

**Principles of Communication- Part I**  
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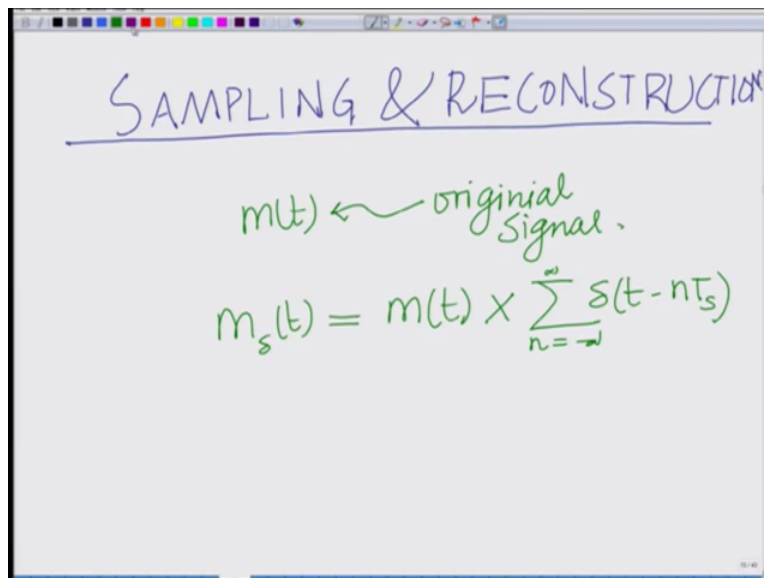
**Module No 6**

**Lecture 37**

**Ideal Impulse Train Sampling, Reconstruction of Original Signal from Samples, Sinc Interpolation**

Hello welcome to another module in this massive open online course, so we are looking at the sample sampling of analog signals. In this module let us start looking at also the reconstruction aspect that is from the (recons) that is from the samples of the sampled signal how can we reconstruct the original signal?

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Alright, so let us look at so we are looking at sampling let us also start looking at reconstruction that is how do you reconstruct? That is how do you reconstruct? That is if you want to reconstruct the original samples how do you reconstruct the original? How do you reconstruct the original signal from the sample? So we want to look at the reconstruction aspects, so sampling and reconstruction now we have already seen that the sampled signal if  $m(t)$  is the original signal this is the your original signal, now the sampled signal  $m_s(t)$  is given by multiplying this by an impulse train that is  $m(t)$  times summation  $n$  equal to minus infinity to infinity  $\delta(t - nT_s)$  this is your impulse train, correct?

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$$m_s(t) = m(t) \times \sum_{n=-\infty}^{\infty} \delta(t - nT_s)$$

↑  
impulseTrain  
↑  
 $f_s = \frac{1}{T_s}$   
Sampling Frequency.

So this is your impulse train which is a series of impulses at every multiple of the sampling time interval  $T_s$ , okay where a sampling frequency  $F_s$  equals  $1$  over  $T_s$ , okay.  $F_s$  is the sampling frequency this is our sampling frequency and now we are multiplying  $m(t)$  by this (impul) (tra) impulse train that is  $m(t)$  times summation  $n$  equal to minus infinity to infinity delta  $t$  minus  $nT_s$  and that each impulse at  $nT_s$ , right? Picks the value of the original signal  $m$  at  $nT_s$ .

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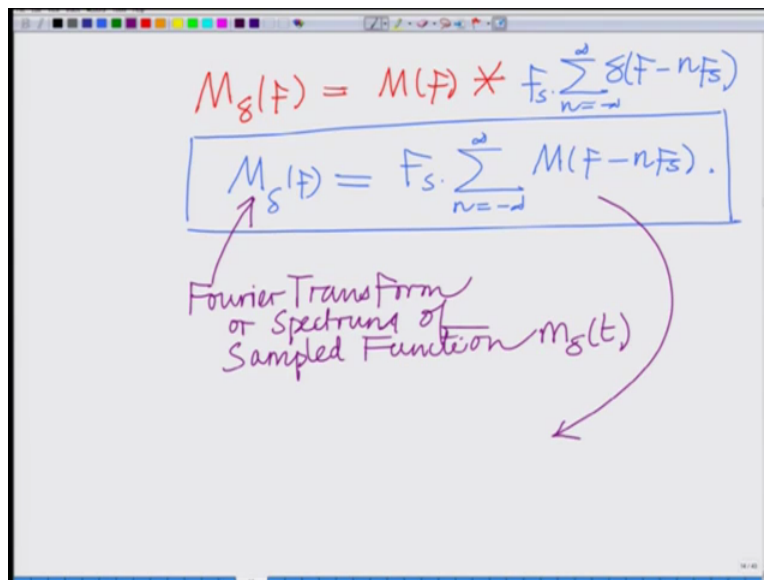
$$m_s(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \delta(t - nT_s)$$

↑  
Sum of impulses  
at  $nT_s$  scaled by  
 $m(nT_s)$ .

$$m_s(t) \leftrightarrow M_s(f)$$

So we have from this from this impulse sampling what we are going to get is your  $m \Delta T$  equals summation  $n$  equal to minus infinity to infinity  $m$  of  $nT_s$  times  $\delta(t - nT_s)$  that is impulse at  $nT_s$  scaled by the function  $nT_s$ , so impulse so what is this? Summation of impulses at  $nT_s$  scaled by impulse at  $nT_s$  scaled by the function that is  $m$  at function at  $nT_s$ . Now also when we looked at the frequency (respo) we looked at the Fourier transform of this that is if  $m \Delta t$  has a Fourier transform has a Fourier transform  $m \Delta F$ .

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The image shows a whiteboard with handwritten mathematical equations. At the top, the equation  $M_s(f) = M(f) * f_s \sum_{n=-\infty}^{\infty} \delta(f - n f_s)$  is written in red. Below it, the equation  $M_s(f) = f_s \sum_{n=-\infty}^{\infty} M(f - n f_s)$  is written in blue and enclosed in a blue rectangular box. A purple arrow points from the text "Fourier Transform or Spectrum of Sampled Function  $m_s(t)$ " to the boxed equation. Another purple arrow points from the boxed equation down and to the right.

We said that  $m \Delta F$  is basically  $M$  of  $F$  convolved with since in the time domain we are multiplying in the frequency domain we have to convolved with the frequency response of the impulse train in the frequency for your transfer of the impulse train we have already seen is summation is  $F_s$  summation  $n$  equal to minus infinity to infinity  $\delta(f - n f_s)$  which is equal to  $F_s$  summation  $n$  equal to minus infinity to infinity  $M$  of  $F$  minus  $F_s$ .

So this is your  $M \Delta F$  this is the Fourier transform of the sample function, okay. This is the Fourier transform of your sampled function or basically the spectrum the spectrum of the sampled function  $m \Delta t$ . And if you write is what you will see? Is that this is  $F_s$  so  $F_s$  is a scaling factor and  $MF$  the original spectrum  $MF$  shifted to each  $n f_s$  that is  $MF$  minus  $n f_s$  shifted every multiple of the sampling frequency and the sum of all these spectral copies.

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Fourier Transform  
or Spectrum of  
Sampled Function  $m_s(t)$

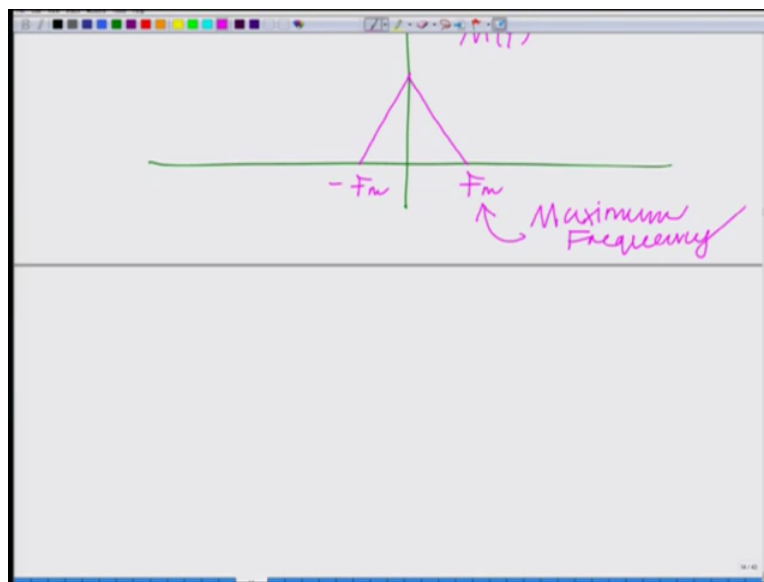
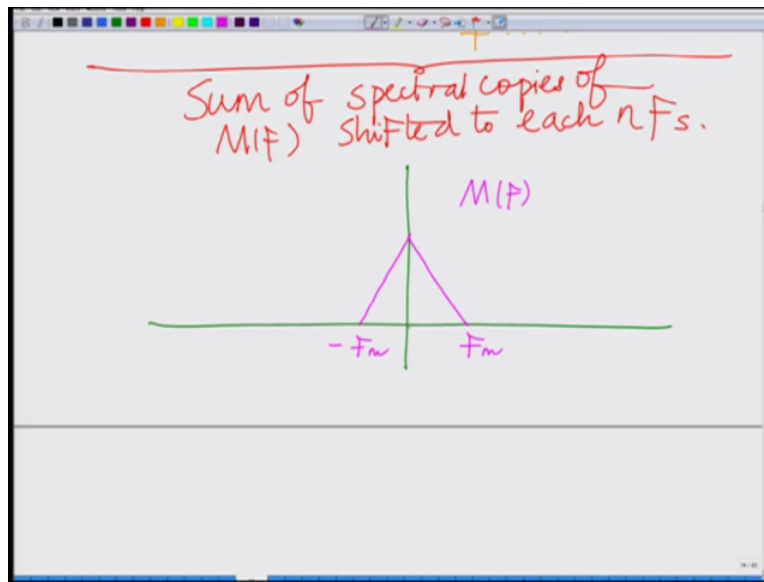
$$\dots + F_s M(f + 2F_s) + F_s M(f + F_s) + F_s M(f) + F_s M(f - F_s) + F_s M(f - 2F_s) + \dots$$


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Sum of spectral copies of  
 $M(f)$  shifted to each  $nF_s$ .

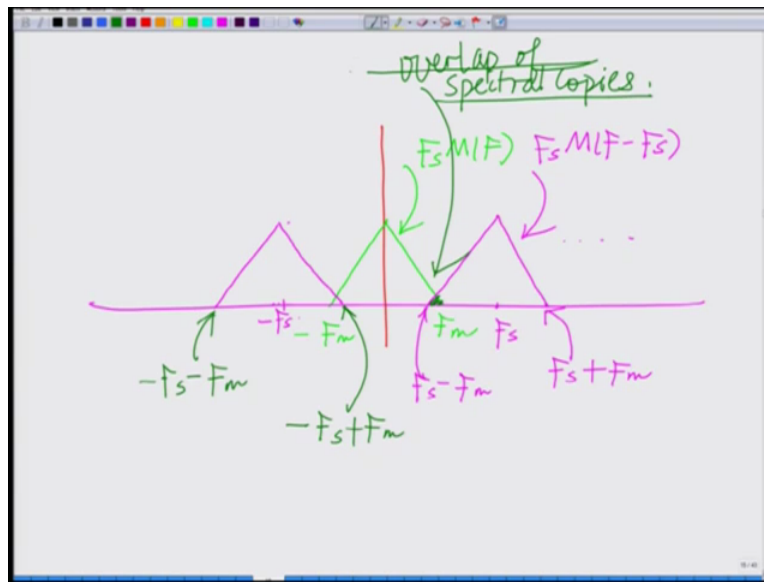
So what you have is very interesting you have  $F_s$  times  $M$  of  $F$   $F_s$  times  $M$  of well,  $F$  minus  $F_s$  plus  $F_s$  times  $M$  of  $F$  minus  $2F_s$  plus so on plus also  $F_s$  that is this is copy shifted to minus  $F_s$ ,  $F_s$   $M$  of  $F$  plus  $F_s$ ,  $F_s$   $M$  of  $F$  plus  $F_s$  plus  $F_s$   $M$  of  $F$  plus  $2F_s$  there is copy shifted to minus 2 and so on, alright. So this is basically the sum of spectral copies sum of spectral copies shifted to of  $MF$  shifted to each  $nF_s$  where  $n$  shifted to each  $nF_s$  where  $n$  is any integer  $F_s$  is the sampling frequency. So therefore this has a very beautiful frequency domain interpretation if you look at it in the frequency domain what we have is you have your original spectrum.

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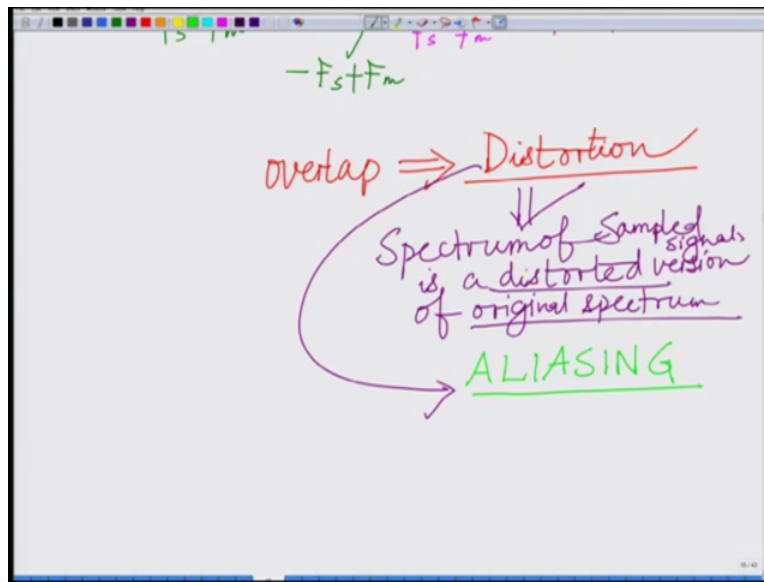
If you have your original spectrum that is MF which is has maximum frequency  $F_m$  let us say minimum frequency that is from minus  $F_m$  to  $F_m$  that is its support or where it is nonzero is from minus  $F_m$  to  $F_m$  the maximum frequency is  $F_m$ , this is the maximum frequency and therefore now what you are doing is basically you have  $F_s$  times MF.

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So the sampled signal  $m \Delta F$  will be well,  $F_s$  times  $M/F$  so this is  $F$  of  $M$   $F_m$  minus  $F_m$ , so this is  $F_s$  times  $M/F$  then you have another copy at  $F_s$  shifted by  $F_s$  which can be this is  $F_s$ , so naturally this point is going to be  $F_s$  minus  $F_m$  this point is going to be  $F_s$  plus  $F_m$  this is your  $F_s$  times  $M$  of  $F$  minus  $F_s$ , so on your going to be have copies  $F_s M$  of  $F$  minus  $2F_s$  copy at  $2F_s$  and also similarly you can have a copy at that is if this is your minus  $F_s$  this is minus  $F_s$ , so naturally this point is going to be minus  $F_s$  plus  $F_m$ , this point is going to be minus  $F_s$  minus  $F_s$  and now you can see there is an overlap here there is a possibility of overlap of spectral copies and this overlap of spectral copies this leads to distortion that is what we have said overlap of spectral copies overlap leads to distortion, okay.

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Distortion means that the sampled spectrum is a distorted version of the original spectrum that is  $m \Delta F$  that is sampled spectrum what does this implies? Sampled spectrum for I should not say sampled spectrum rather I should say spectrum of sampled signal we are not sampling the spectrum but rather the spectrum of sampled signal spectrum of sampled signal is a distorted version of the original spectrum and this distortion is termed as this is very important this distortion is termed as Aliasing.

We have said that this distortion is termed as Aliasing that is these different spectral copies alias, okay. And the condition for aliasing, remember we have also derived the condition for aliasing, aliasing occurs when  $F_s - F_m$  is less than  $4F_m$ .

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Aliasing occurs if

$$f_s - f_m < f_m$$
$$\Rightarrow \boxed{f_s < 2f_m}$$

Therefore, to avoid Aliasing,  
we need,

$$\boxed{f_s \geq 2f_m}$$

So aliasing occurs when occurs if  $f_s$  minus  $f_m$  is less than  $f_m$  that implies  $f_s$  less than  $2f_m$  that is sampling frequency is less than twice the maximum frequency, so this is the condition for aliasing therefore to avoid aliasing we need we need  $f_s$  greater than or equal to  $2f_m$  that is the minimum sampling frequency to avoid aliasing or the distortion caused by aliasing is  $f_s$  greater than equal to, to avoid aliasing we need  $f_s$  greater than equal to  $2f_m$  where  $f_m$  is the maximum frequency.

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Therefore, to avoid Aliasing,  
we need,

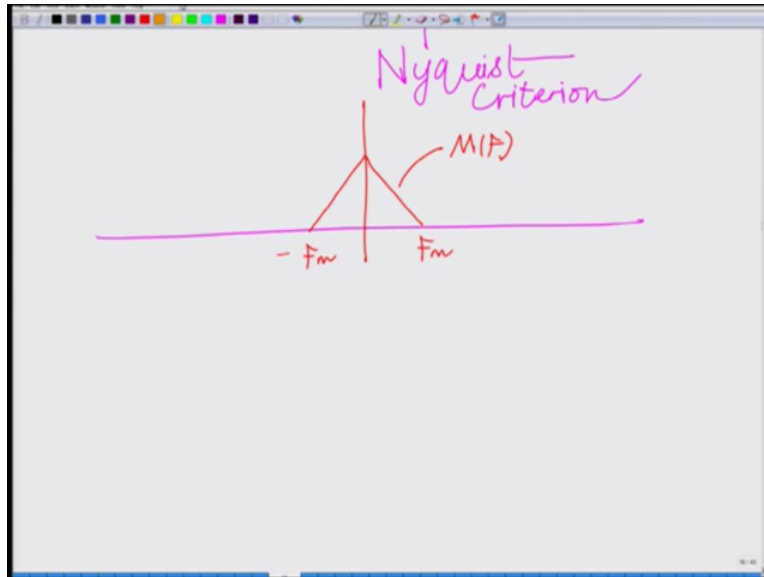
$$\boxed{f_s \geq 2f_m}$$

Nyquist  
Criterion



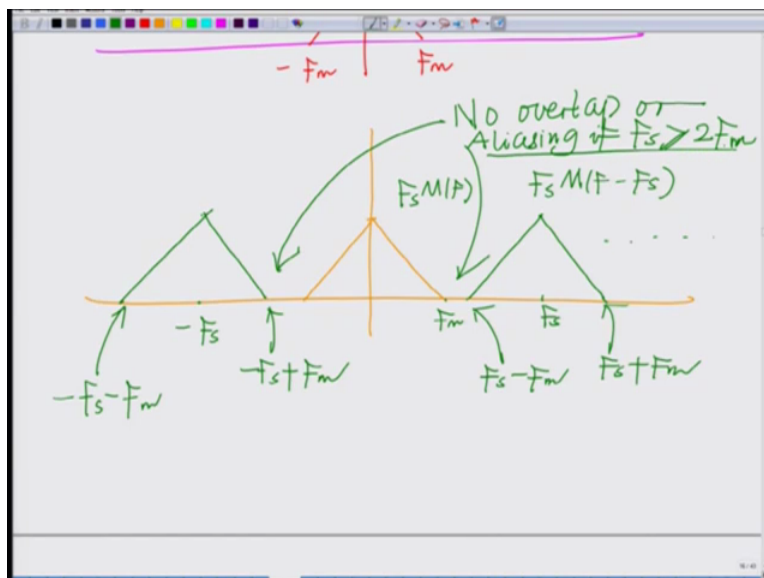
And this is known as the Nyquist criterion this is known as the Nyquist sampling theorem or this is known as the Nyquist, this is known as the Nyquist criterion.

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And remember when  $F_s$  is greater than  $F_m$  what is going to happen? This is your original spectrum let us say this is your original spectrum.

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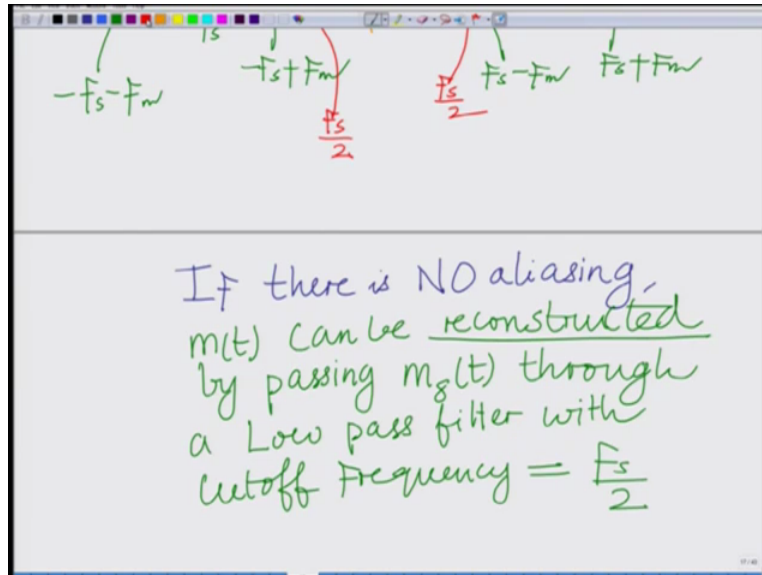
Similarly here you have at minus  $F_s$  this is your minus  $F_s$  plus  $F_m$  this is your minus  $F_s$  minus  $F_m$  and now realize there is no overlap or no aliasing or aliasing if  $F_s$  greater than or equal to  $F_s$  is greater than or equal to  $2F_m$ , so there is no overlap or aliasing if  $F_s$  is greater than or equal to  $2F_m$  and now there is no distortion, alright since the sampling frequency is greater than equal to twice the maximum frequency there is no distortion, now you can see that we can reconstruct the original message signal from the sampled spectrum and the way to do that is by filtering this copy the original the copy of the spectrum which is intact at the (series) at the at the frequency 0 that is we have  $F_s$  times  $MF$  which is intact at in the baseband that is which is intact at which is centered around the 0 frequency I can extract that by low pass filtering this with an ideal low pass filter, okay.

Diagram illustrating the Nyquist sampling theorem. The central orange triangle represents the baseband signal spectrum, centered at 0. The green triangles represent the aliased spectra, centered at  $-F_s$  and  $F_s$ . The baseband triangle has a peak value of  $\frac{1}{2}$  and a base extending from  $-\frac{F_s}{2}$  to  $\frac{F_s}{2}$ . The aliased triangles have bases extending from  $-F_s - \frac{F_s}{2}$  to  $-F_s + \frac{F_s}{2}$  and from  $F_s - \frac{F_s}{2}$  to  $F_s + \frac{F_s}{2}$ . The condition for no overlap (no aliasing) is  $F_s \geq 2F_m$ .

So now I can extract this now this is remember this is  $F_s$  this is 0 so this is your  $F_s$  by 2  $F_s$  over 2 sampling frequency by 2 this is your sampling frequency by 2, now I can extract this by low pass

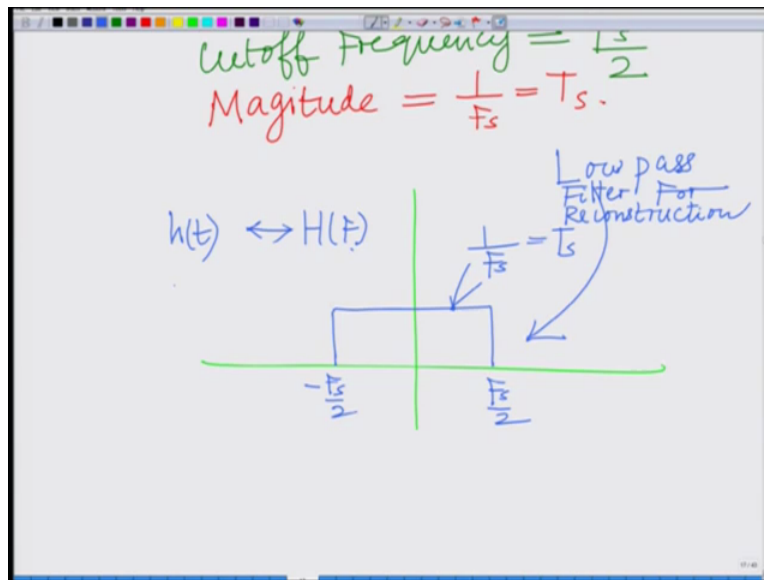
filtering this by low pass filtering this you can clearly see if I have low pass filter with cut-off at  $F_s/2$  and there is a scaling factor of  $F_s$ , so this has to be  $1/F_s$  the height of this has to be  $1/F_s$  equal to  $T_s$ .

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So now I can reconstruct the original so now we realize something interesting I reconstruct if there is no alias remember if there is no aliasing that is an important condition if there is no aliasing reconstruct or original message signal or  $m(t)$  can be reconstructed there is a ready signal can be reconstructed by passing  $m \Delta t$  through a low pass filter with cut-off frequency equals  $F_s/2$  the cut-off frequency of low pass filter is  $F_s/2$  that is it is one if  $F$  magnitude  $F$  is less than  $F_s/2$  and 0 otherwise and also there is a scaling of  $F_s$ . So to invert that you have to multiply by  $1/F_s$  therefore and so cut-off frequency is  $F_s/2$  and the height of the response is  $1/F_s$  that is  $T_s$ .

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And frequency and height and frequency response magnitude equals  $1$  over  $f_s$  equals  $T_s$ , so the filter looks like this basically I have my filter, okay. And now I have a low pass filter with cut-off equals  $f_s$  by  $2$ . So this is  $f_s$  by  $2$  this is minus  $f_s$  by  $2$  the height of this is  $1$  over  $f_s$  equals  $T_s$ , okay. So this is my reconstruction low pass filter low pass filter for this is my low pass filter for reconstruction let us term this as  $h$  of  $t$ , so this is this is  $H$  of  $F$  which is the response of let us say the impulse responses  $h$  of  $t$ . So we have  $H$  of  $F$  the response is basically it is easy to write this down this is  $H$  of  $F$  which is equal to  $1$  or basically  $T_s$  if  $F$  magnitude of  $F$  is less than or equal to  $f_s$  by  $2$  where  $f_s$  is sampling frequency and this is  $0$  and this is  $0$  otherwise, okay. So this is basically the low pass filter that the LPF or the low pass filter that we want to employ for reconstruction.

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$$H(f) = \begin{cases} T_s & \text{if } |f| \leq \frac{F_s}{2} \\ 0 & \text{otherwise.} \end{cases}$$
$$h(t) = \text{sinc}(F_s t)$$
$$\text{sinc}(x) = \frac{\sin \pi x}{\pi x}$$
$$h(t) = \text{sinc}(F_s t) = \frac{\sin(\pi F_s t)}{\pi F_s t}$$

Okay, which is magnitude  $T_s$  if  $F$  is less than or equal to magnitude  $F$  is less than or equal to  $F_s$  by 2 0 otherwise, okay. And now this low pass filter you can (re) you can find out the impulse response of this that is the time domain impulse response and the impulse response of this we already know this is a pulse remember if it's if the signal is a pulse in the time domain then in the frequency domain it is  $(\text{ })$ (20:17) and if you apply duality if it is a pulse in the frequency domain naturally the impulse response of the response in the time domain must be a sinc function and the response of this in the sinc domain that can be seen as follows that is that can be seen as follows.

From duality this can be easily obtained as this is sinc of the corresponding  $h$  of  $t$  equals sinc of  $F_s t$ , okay. So this is basically the sinc of  $F_s t$  that is a sinc function in the time domain this is a where sinc of  $x$  remember sinc of  $x$  equals  $\sin \pi x$  by  $\pi x$ , so sinc of  $F_s t$   $h$  of  $t$  equals sinc of  $F_s t$  equals  $\sin$  of  $\pi F_s t$  divided by  $\pi F_s t$  where  $F_s$  is the sampling frequency, okay.

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$$H(f) = \begin{cases} T_s & \text{if } |f| \leq \frac{F_s}{2} \\ 0 & \text{otherwise.} \end{cases}$$

This can be derived by duality.

$$h(t) = \text{sinc}(F_s t)$$
$$\text{sinc}(x) = \frac{\sin \pi x}{\pi x}$$
$$h(t) = \text{sinc}(F_s t) = \frac{\sin(\pi F_s t)}{\pi F_s t}$$

So this is the sinc function in the time domain and we can derive this and this can be derived by let us note that this can be derived by duality. And this can be derived by duality, okay. So this can be derived using this can be derived using duality, okay. And therefore what we have this can be derived by using the principle of duality we know that the principle of duality, okay. Now therefore we have the signal  $m(t)$ , so what we have to do to reconstruct the signal for reconstruction.

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$$h(t) = \text{sinc}(F_s t) = \frac{\sin(\pi F_s t)}{\pi F_s t}$$

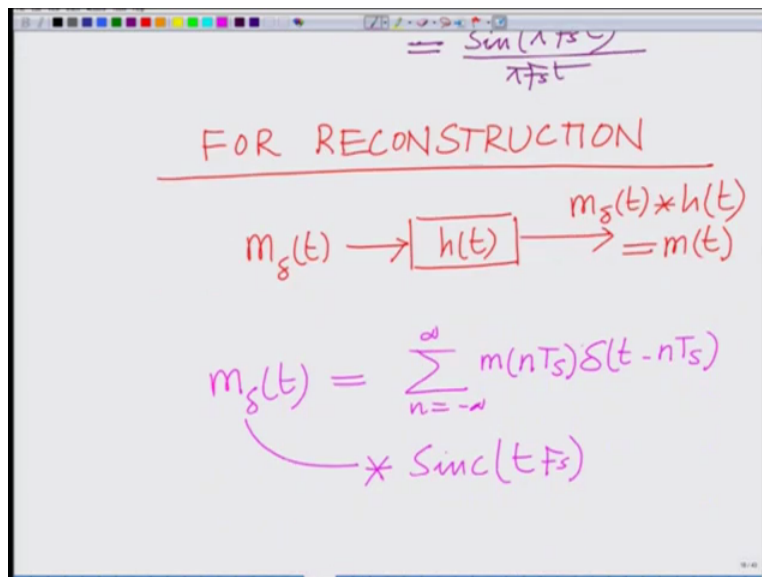
FOR RECONSTRUCTION

$$m_g(t) \rightarrow \boxed{h(t)} \xrightarrow{m_g(t) * h(t)} m(t)$$

Now let us highlight that procedure for reconstruction we take our sampled signal  $m_s(t)$  pass it through the system with impulse response  $h(t)$  and that gives this, so  $m_s(t)$  convolved with  $h(t)$  that gives back our original signal  $m(t)$  when there is no aliasing, right? Because we are low pass filtering to extract the component at 0 frequency that is the component between  $-\frac{F_s}{2}$  and  $\frac{F_s}{2}$  which is simply a replica of the spectrum of the original (sa) signal  $m(t)$  scaled by  $F_s$ .

So we are multiplying it by a low pass filter which is response which has which has a cut off frequency  $F_s/2$  and obviously to invite for that scaling to account for that scaling, we are scaling this by  $1/F_s$  or basically  $T_s$ , okay. That gives us back the original spectrum it gives us back the original spectrum in the frequency domain in the time domain naturally in the frequency domain we are multiplying so in the time domain it is basically convolution, okay.

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$$\frac{\sin(\pi f_s t)}{\pi f_s t}$$

FOR RECONSTRUCTION

$$m_s(t) \rightarrow [h(t)] \rightarrow m_s(t) * h(t) = m(t)$$

$$m_s(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \delta(t - nT_s)$$

$$\times \text{sinc}(tF_s)$$

And that can be expressed as follows, now if you look at that our  $m_s(t)$  the sampled signal is summation  $n$  equal to minus infinity to infinity  $m(nT_s) \delta(t - nT_s)$ , now I have to take  $m_s(t)$  and I have to convolve with sinc of  $tF_s$  that is sinc of  $F_s t$ , remember this is our reconstruction filter sinc of  $tF_s$  or basically synchronous  $F_s t$  and therefore what we can see is  $m(t)$  equals  $m_s(t)$  convolved with sinc of  $F_s t$  which is equal to basically, now we have our  $m_s(t)$  also I can write this as convolution is commutative, so I can write this as sinc of  $F_s t$  convolved I can interchange  $m_s(t)$  and sinc.

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Handwritten derivation on a whiteboard:

$$m(t) = m_s(t) * \text{sinc}(tF_s)$$

$$= \text{sinc}(F_s t) * m_s(t)$$

$$= \text{sinc}(F_s t) * F_s \sum_{n=-\infty}^{\infty} m(nT_s) \delta(t - nT_s)$$

There is a handwritten note at the top:  $m_s(t) \xrightarrow{n=-\infty}$  with an arrow pointing to the sinc function.

Continuation of the handwritten derivation on a whiteboard:

$$= F_s \sum_{n=-\infty}^{\infty} \text{sinc}(F_s t) * \delta(t - nT_s)$$

$$= F_s \sum_{n=-\infty}^{\infty} \text{sinc}(F_s(t - nT_s))$$

So this is sinc of  $F_s t$  convolved with  $m_s(t)$  we have already seen that is  $n$  equal to minus infinity to infinity  $m(nT_s)$  into  $\delta(t - nT_s)$  which is basically equal to which equals well,  $F_s$  taking the sinc inside that is  $n$  equal to minus infinity to infinity sinc of  $F_s t$  convolved with  $\delta(t - nT_s)$ . Now a function convolved with  $\delta(t - nT_s)$  the function is simply shifted to  $nT_s$  so that is  $F_s$  shifted by  $nT_s$ , so that is  $n$  equal to minus infinity to infinity sinc of  $F_s(t - nT_s)$  where we are once again using the property that is function  $x(t)$  convolved with



Delta t minus t not is simply x of t minus t not that is x of t shifted by t not, so sinc of Fst convolved by delta t minus nTs is simply sinc of Fs t minus nTs.

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$$\begin{aligned}
 &= \text{sinc}(F_s t) * m_s(t) \\
 &= \text{sinc}(F_s t) * f_s \sum_{n=-\infty}^{\infty} m(nT_s) \delta(t - nT_s) \\
 &= f_s \sum_{n=-\infty}^{\infty} m(nT_s) \text{sinc}(F_s t) * \delta(t - nT_s) \\
 &= f_s \sum_{n=-\infty}^{\infty} m(nT_s) \text{sinc}(F_s(t - nT_s))
 \end{aligned}$$

$F_s T_s = 1$

$$m(t) = f_s \sum_{n=-\infty}^{\infty} \text{sinc}(F_s t - n)$$

And now you will realize that  $F_s$  into  $nT_s$  is one because  $F_s$  is a sampling frequency  $F_s$  into  $T_s$  because  $T_s$  is 1 over  $F_s$ , so  $F_s$  into  $T_s$  equals 1, so if I use this principle over here what I get is summation  $n$  equal to minus infinity to infinity  $F_s$  sinc of  $F_s t$  minus  $F_s$  into  $nT_s$  that is simply  $n$  and that is the original reconstructed signal that is how we can reconstruct the original signal by passing it through the sinc filter or this is basically a filter whose response is basically this will be your I am sorry I am missing the the  $m(nT_s)$  there has to be an  $m(nT_s)$  and this is also there will be  $m$  of  $nT_s$ .

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Handwritten equation on a whiteboard:

$$m(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \text{sinc}(F_s(t - nT_s))$$

Annotations:

- $F_s T_s = 1$  (top right)
- Arrow pointing to the sinc function:  $\text{sinc}(F_s(t - nT_s))$
- Text below: "Shifting  $\text{sinc}(F_s t)$  to each  $nT_s$ , Scaling by  $m(nT_s)$  Followed by sum" (green)

So what we are doing is basically we are shifting, correct? we are shifting of course there is no  $F_s$  there is no  $F_s$  because this is simply your  $m(nT_s)$  into  $\Delta t$  minus  $T_s$ , so this is simply  $m$  of  $nT_s$  so anyway what we are doing is basically we are taking each sinc  $F_s$  shifting it to  $nT_s$  where as shifting sinc of  $F_s t$  to each  $nT_s$  to each  $nT_s$  scaling by  $m(nT_s)$  followed by the.

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Handwritten derivation on a whiteboard:

Scaling by  $m(nT_s)$ , Followed by sum,

$$\text{sinc}(F_s t) = \frac{\sin(\pi F_s t)}{\pi F_s t}$$

zeros are at  $kT_s$

$$= \frac{\sin(\pi F_s kT_s)}{\pi F_s kT_s}$$

$$= \frac{\sin(k\pi)}{k\pi} = 0.$$

So what we have essentially is basically and remember sinc of  $F_s t$  sinc of  $F_s t$  the zeros this is equal to basically  $\frac{\sin(\pi F_s t)}{\pi F_s t}$  (28:33)  $\sin$  of  $\pi F_s t$  divided by  $\pi F_s t$ . So this is zeros are at the zeros of

this are at integer multiples that is at  $k$  times  $T_s$  because this is at every  $k$  times  $T_s$  this is  $\sin$  of  $\pi F_s \pi F_s k \text{ times } T_s$  divided by  $\pi F_s k \text{ times } T_s F_s$  into  $T_s$  is one, so this is  $\sin$  of  $k \pi$  divided by  $k \pi$  which is 0. So the zeros are at that is at every integer multiples at integer multiples of Time scale.

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So therefore now what you are doing is essentially if you can look at this you have your you have your sampled signal  $m \Delta t$  which consist of impulses, correct? So this is your  $m$  of this is your sample which is basically your  $m$  of  $nT_s$ , so let us say  $n$  equal to 0 so this is your sample which is  $m$  of  $T_s$  this is your sample which is  $m$  of  $2T_s$  and so on and this is your sample which is  $m$  of minus  $T_s$  this is your sample which is  $m$  of  $m$  of minus  $2T_s$  and so on. Now what you are doing you are shifting the sinc to  $T_s$  scaling it by  $m$  of  $nT_s$  or  $m$  of 0 and then you are super imposing this sinc, okay.

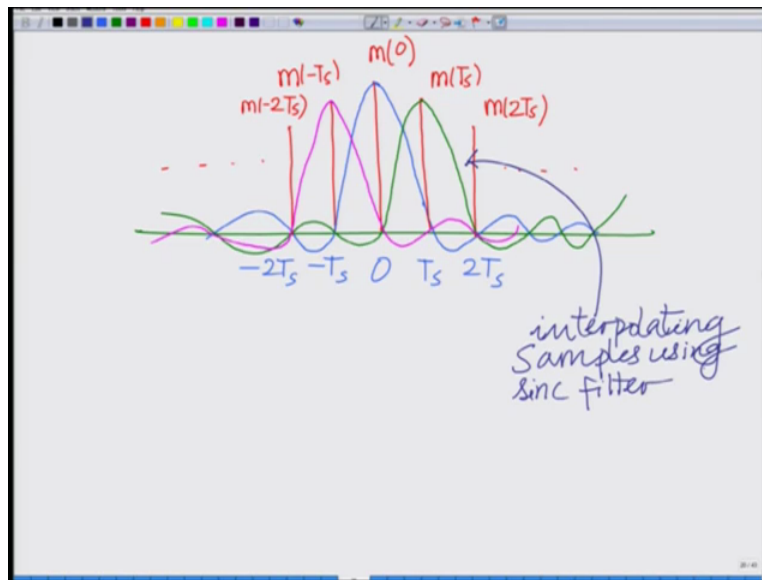
So you are taking the sinc scaling it by  $m$  of 0 sinc amplitude at 0 is one so you are scaling it by  $m$  of zero so now and remember that is one sinc whose zeros are at remember this, so this is  $T_s$ , this is 0, this is  $2T_s$ , okay. This at 0 so this is 0  $T_s$ ,  $2T_s$ , minus  $T_s$ , minus  $2T_s$  and now when you scale it by  $m$  0 remember the zeros are still at  $T_s$ ,  $2T_s$  and so on. So what you are getting is the sinc which are which is like this that is the zeros are at basically the various samples.

And now then you are taking a second sinc scaling by  $m$  of  $T_s$  and the zeros of this sinc will again be at this is an important property of the **the** zeros of this which is basically scaled by  $2T_s$  are going to be at scaled at  $T_s$  is going to be a  $2T_s$  0 and so on and then you are taking another sinc shifting it to minus  $T_s$  and the zeros of this are going to be at these are going to be at well, minus  $T_s$  is going to have zeros and so on and what you are observing is basically from this figure, what you can observe? is basically the reconstruction is nothing but a some of shifted sincs.

You are taking a each sinc function sinc of  $F_s t$  shifting it to  $n$  times  $T_s$  and scaling it by the value of the sample that is  $m$  of  $n$  times  $T_s$ , okay. Scaling it by the sample  $m$  of  $n$  times  $T_s$  and then basically summing of all summing of all these shifted scaled sinc copies and the interesting properties are the zeros that is wherever your scaling it by  $n$  of  $T_s$  wherever you are the shifting the sinc scaling it by  $m$  of  $nT_s$  shifting it to  $n$  of  $n$  of  $T_s$  and scaling it by  $m$  of  $nT_s$  the zeros of this sinc coincide with basically the other samples, so it is not affecting the other samples so the reconstruction at the samples exactly is not affected and that is true because if you have to reconstruct the original signal the original signal back then we know the values at the samples they are simply  $m$  of 0  $m$  of  $T_s$   $m$  of  $nT_s$  and so on, alright.

So the reconstruction is not affecting the samples at the locations, alright. And this is an interpolation, alright. What we are doing essentially we have this samples and we are taking we are interpolating these samples to create the original signal original signal  $m(t)$  and this is interpolation using the sinc filter.

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So what we are doing is literally, basically we are interpolating using the sinc filter this process is basically we are interpolating the samples. We are interpolating the samples using the sinc filter and what is our reconstruction?

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$$m(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \text{sinc}\left(\frac{F_s}{T_s}(t - nT_s)\right)$$

interpolating samples using sinc filter

Shift  $\text{sinc}(F_s t)$  to each  $nT_s$   
Scale by  $m(nT_s)$   
Sum all shifted copies of  $\text{sinc}(F_s t)$

If you look at this relation over here that is  $m(t)$  equals summation  $n$  equal to minus infinity to infinity  $m(nT_s) \text{sinc}(F T_s \text{ minus } nT_s)$  that is this is basically shift that is basically shifts  $\text{sinc}(F T_s)$  to each  $nT_s$  scaled by  $m(nT_s)$  and sum all shifted copies of this function  $\text{sinc}(F t)$  sorry this is

sinc of  $F_s t$  sinc of  $F_s t$  sinc of  $F_s t$  minus  $F_s t$  minus  $nT_s$ , correct? Sinc of  $F_s t$  minus  $nT_s$  so you are shifting this sinc of  $F_s t$  scaling by  $m(nT_s)$  and sum of all shifted copies of the function sinc of  $F_s t$  and that is what you are doing.

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The slide shows the following equation and annotations:

$$m(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \text{sinc}\left(\frac{F_s}{T_s}(t - nT_s)\right)$$

Annotations on the slide:

- Top right: reconstruction samples using sinc filter
- Left of the equation: sinc interpolation and interpolation filter is sinc filter.
- Right of the equation: Shift sinc( $F_s t$ ) to each  $nT_s$ , Scale by  $m(nT_s)$ , and Sum all shifted copies of sinc( $F_s t$ )

And this is basically your sinc interpolation or basically interpolation sinc is the interpolation filter, so interpolation filter is sinc function has a sinc response basically in the time domain. This is the interpolation filter the sinc function sinc of  $F_s t$  is nothing but the response of the interpolation filter, alright. So basically what we have illustrated is how to reconstruct the original signal, which is very simple when we do not have aliasing when  $F_s$  is greater than  $2 F_m$  then we can low pass filter it with cut-off frequency  $F_s$  by 2 and to invert the scaling we can have the height of the filter the magnitude response of the filter is  $1$  over  $F_s$  that is  $T_s$ , alright.

The corresponding response in the time domain is sinc  $F_s t$ , so you are passing the sampled signal through this filter with response sinc of  $F_s t$  we are able to interpolate the samples, alright. Interpolate the samples of the sampled signal and basically reconstruct the original message signal  $m(t)$  without any distortion since there is no aliasing, alright. So that is that concludes the reconstruction aspect of the original signal, alright. So will stop here and look at other aspects in the subsequent modules, thank you.