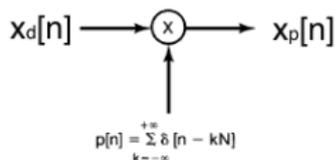
$$x_c(t) = \sum_{n=-\infty}^{\infty} x_p(nT)\delta(t - nT) * \text{sinc}\left(\frac{t}{T}\right) = \sum_{n=-\infty}^{\infty} x_p(nT)\text{sinc}\left(\frac{t-nT}{T}\right)$$



# Reconstruction of $x_c(t)$ from $x_d[n]$



# Sampling and Reconstruction of Discrete Signals

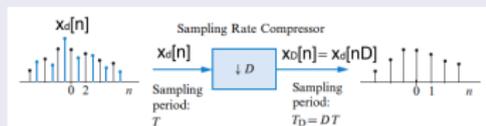


$$X_p(\omega) = \frac{1}{2\pi} \left[ X_d(\omega) * P(\omega) \right] = \frac{1}{2\pi} \left[ X_d(\omega) * \sum_{k=-\infty}^{\infty} \frac{2\pi}{N} \delta(\omega - \omega_s k) \right]$$

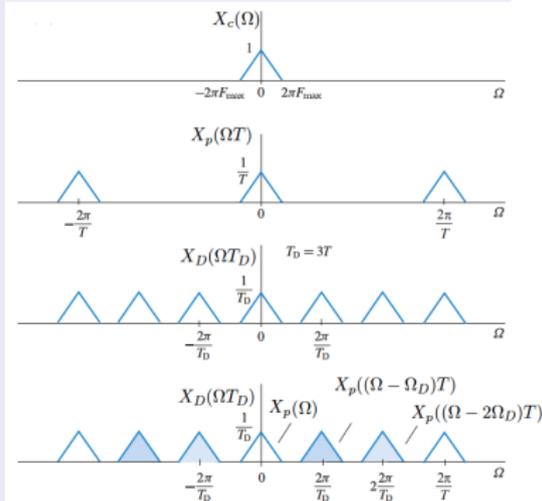
$$= \frac{1}{N} \sum_{k=-\infty}^{\infty} X_d(\omega - \omega_s k) \text{ where } \omega_s = \frac{2\pi}{N}$$

$$X_d(\omega) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(\omega - 2\pi n) \text{ or } X_d(\Omega T) = \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(\Omega - \frac{2\pi}{T} n)$$

## Relation between the spectra of $x_c(t)$ and $x_p[n]$



$$\begin{aligned} X_D(\Omega T_D) &= \frac{1}{T_D} \sum_{n=-\infty}^{\infty} X_c(\Omega - n\Omega_D) \\ &= \frac{1}{DT} \sum_{n=-\infty}^{\infty} X_c(\Omega - n\frac{2\pi}{DT}) \\ T_D &= DT \text{ and } \Omega_D = \frac{2\pi}{T} \end{aligned}$$



$$X_D(\Omega T_D) = \frac{1}{D} \sum_{m=0}^{D-1} \left[ \frac{1}{T} \sum_{n=-\infty}^{\infty} X_c(\Omega - n\frac{2\pi}{T} - m\frac{2\pi}{DT}) \right] = \frac{1}{D} \sum_{m=0}^{D-1} X_p((\Omega - m\Omega_D) T)$$

## Downsampling by a factor of 2

The Fourier transform of a discrete sequence  $x_d[n]$  is

$$X_d(\omega) = \sum_{n=-\infty}^{\infty} x_d[n] e^{-j\omega n}$$

Downsampling  $x_d[n]$  yields a sequence  $x_D[n] = \{\dots, x_d[-2], x_d[0], x_d[2], \dots\}$

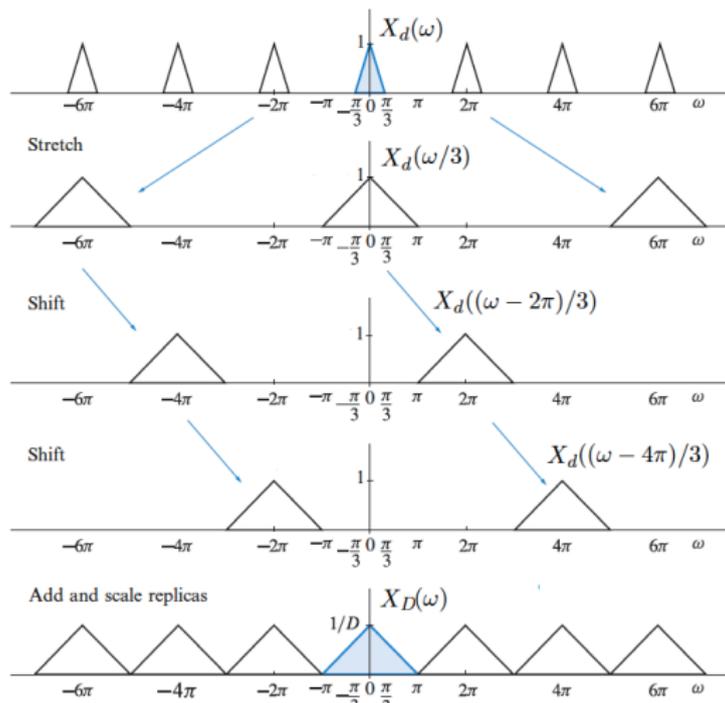
$$X_D(\omega) = \dots + x_d[-2]e^{j\omega} + x_d[0] + x_d[2]e^{-j\omega} + \dots = \sum_{k=-\infty}^{\infty} x_d[2k] e^{-j\omega k}$$

If we define  $x_u[n] = \frac{1}{2}x_d[k] + (-1)^k x_d[k]$

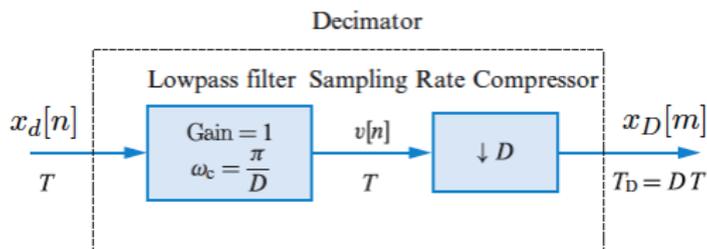
$$X_D(\omega) = \sum_{k=-\infty}^{\infty} x_u[2k] e^{-j\omega k} = X_u(\omega/2)$$

$$X_u(\omega) = \frac{1}{2} \sum_{k=-\infty}^{\infty} [x_d[k] + (-1)^k x_d[k]] e^{-j\omega k} = \frac{1}{2} [X_d(\omega) + X_d(\omega - \pi)]$$

$$X_D(\omega) = X_u(\omega/2) = \frac{1}{2} [X_d(\omega/2) + X_d(\omega/2 - \pi)]$$



$$X_D(\omega) = \frac{1}{3} [X_p(\omega/3) + X_p(\omega/3 - 2\pi/3) + X_p(\omega/3 - 4\pi/3)]$$



$$x_D[m] = v[mD] = \sum_{k=0}^{M-1} h[k]x_d[mD - k]$$

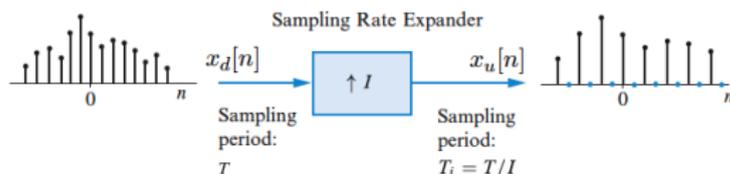
The continuous-time signal  $x_c(t)$  can be perfectly reconstructed from the samples  $x_D[m]$  using the bandlimited interpolation formula

$$x_c(t) = x_D[m] * \text{sinc}(t/TD) = \sum_{m=-\infty}^{\infty} x_D[m] \frac{\sin(\pi(t/TD - m))}{\pi(t/TD - m)}$$

Since  $x_p[n] = x_c(t)|_{t=nT}$ , perfect reconstruction of  $x_p[n]$  is possible

$$x_d[n] = \sum_{m=-\infty}^{\infty} x_D[m] \frac{\sin(\pi(n/D - m))}{\pi(n/D - m)}$$

# Upsampling



The upsampled signal  $x_I[n] \triangleq x_c(nT_i) = x_c(nT/I)$

We can obtain  $x_I[n] = x_d[m] * \frac{\sin(\pi(n/I))}{\pi(n/I)} = \sum_{m=-\infty}^{\infty} x_d[m] \frac{\sin(\pi(n-mI)/I)}{\pi(n-mI)/I}$

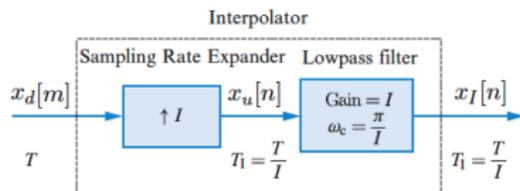
The interpolator function has the general property

$$x_I[n] = \sum_{m=-\infty}^{\infty} x_d[m] h[n - mI], \quad h[n] = \begin{cases} 1 & n = 0 \\ 0 & n = \pm I, \pm 2I, \dots \end{cases}$$

$$X_I(\omega) = \sum_{m=-\infty}^{\infty} x_d[m] H(\omega) e^{-jI\omega m} = H(\omega) \sum_{m=-\infty}^{\infty} x_d[m] e^{-jI\omega m} = H(\omega) X(I\omega)$$

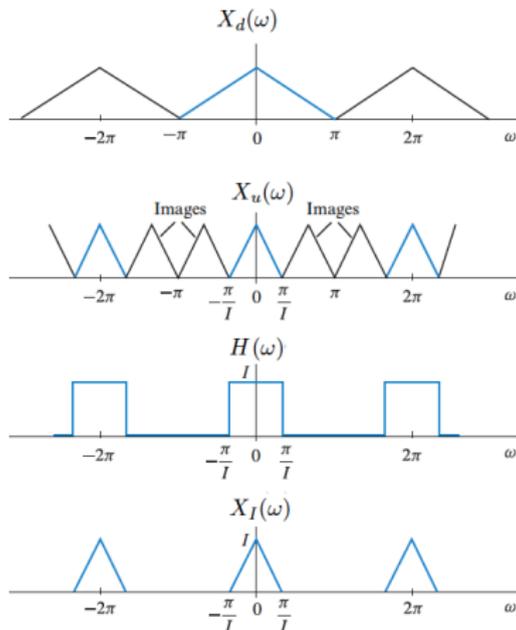
We define  $x_u[n]$  by inserting  $I - 1$  zeros between samples of  $x_d[n]$

$$x_u[n] \triangleq \begin{cases} x_d[n/I] & n : \text{multiple of } I \\ 0 & \text{otherwise} \end{cases} \longleftrightarrow X_u(\omega) = X(I\omega)$$

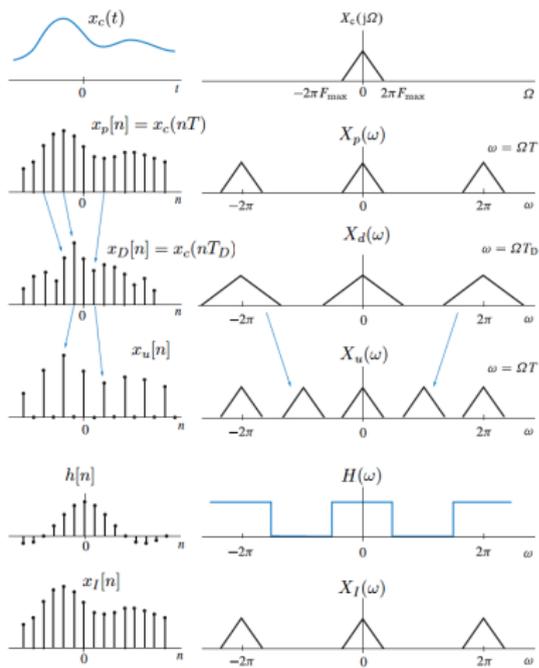
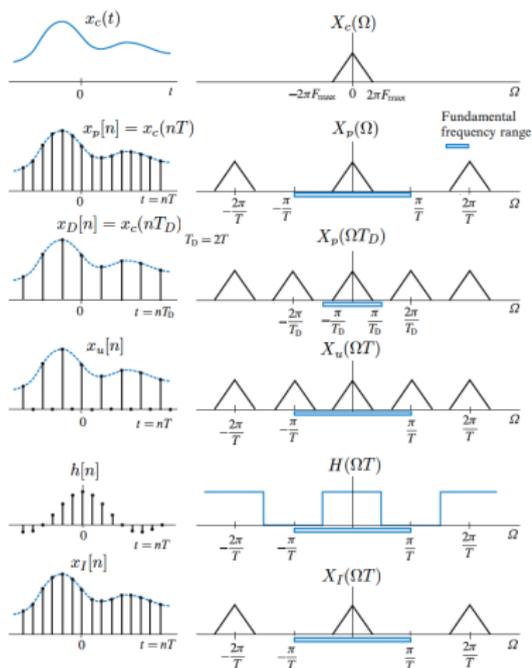


$$X_I(\omega) = H(\omega)X(I\omega)$$

$$x_I[n] = x_u[n] * h[n]$$



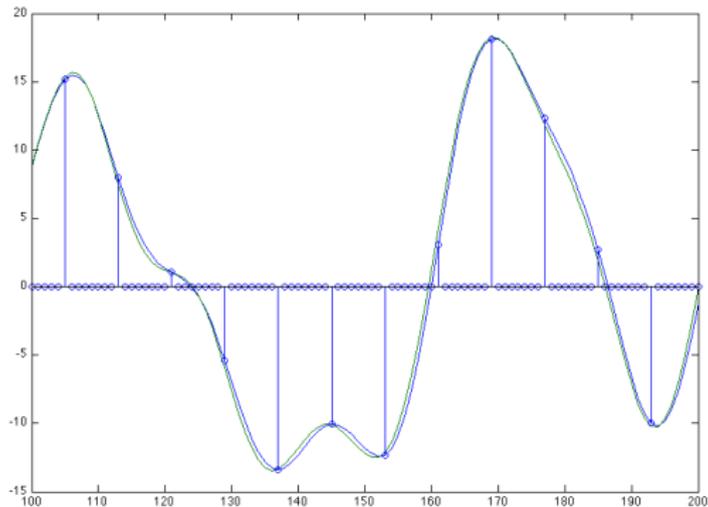
# Decimation and interpolation operations in time and frequency domains



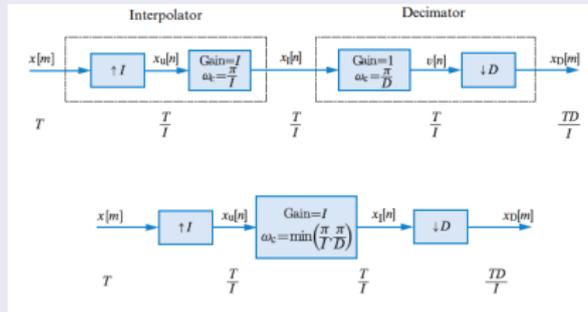
## Tutorial for Interpolation of a downsampled signal using sinc function

```
% Generate a signal as a sum of sinusoids
Nf=5; % number of sinusoids
Fs = 400; % Original sampling rate
Td=1; % signal duration
Af = [10 6 5 4 3]' ; % amplitudes of the sinusoids
f = [ 5 8 10 12 20]; % frequencies of individual components in the signal
T =1/Fs; % sampling interval of the original signal
t = linspace(0,Td,Td*Fs)'; % time axis of the resampled signal
x = sin(t*2*pi*f)*Af; % original signal sampled as a sum of the sinusoids
D = 8 ; % downsample by a factor of D
td=1:D:length(x); % downsampled time index
L = length(td) ; % number of samples in the downsampled signal
M = length(t) ; % number of samples in the original signal
xd=x(td); % downsampled version of the original signal by a factor of D
% Convolution matrix of sinc interpolation function sampled at every T secs
S = sinc( (t*ones(1,L) - ones(M,1)*(td-1)*T)/(D*T) );
x_int = S*xd; % interpolated version of the downsampled signal
x_ups = zeros(size(x)); x_ups(1:D:end) = xd; % upsampled version of the downsampled signal
t1=100:200; % extracting a portion of the signal for plotting
plot(t1,x(t1),t1,x_int(t1)); hold ; stem(t1,x_ups(t1))
```

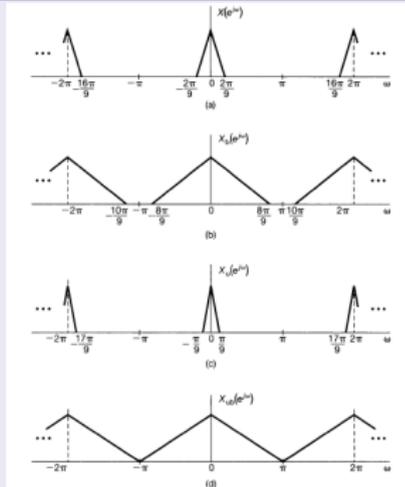
## Fractional delay of a sample



# Sampling rate change by a fractional factor



$$H(\omega) = \begin{cases} 1 & 0 \leq |\omega| \min\left(\frac{\pi}{T}, \frac{\pi}{D}\right) \\ 0 & \min\left(\frac{\pi}{T}, \frac{\pi}{D}\right) \leq |\omega| \leq \pi \end{cases}$$



Original spectrum

Spectrum after downsampling by 4

Spectrum after upsampling  $x[n]$  by 2

Spectrum after upsampling  $x[n]$  by 2

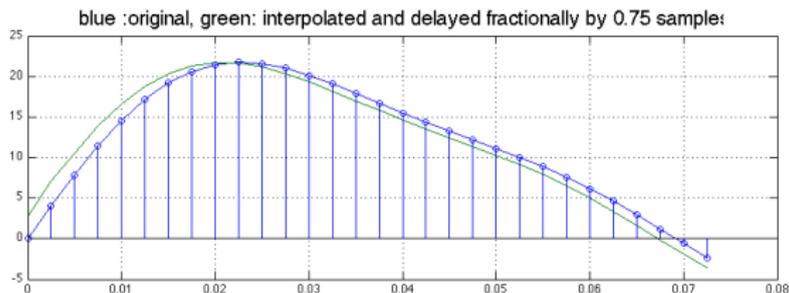
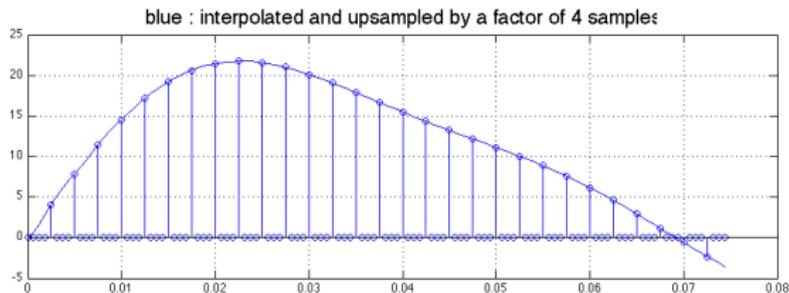
then downsampling by 9.

## Fractional delay of a sample

```
NO = 0.75 % amount of fractional shift of a sampling interval T
D =round(1/min([1-NO NO])) ;
Fs1 = Fs*D; % Sampling rate after upsampling
T =1/Fs1; % sampling interval of the original signal
Td=1; % signal duration
n = [1:length(x)] ; % sampled original signal time index
t = linspace(0,Td,Td*Fs1)'; % time axis of the resampled signal
t_ups=1:D*length(x); % upsampled time axis
x_ups = zeros(size(t_ups)); x_ups(1:D:end) = x; % upsampled version of the original sampled signal
M = length(t_ups) ; % number of samples in the upsampled signal
L = length(x) ; % number of samples in the original signal
% interpolation functions sampled at every T/D secs
S = sinc( ( t*ones(1,L) - ones(M,1)*(n-1)*T*D ) / (D*T) ); % sinc((t-nT)/T)
x_int = S*x; % interpolated version of the resampled signal
x_d = x_int(1+(round(D*NO)):D:end); % downsampled signal with a fractional delay

t1=1:30; % extracting a time interval for the original signal
% extracting the same time interval for the upsampled and interpolated signal
t2=[(D*(min(t1)-1)+1 ):D*max(t1)];
t0=t1-1;
subplot(2,1,2);plot(t0/Fs,x(t1),t0/Fs,x_d(t1));hold ;stem(t0/Fs,x(t1)); grid ;hold
title([ 'blue :original, green: interpolated and delayed fractionally by ' num2str(NO) ' samples' ] );
t0=t2-1;
subplot(2,1,1);plot(t0/Fs1,x_int(t2)); hold ;stem(t0/Fs1,x_ups(t2));hold;grid ;
title([ 'blue : interpolated and upsampled by a factor of ' num2str(D) ' samples' ] )
```

## Fractional delay of a sample



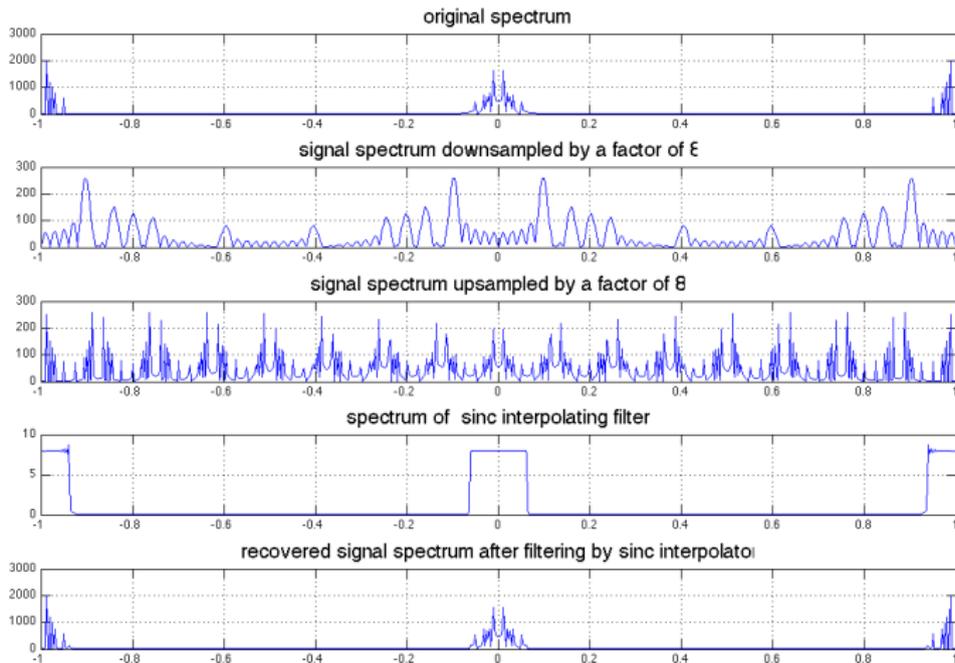
## Frequency spectra of an original, downsampled and interpolated signals

```
Nf=5; % number of sinusoids
Fs = 400; % Original sampling rate
Td=1; % signal duration
Af = [10 6 5 4 3]' ; % amplitudes of the sinusoids
f = [ 5 8 10 12 20]; % frequencies of individual components in the signal
T =1/Fs; % sampling interval of the original signal
t = linspace(0,Td,Td*Fs)'; % time axis of the resampled signal
x = sin(t*2*pi*f)*Af; % original signal sampled as a sum of the sinusoids
D = 8 ; % downsample by a factor of D
td=1:D:length(x); % downsampled time index
L = length(td) ; % number of samples in the downsampled signal
M = length(t) ; % number of samples in the original signal
xd=x(td); % downsampled version of the original signal by a factor of D
% Convolution matrix of sinc interpolation function sampled at every T secs
S = sinc( (t*ones(1,L) - ones(M,1)*(td-1)*T)/(D*T) );
x_int = S*xd; % interpolated version of the downsampled signal
x_ups = zeros(size(x)); x_ups(1:D:end) = xd;
```

## Frequency spectra of an original, downsampled and interpolated signals

```
Df = 800; % sampling rate of frequency spectrum
w = linspace(-1,1,Df);
% Estimating frequency responses
H_x = freqz(x,1,w*2*pi);
H_xups = freqz(x_ups,1,w*2*pi);
H_xd = freqz(xd,1,w*2*pi);
H_I = freqz(S(:,end/2),1,w*2*pi);
H_xint = freqz(x_int ,1,w*2*pi);
subplot(5,1,1); plot(w,abs(H_x));grid; title('original spectrum','fontsize',16)
subplot(5,1,2); plot(w,abs(H_xd));
grid;title('signal spectrum downsampled by a factor of 8','fontsize',16);
subplot(5,1,3); plot(w,abs(H_xups));
grid;title('signal spectrum upsampled by a factor of 8','fontsize',16);
subplot(5,1,4); plot(w,abs(H_I));
grid;title('spectrum of sinc interpolating filter','fontsize',16);
subplot(5,1,5); plot(w,abs(H_xint));
grid;title('recovered signal spectrum after filtering by sinc interpolator','fontsize',16);
```

## Frequency spectra of an original, downsampled and interpolated signals



## Sampling rate compressor

$$\text{Sampling train : } \delta_D[n] = \sum_{k=-\infty}^{\infty} \delta[n - kD] = \frac{1}{D} \sum_{k=0}^{D-1} W_D^{-kn}, \quad W_D = e^{-\frac{j2\pi}{D}}$$

$z$ -transform of sampled sequence  $v[n] = x[n]\delta_D[n]$ :

$$\begin{aligned} V(z) &= \sum_{n=-\infty}^{\infty} x[n]\delta_D[n]z^{-n} = \sum_{n=-\infty}^{\infty} x[n] \left[ \frac{1}{D} \sum_{k=0}^{D-1} W_D^{-kn} \right] z^{-n} \\ &= \frac{1}{D} \sum_{n=-\infty}^{\infty} \sum_{k=0}^{D-1} x[n] (W_D^k z)^{-n} = \frac{1}{D} \sum_{k=0}^{D-1} X(W_D^k z) \end{aligned}$$

Upsampler output  $y[n] = v[nD] = x[nD]$  and its  $z$ -transform

$$Y(z) = \sum_{n=-\infty}^{\infty} v[nD]z^{-n} = \sum_{m=-\infty}^{\infty} v[m]z^{-m/D} = V(z^{1/D}) = \frac{1}{D} \sum_{k=0}^{D-1} X(W_D^k z^{1/D})$$

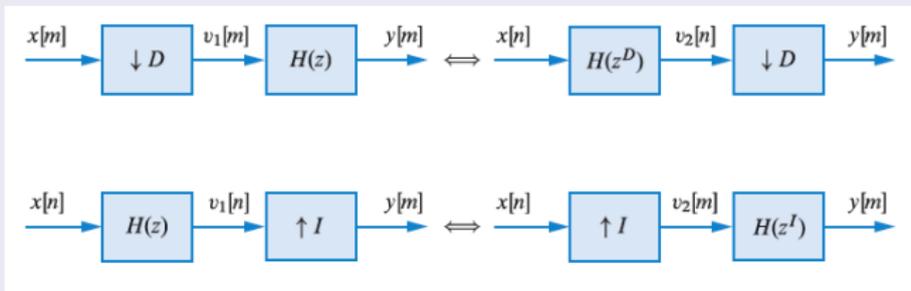
## Sampling rate expander

$$y[n] = \begin{cases} x[n/I] & n = 0, \pm I, \pm 2I, \dots \\ 0 & \text{otherwise} \end{cases}$$

The  $z$ -transform of the output

$$Y(z) = \sum_{n=-\infty}^{\infty} x[n/I]z^{-n} = \sum_{m=-\infty}^{\infty} x[m]z^{-mI} = X(z^I)$$

## Multirate identities: Interchangibility of up or down sampling with filtering



$$Y(z) = H(z)V_1(z) = H(z)\frac{1}{D} \sum_{k=0}^{D-1} X(W_D^k z^{1/D}) \Leftrightarrow$$

$$V_2(z) = H(z^D)X(z) \text{ and } Y(z) = \frac{1}{D} \sum_{k=0}^{D-1} V_2(W_D^k z^{1/D})$$

$$Y(z) = H(z)\frac{1}{D} \sum_{k=0}^{D-1} X(W_D^k z^{1/D}) \text{ since } H(W_D^k z^{D/D}) = H(z)$$

$$V_1(z) = H(z)X(z) \text{ and } Y(z) = V_1(z^I) \text{ and } Y(z) = H(z^I)X(z^I)$$

$$V_2(z) = X(z^I) \text{ and } Y(z) = H(z^I)V_2(z) \text{ and } Y(z) = H(z^I)X(z^I)$$

## Polyphase Filters

An FIR filter with  $N = ML$  is realized with  $M$  filters each having length  $L$  as for instance

$N = 6$  and  $M = 2$

$$H(z) = h[0] + h[1]z^{-1} + h[2]z^{-2} + h[3]z^{-3} + h[4]z^{-4} + h[5]z^{-5}$$

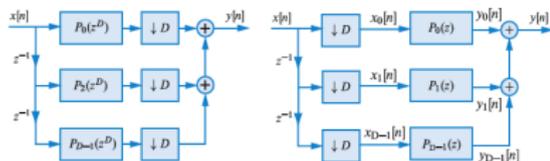
$$= (h[0] + h[2]z^{-2} + h[4]z^{-4}) + z^{-1}(h[1] + h[3]z^{-2} + h[5]z^{-4})$$

$$H(z) = P_0(z^2) + z^{-1}P_1(z^2),$$

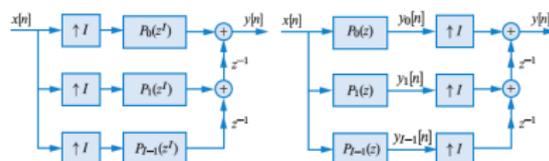
In general,

$$p_k[n] \triangleq h[nM + k], \quad k = 0, 1, 2, \dots, M - 1$$

$$H(z) = \sum_{k=0}^{M-1} z^{-k} P_k[z^M], \quad P_k(z) = \sum_{n=0}^{L-1} P_k[n]z^{-n}$$



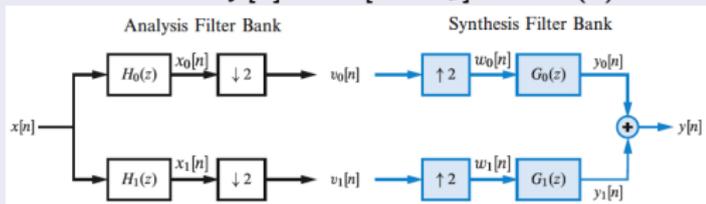
Polyphase Decimation



Polyphase Upsampling

## Perfect reconstruction filters

Condition for perfect reconstruction:  $y[n] = Gx[n - n_d] \xleftrightarrow{Z} Y(z) = Gz^{-n_d} X(z)$



Input and output of upper channel using  $Y(z) = \frac{1}{D} \sum_{k=0}^{D-1} X(W_D^k z^{1/D})$  and  $Y(z) = X(z^I)$

$$V_0(z) = \frac{1}{2} H_0(z^{1/2}) X(z^{1/2}) + \frac{1}{2} H_0(-z^{1/2}) X(-z^{1/2})$$

$$Y_0(z) = V_0(z^2) G_0(z)$$

$$Y_0(z) = \frac{1}{2} [H_0(z) X(z) + H_0(-z) X(-z)] G_0(z)$$

Input and output of lower channel

$$V_1(z) = \frac{1}{2} [H_1(z^{1/2}) X(z^{1/2}) + H_1(-z^{1/2}) X(-z^{1/2})]$$

$$Y_1(z) = V_1(z^2) G_1(z)$$

$$Y_1(z) = \frac{1}{2} [H_1(z) X(z) + H_1(-z) X(-z)] G_1(z)$$

$$Y(z) = \frac{1}{2} [T(z) X(z) + A(z) X(-z)]$$

$$T(z) = H_0(z) G_0(z) + H_1(z) G_1(z) \quad A(z) = H_0(-z) G_0(z) + H_1(-z) G_1(z)$$

*Aliasing*

*phase / magnitude distortion*

$$T(z) = H_0(z) G_0(z) + H_1(z) G_1(z) = Gz^{-n_d} \quad A(z) = H_0(-z) G_0(z) + H_1(-z) G_1(z) = 0$$

Perfect reconstruction condition:

## Perfect reconstruction condition

$$\begin{bmatrix} H_0(z) & H_1(z) \\ H_0(-z) & H_1(-z) \end{bmatrix} \begin{bmatrix} G_0(z) \\ G_1(z) \end{bmatrix} = \begin{bmatrix} Gz^{-n_d} \\ 0 \end{bmatrix}$$
 Setting  $G = 2$  preserves the amplitude of input signal.

## Synthesis filters

$$G_0(z) = \frac{2z^{-n_d}}{\Delta_m(z)} H_1(-z), \quad G_1(z) = -\frac{2z^{-n_d}}{\Delta_m(z)} H_0(-z),$$
$$\Delta_m(z) = H_0(z)H_1(-z) - H_0(-z)H_1(z) = -\Delta_m(-z)$$

Perfect reconstruction conditions:

$$\underbrace{z^{n_d} H_0(z) G_0(z)}_{R(z)} + \underbrace{z^{n_d} H_1(z) G_1(z)}_{R(-z)} = 2 \quad \text{Since } H_1(z) = G_0(-z) \Delta_m(-z) \frac{1}{2} (-z)^{n_d}$$

Polyphase decomposition of  $R(z) = R_0(z^2) + z^{-1}R_1(z^2)$  with PR yields  $R_0(z^2) + z^{-1}R_1(z^2) + R_0(z^2) - z^{-1}R_1(z^2) = 2$  or  $R_0(z^2) = 1$  which makes  $R(z)$  a half-band filter *i.e.*  $\mathcal{F}^{-1}$ [rectangular filter with cutoff  $\pi/2$ ]

## Design of a two-channel PR filter bank

- Obtain an  $R(z)$  satisfying  $R(z) = 1 + z^{-1}R_1(z^2)$
- perform the factorization  $R(z) = z^{n_d} H_0(z) G_0(z)$
- assign the remaining filters from  $G_0(z) = \frac{2z^{-n_d}}{\Delta_m(z)} H_1(-z), G_1(z) = -\frac{2z^{-n_d}}{\Delta_m(z)} H_0(-z)$



## Quadrature mirror filter (QMF) banks

Design a low-pass filter  $H(z)$  and then determine the filters in the bank as;

$$H_0(z) = H(z), H_1(z) = H(-z)$$

$$G_0(z) = H(z), G_1(z) = -H(-z)$$

Time and frequency domain characteristics :

$$h_0[n] = h[n] \quad \xleftrightarrow{DTFT} \quad H_0(\omega) = H(\omega)$$

$$h_1[n] = (-1)^n h[n] \quad \xleftrightarrow{DTFT} \quad H_1(\omega) = H(\omega - \pi)$$

$$g_0[n] = h[n] \quad \xleftrightarrow{DTFT} \quad G_0(\omega) = H(\omega)$$

$$g_1[n] = (-1)^{n+1} h[n] \quad \xleftrightarrow{DTFT} \quad H_1(\omega) = -H(\omega - \pi)$$

Since  $H_0(\omega) = H_1(\omega - \pi)$ ,  $\omega \rightarrow \omega + \pi/2$ , leads to  $|H_0(\omega + \pi/2)| = |H_1(\pi/2 - \omega)|$   
Symmetric about quadrature frequency  $2\pi/4$

## Perfect reconstruction property

$$A(z) = H_0(-z)G_0(z) + H_1(-z)G_1(z) = H(-z)H(z) - H(z)H(-z) = 0$$

$$T(z) = H_0(z)G_0(z) + H_1(z)G_1(z) = H^2(z) - H^2(-z)$$

$$\Delta_m(z) = H_0(z)H_1(-z) - H_0(-z)H_1(z) = H^2(z) - H^2(-z)$$

## Polyphase decomposition

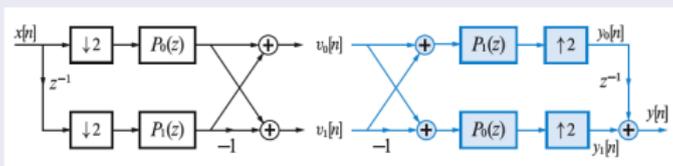
$$H(z) = P_0(z^2) + z^{-1}P_1(z^2)$$

$$T(z) = \Delta_m(z) = 4z^{-1}P_0(z^2)P_1(z^2)$$

If  $P_0(z^2) = b_0z^{-n_0}$  and  $P_1(z^2) = b_1z^{-n_1}$

then FIR PR filter is a pure delay

$H(z) = b_0z^{2n_0} + b_1z^{-(2n_1+1)}$  but it has not much of a practical value.



## Power complementary filters

A low pass filter with linear phase (Type I or II)  $H(\omega) = A(\omega)e^{-j\omega M/2}$  has even symmetry

$$T(\omega) = H^2(\omega) - H^2(-\omega) = e^{-j\omega M} \left[ |H(\omega)|^2 - (-1)^M |H(\omega - \pi)|^2 \right]$$

$M = 0$  causes  $T(\omega) = 0$  leading to severe amplitude distortion.

Choose  $M$  odd (Type II FIR)

$$T(\omega) = e^{-j\omega M} \left[ |H(\omega)|^2 + |H(\omega - \pi)|^2 \right] = e^{-j\omega M} \underbrace{\left[ |H_0(\omega)|^2 + |H_1(\omega)|^2 \right]}$$

$$|H_0(\omega)|^2 + |H_1(\omega)|^2 = 1$$

No exact solution other than  $H(z) = b_0 z^{2n_0} + b_1 z^{-(2n_1+1)}$

An approximate solution is obtained by

$$J = \underbrace{\alpha \int_{\omega_s}^{\pi} |H(\omega)|^2 d\omega}_{\text{aliasing}} + \underbrace{(1 - \alpha) \int_0^{\pi} (1 - |T(\omega)|^2) d\omega}_{\text{perfect reconstruction}}, \quad 0 < \alpha < 1$$

