

Principles of Communication- Part I
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Module No 6

Lecture 36

Spectrum of Sampled Signal, Aliasing and Nyquist Sampling Theorem

Hello welcome to another module in this massive open online course, so we are looking at the sampling of an analog signal to (con) to extract samples from it to convert it to a digital signal, alright. And we have said that the signal can be sampled by multiplying it by an by an impulse train, alright which is termed as ideal sampling, alright.

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The image shows a whiteboard with handwritten mathematical derivations for the sampled signal. At the top, the sampling function is defined as $g_s(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_s)$. Below this, the sampled signal is derived as follows:

$$\begin{aligned} \text{Sampled signal:} &= m(t) \times g_s(t) \\ &= m(t) \times \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \\ &= \sum_{n=-\infty}^{\infty} m(t) \delta(t - nT_s) \end{aligned}$$

Arrows indicate the relationship between the terms: an arrow from "original signal" points to $m(t)$, and an arrow from "Sampled signal" points to the final sum. Below the sum, the equation is further simplified for a specific sample $n=0$:

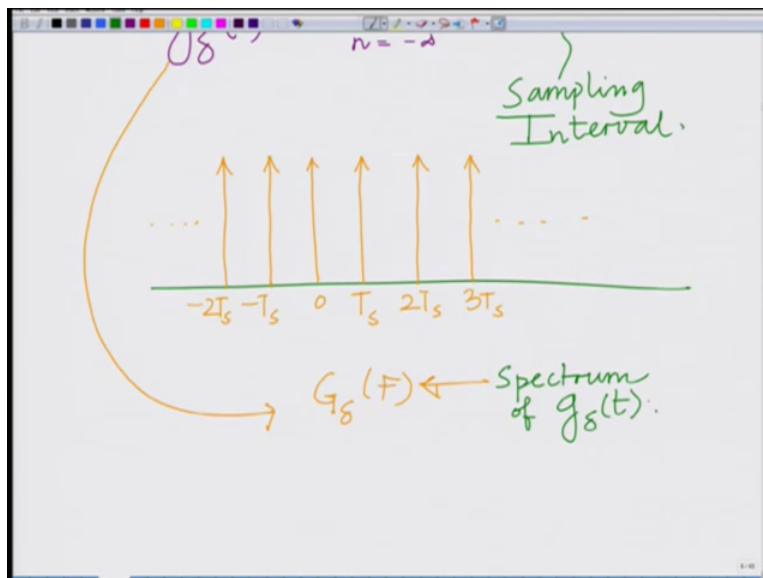
$$\begin{aligned} &= m(nT_s) \delta(t - nT_s) \\ &= m(t) \delta(t - t_0) \\ &= m(t_0) \delta(t - t_0) \end{aligned}$$

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

Sampling Interval.

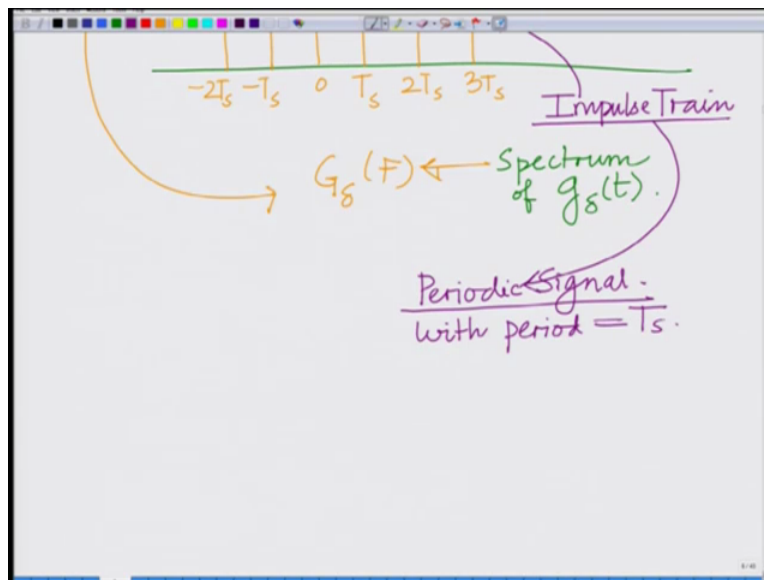
So let us say let us consider our impulse train which is given by $g_s(t)$ equals summation from $n = -\infty$ to ∞ of $\delta(t - nT_s)$ where T_s is the sampling interval, alright which is basically the time interval between the successive sampling instants, okay. Delta is of course the direct Delta function this is basically we have our sequence of impulses at train of impulses at every multiple of Time scale.

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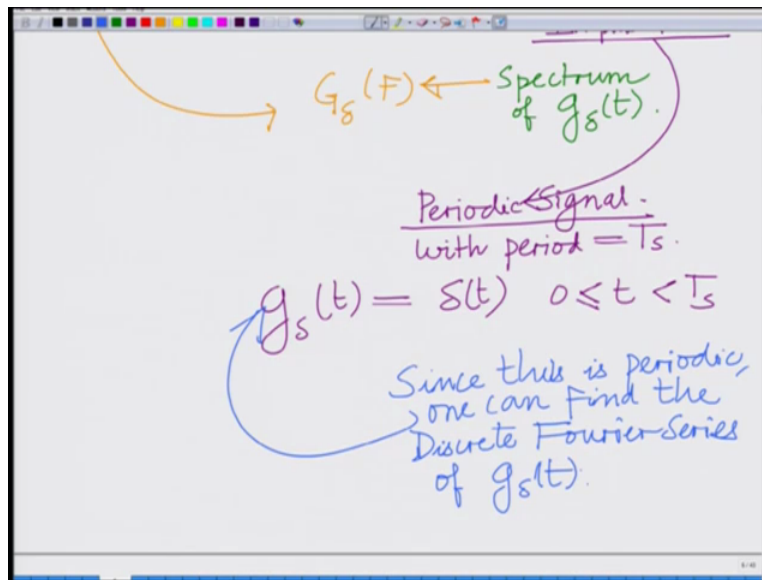
So we have impulse at T_s we have impulse so this is our impulse at T_s we have impulse at $2T_s$, we have impulse at $3T_s$ and so on and of course we have impulse at 0 we have impulse we have an impulse at we have an impulse at minus T_s , minus $2T_s$ etc and now what we want to do is we want to find the spectrum $G_s(f)$ that is the spectrum $G_s(f)$ which is basically your spectrum of $g_s(t)$, okay. We want to start with that towards eventually find the spectrum of the spectrum of the sampled signal.

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Now you will see that this signal if you can if you look at this signal this impulse train, correct? This impulse train this impulse train you will realize this is a periodic signal, correct? The impulse train is a periodic signals is there is one impulse every T_s . So we have impulse at T_s impulse at 0 impulse at T_s , $2T_s$, minus T_s , minus $2T_s$ and so on, alright. So this is a periodic signal with period T_s , okay. So this is a periodic signal with period equal to T_s , okay.

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So in one period let us say this is basically your $x(t)$ or your g delta t equals basically Delta t 0 less than equal to t less than T_s , alright this is the signal and therefore since this is periodic we can find the Fourier discrete Fourier (se) series representation, since this is periodic since this is periodic one can find the one find the discrete Fourier series of g delta t , okay.

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Handwritten equation on a whiteboard:

$$g_s(t) = \sum_{k=-\infty}^{\infty} c_k \cdot e^{j2\pi k f_s t}$$

Below the equation, it is noted that $f_s = \frac{1}{T_s}$.

So let us express g delta t so g delta t can be expressed as summation well, k equals minus infinity to infinity $C_k e$ to the power of $j 2\pi k f_s t$ where f_s equals 1 over T_s the sampling

frequency is the fundamental frequency, alright. Fundamental frequency is 1 over time period, correct? The fundamental frequency is 1 over the time period the time period is nothing but the sampling interval, right? Sampling interval that is T_s therefore the sampling frequency F_s itself is the fundamental frequency, okay.

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Handwritten whiteboard showing the definition of the sampling coefficient C_k . At the top right, it states $f_s = \frac{1}{T_s}$ with an arrow pointing to it. Below this, it says "Fundamental Freq = Sampling Frequency". The main equation is $C_k = \frac{1}{T_s} \int_{-T_s/2}^{T_s/2} g_s(t) dt$.

Handwritten whiteboard showing the derivation of the sampling coefficient C_k . It starts with the same equation as the previous slide: $C_k = \frac{1}{T_s} \int_{-T_s/2}^{T_s/2} g_s(t) dt$. Above this, it says "Fundamental Freq = Sampling Frequency". Below the integral, it adds the exponential term $e^{-j2\pi k f_s t}$ inside the integrand. The equation then becomes $C_k = \frac{1}{T_s} \int_{-T_s/2}^{T_s/2} g_s(t) e^{-j2\pi k f_s t} dt$. Finally, it shows the substitution $s(t) = g_s(t)$, resulting in $C_k = \frac{1}{T_s} \int_{-T_s/2}^{T_s/2} s(t) e^{-j2\pi k f_s t} dt$.

So the fundamental frequency equals the equals the sampling frequency, alright. And further now we have to find these coefficients C_k and you can see that this coefficient each C_k equals well, 1 over T_s we know the expression for C_k 1 over T_s integral 0 to T_s , correct? $G \Delta t$ or we can

make it from minus $T_s/2$ to $T_s/2$ $g \Delta t \, dt$ which is equal to $1/T_s$ minus $T_s/2$ to $T_s/2$ Δt sorry $g \Delta t$ e to the power of $-j 2\pi k F_s t$ e to the power of e to the power of $-j 2\pi k F_s t \, dt$, alright there we have replaced $g \Delta t$ by Δt that interval minus $T_s/2$ to $T_s/2$.

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The image shows a whiteboard with handwritten mathematical derivations. At the top, the integral for the k th coefficient is written as:

$$= \frac{1}{T_s} \int_{-T_s/2}^{T_s/2} s(t) e^{-j 2\pi k F_s t} dt$$

Below this, the expression for the k th coefficient is shown with the integral limits and the exponential term:

$$k^{\text{th}} \text{ coefficient} = \frac{1}{T_s} \cdot e^{-j 2\pi k F_s t} \Big|_{t=0}$$

Then, the final simplified expression for the coefficient C_k is boxed:

$$C_k = \frac{1}{T_s} \times 1 = \frac{1}{T_s}$$

There are additional handwritten notes on the left side of the box: "kth coefficient in Discrete Fourier Series of $g(t)$ ".

Now the integration multiplying by Δt is very simple that is when I multiply a function by Δt and integrate that extracts the value of the function at t equal to 0. So this is simply e to the power of $-j 2\pi k F_s t$ evaluated at t equal to 0 that is simply 1 over that is simply 1 so this quantity is simply $1/T_s$ into 1, so $1/T_s$ so each C_k has a very simple expression each C_k this is the k th coefficient in the discrete Fourier series k th coefficient in the discrete Fourier series of $g \Delta t$ that is your that is basically your that is basically your C_k , so we have evaluated each coefficient C_k as $1/T_s$.

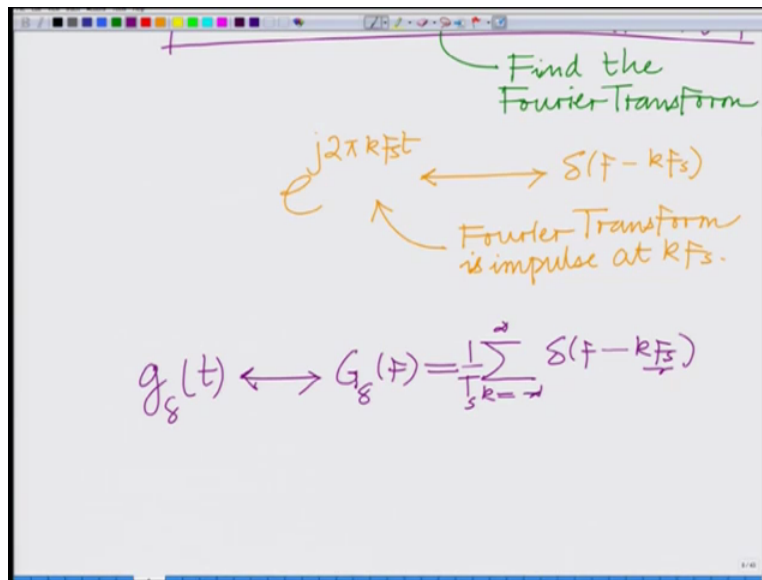
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$$g_{\delta}(t) = \sum_{k=-\infty}^{\infty} \frac{1}{T_s} \cdot e^{j2\pi k F_s t}$$
$$g_{\delta}(t) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} e^{j2\pi k F_s t}$$

Find the Fourier Transform

And therefore now the discrete Fourier series representation of this impulse train is basically given as I have $g_{\delta}(t)$ equals summation k equals minus infinity to infinity $\frac{1}{T_s} e^{j2\pi k F_s t}$ which is also basically you can write it since $\frac{1}{T_s}$ is common so you can take it outside of the summation $\frac{1}{T_s} \sum_{k=-\infty}^{\infty} e^{j2\pi k F_s t}$ this is your discrete Fourier series. This is basically your discrete Fourier series this is basically the discrete Fourier series representation okay, correct?

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Find the Fourier Transform

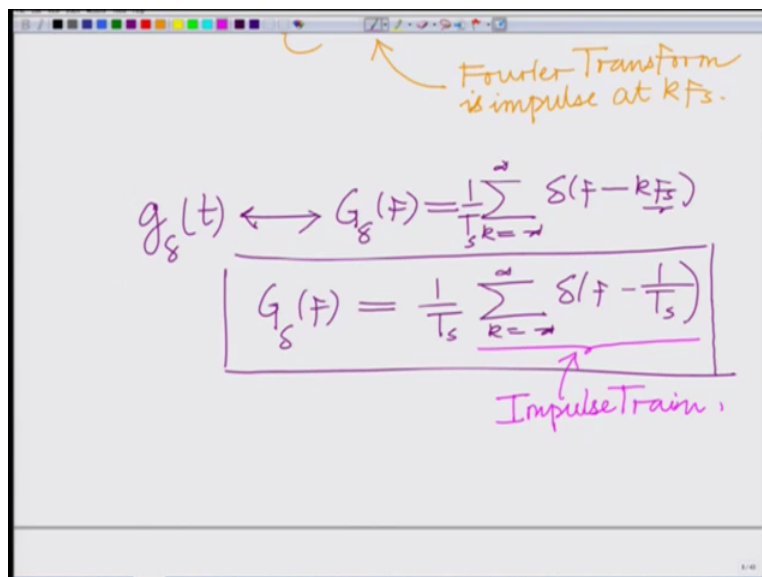
$e^{j2\pi k F_s t} \longleftrightarrow S(f - k F_s)$

Fourier Transform is impulse at $k F_s$.

$g_s(t) \longleftrightarrow G_s(f) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} S(f - k F_s)$

And now from this discrete Fourier series representation we can find the Fourier transform so from this find the Ft or the Fourier transform and that is simple one can extract the Fourier transform because $e^{j2\pi k F_s t}$ if you look at $e^{j2\pi k F_s t}$ $e^{j2\pi k F_s t}$ is nothing but an impulse this Fourier transform of this is an impulse at $k F_s$, correct? Fourier transform is an impulse at $k F_s$.

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Fourier Transform is impulse at $k F_s$.

$g_s(t) \longleftrightarrow G_s(f) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} S(f - k F_s)$

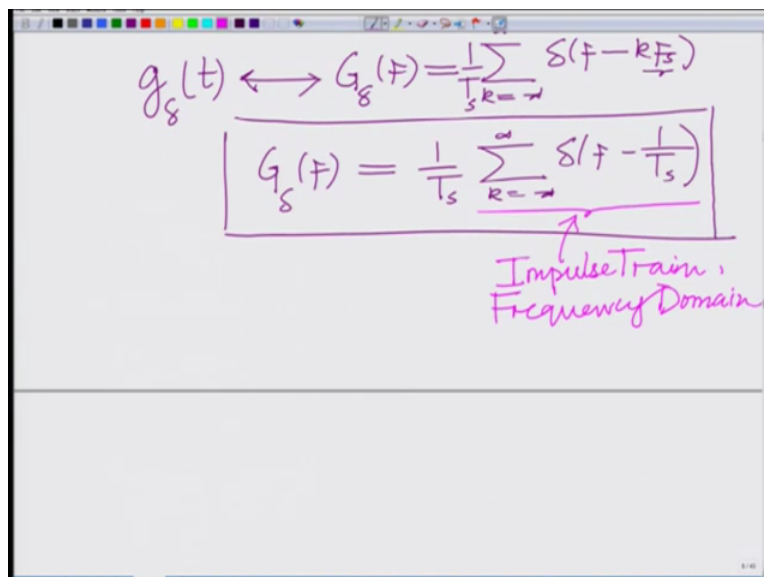
$G_s(f) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} S(f - \frac{1}{T_s})$

Impulse Train

The Fourier transform or basically spectrum is an impulse at kF_s which means now the Fourier transform of this is basically $g_\delta(t)$ is basically each Fourier transform of each is basically an impulse ΔF minus kF_s therefore the Fourier transform $g_\delta(t)$ which is basically the Fourier transform of $g_\delta(t)$ is basically, so if you look at $g_\delta(t)$ and let us say we consider it's Fourier transform capital $G_\delta(F)$ that is simply summation k equal to minus infinity to infinity 1 over well, there is a 1 over T_s each e to the power of $j 2\pi F k F_s$ kF e to the power of $j 2\pi k F_s t$ has a spectrum ΔF minus kF_s , okay.

And F_s remember F_s is basically 1 over T_s , so this is basically nothing but 1 over T_s summation k equal to minus infinity to infinity ΔF minus 1 over T_s , okay. And therefore you can see something very interesting which is basically if you look at this summation if you look at this summation 1 over T_s summation k equal to minus infinity to infinity ΔF minus 1 over T_s this is basically this is basically another impulse train.

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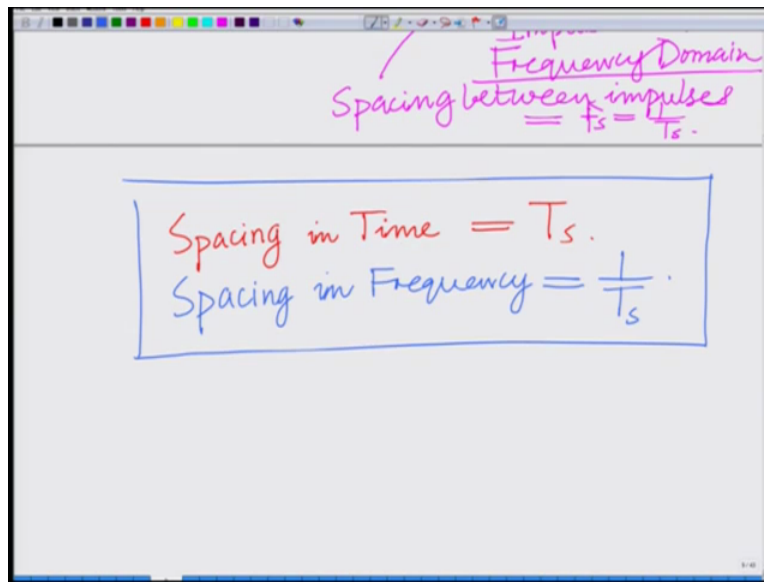
$$g_s(t) \leftrightarrow G_s(F) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} \delta\left(f - \frac{k}{T_s}\right)$$

$$\boxed{G_s(F) = \frac{1}{T_s} \sum_{k=-\infty}^{\infty} \delta\left(f - \frac{1}{T_s}\right)}$$

Impulse Train, Frequency Domain

Now you can see that this except for the scaling factor of 1 over T_s this is basically another impulse train and this is a train of impulses in the frequency domain, alright. So we have started with the train of impulses, impulse train in the time domain and what we are seeing is its Fourier transform or spectrum is basically an impulse train in the frequency domain impulse train in the frequency domain is an impulse train in the frequency domain.

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Further the spacing between the impulses that is each impulse I am missing a factor of k over here this is F minus k over T_s , so spacing between impulses equals F_s equals 1 over T_s remember spacing in time it was T_s the impulses are spaced at because T_s is a sampling interval, alright. So the impulses are spaced T_s apart but in frequency impulses are spaced, so spacing in frequency equals spacing in frequency equals this equals 1 over T_s , alright.

So spacing in for x as the (spaco) spacing in time is T_s , spacing in frequency is 1 over T_s so basically as the sampling as the sampling interval T_s increases the spacing in time increases the spacing in frequency that is 1 over T_s decreases because 1 over T_s is the sampling frequency as the sampling interval increases F_s that is a sampling frequency decreases, so spacing in frequency is the sampling frequency alright a sampling time sampling interval is increasing the sampling frequency is decreasing therefore spacing in the frequency domain is decreasing.

And conversely as the spacing in the and vice versa, right? As the spacing in the time decreases there is your sampling at finer at a (fi) at a very fine at finer and finer interval, right? that is T_s is decreasing the sampling interval T_s is decreasing which means the sampling (increa) the sampling frequency F_s is increasing the spacing in frequency increases, alright. So this is spacing in frequency is T_s equals 1 over T_s equals to F_s .

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The image shows a digital whiteboard with handwritten notes. At the top, a box contains two equations: "Spacing in Time = T_s ." and "Spacing in Frequency = $\frac{1}{T_s} = F_s$ ". Below the box, there are two lines of text. The first line, in green, says "As spacing in time increases, Spacing in Freq. decreases". The second line, in blue, says "As spacing in time decreases, Spacing in freq. increases.".

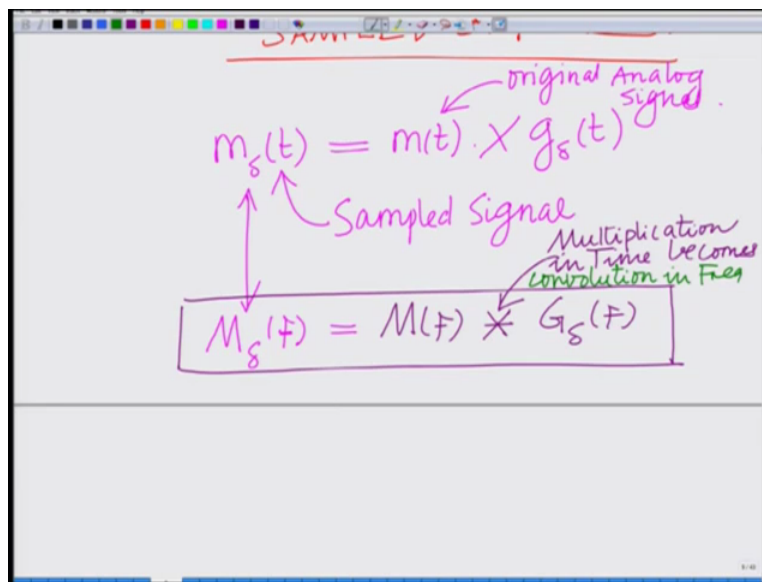
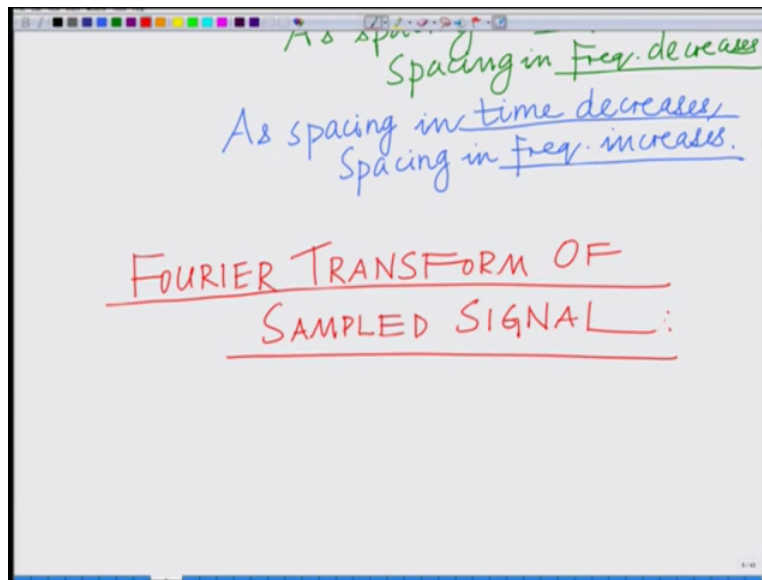
$$\text{Spacing in Time} = T_s.$$
$$\text{Spacing in Frequency} = \frac{1}{T_s} = F_s$$

As spacing in time increases,
Spacing in Freq. decreases

As spacing in time decreases,
Spacing in freq. increases.

So you can note that as spacing in time (decrea) increases spacing in the frequency domain decreases, okay. As spacing in time increases spacing in frequency domain increases and vice versa. As spacing in time decreases as spacing in time decreases spacing in the frequency that basically increases, alright. So that is basically it means that as your sampling finer and finer in time that is T_s is decreasing the spacing in frequency that is the sampling frequency F_s is increasing therefore the spacing between these successive impulses of the impulse train in the frequency domain is increasing.

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Now let us look at what is the Fourier transform of the sampled function, alright. So now we are in a position to look at the Fourier transform of the sampled functions, so we want to look we have found the Fourier transform of the of the impulse train now Fourier transform of the Fourier transform of the sampled analog signal, now you can see that will be given as follows remember $m \Delta t$ or sampled analog signal is simply m of t times $g \Delta t$ this is your sampled analog signal.

This is your sampled signal and $m(t)$ is the original analog signal, your original this is your original analog signal and now if I look at the Fourier transform of this, correct? That is very simple $M_s(f)$ is the Fourier transform of $m(t)$ that is the sampled signal this is equal to now you can see $m(t)$ let us say has a Fourier transform $M(f)$ and $\delta(t)$ we know has a Fourier transform $\delta(f)$ this will be the convolution between the Fourier transform $M(f)$ and $\delta(f)$ because multiplication in the time domain is convolution in the frequency domain. So multiplication in time becomes convolution multiplication in time becomes convolution in the frequency domain.

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$$M_s(f) = M(f) * F_s \sum_{k=-\infty}^{\infty} \delta(f - kF_s)$$

$$M_s(f) = F_s \sum_{k=-\infty}^{\infty} M(f) * \delta(f - kF_s)$$

$$M_s(f) = F_s \sum_{k=-\infty}^{\infty} M(f - kF_s)$$

$$M(f) * \delta(f - F_0) = M(f - F_0)$$

Therefore $M_s(f)$ that is a spectrum of the sampled signal equals $M(f)$ convolved with $\delta(f)$ $\delta(f)$ we have seen is $1/T_s$ summation that is $1/T_s$ summation $\delta(f - kF_s)$ and $1/T_s$ is also equal to remember $1/T_s$ is also equal to F_s . So this is basically your convolution with F_s times summation k equals minus infinity to infinity, correct?

k equals minus infinity to infinity $F_s \delta(f - kF_s)$ where F_s is the sampling frequency and this is equal to well, F_s summation k equal to minus infinity to infinity $M(f)$ convolved with $\delta(f - kF_s)$ which is F_s now k equal to minus infinity to infinity $M(f)$ convolved with $\delta(f - kF_s)$ remember this is nothing but $M(f - kF_s)$, so this is $M(f - kF_s)$ that is your M that is your $M_s(f)$ $M_s(f)$ is basically k equal to minus infinity to infinity $M_s(f)$ $M_s(f)$ is F_s

summation k equal to minus infinity to infinity $M(f)$ minus $M(f)$ minus kF_s $M(f)$ minus kF_s $M(f)$ minus kF_s that is what you are doing here is something very interesting, alright.

First let me elaborate, the property that we have used here again is something that should be very familiar to you that is $M(f)$ convolved with $\delta(f - F)$ not (shim) simply shifts $M(f)$ to F not, so $M(f)$ convolved with each $\delta(f - kF_s)$ simply shifts each $M(f)$ to kF_s and therefore now what you are doing is you are shifting the spectrum $M(f)$ to each kF_s , alright. So for at each kF_s that is k times multiple of the sampling frequency F_s , right?

You have a copy, alright of the original spectrum $M(f)$ so now you are shifting the spectrum $M(f)$ to each kF_s and you are summing all these copies, so the spectrum of the sampled signal will be the sum of the copies of the original spectrum $M(f)$ shifted to every multiple of the sampling frequency F_s , so what you are seeing here is something very interesting, what you have is?

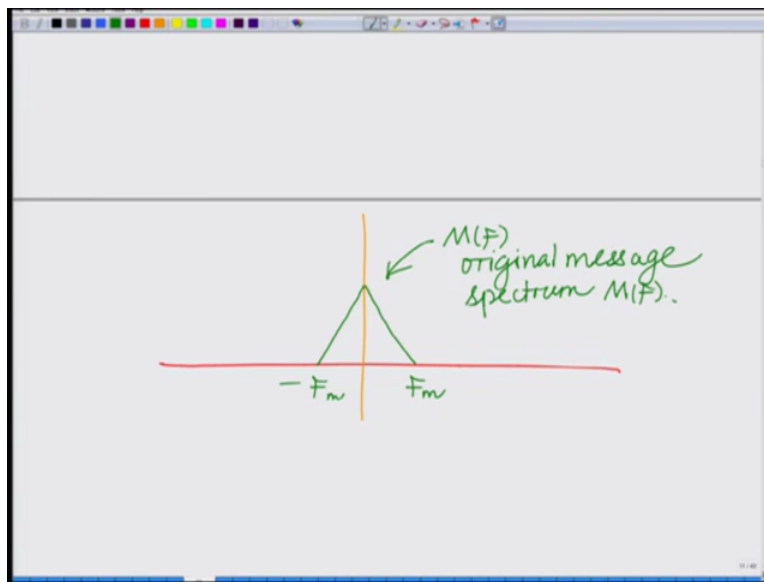
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The image shows a handwritten derivation on a digital whiteboard. At the top, the equation for the spectrum of a sampled signal is written: $M_s(f) = f_s \sum_{k=-\infty}^{\infty} M(f - kF_s)$. Below this, a convolution property is shown: $M(f) * \delta(f - F_0) = M(f - F_0)$. An arrow points from this property to the summation term in the main equation. Below the equation, a note states: "Spectrum of sampled signal = sum of all copies of the original message spectrum $M(f)$ shifted to every kF_s ". To the left of this note, another note says: " k^{th} multiple of sampling frequency F_s ".

This is the (orig) this is the spectrum of sampled signal equals sum of all copies all copies of the original message spectrum $M(f)$ shifted to every kF_s where kF_s remember kF_s is k th basically this is the k th multiple of, this is the k th multiple of the sampling, so this is the sum of all copies of the message spectrum shifted to every kF_s which is very interesting, so you have the original spectrum $M(f)$ now we are shifting this to each kF_s , right?

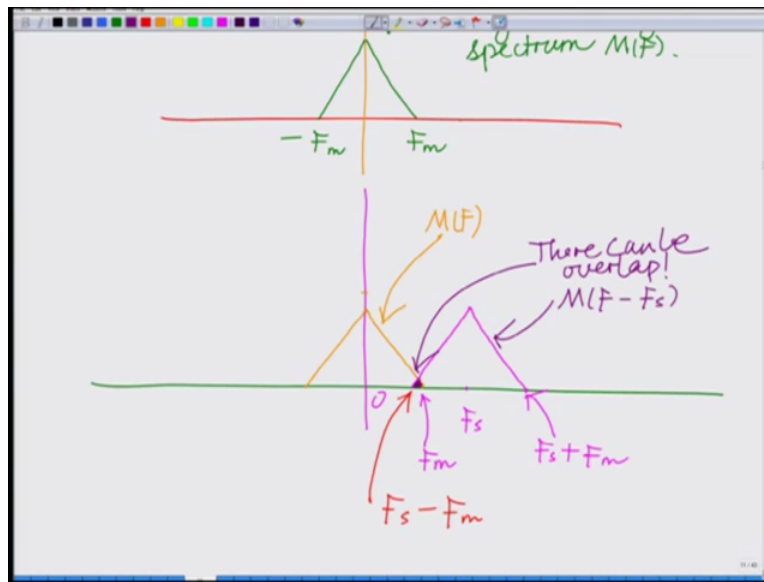
Each multiple of the sampling (freq) each integer multiple of the sampling frequency F_s therefore you are making an infinite number of copies of the original spectrum $M(f)$ and then you are adding all these copies, now let us see in the frequency domain what does what this corresponds to, so you have your original spectrum $M(f)$, okay. So let us say you have your original spectrum this can lead to something very interesting in the frequency domain you have your original spectrum $M(f)$, correct?

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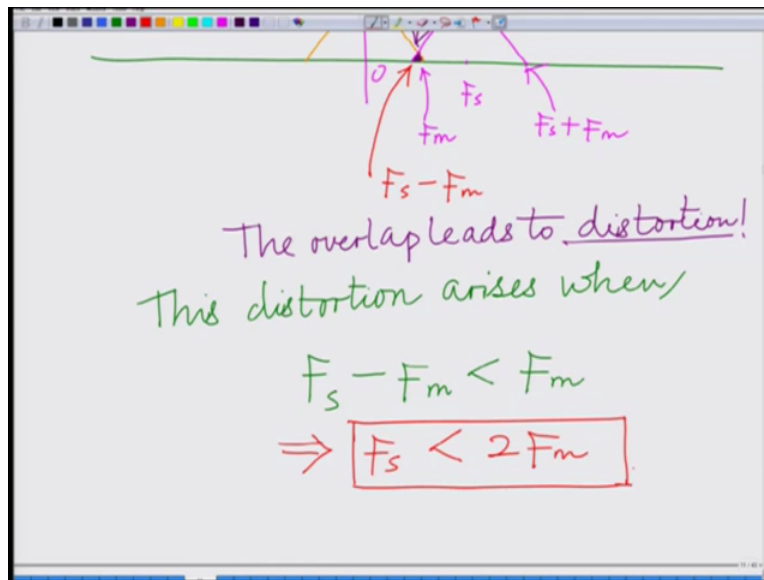
You have the original spectrum $M(f)$, okay. Let us say the maximum frequency is F_m between minus F_m , so this is your original message original message spectrum My family consists of.

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Now let us draw here the sampled spectrum so we are going to have well, we are going to have the original message spectrum intact that is going to be intact, so we are going to have the original message spectrum that is going to be intact and so that is going to be this is going to be $M(f)$. In addition at F_s you will have one copy at F_s , correct? So this is the copy at 0 in addition you will have one copy which is shifted to F_s , correct? So that will be that copy will be, so this point is your F_m now that is a copy at F_s so this point naturally is F_s plus F_m and this point is interesting if you look at this, this is F_s minus F_m . Now that can potentially be an overlap in this region this to copies so this is your $M(f)$ this is your $M(f)$ minus $M(f)$.

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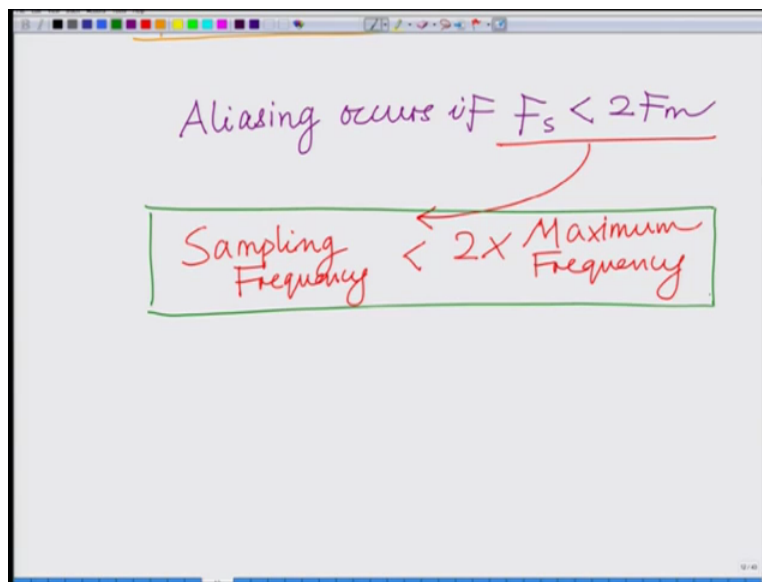
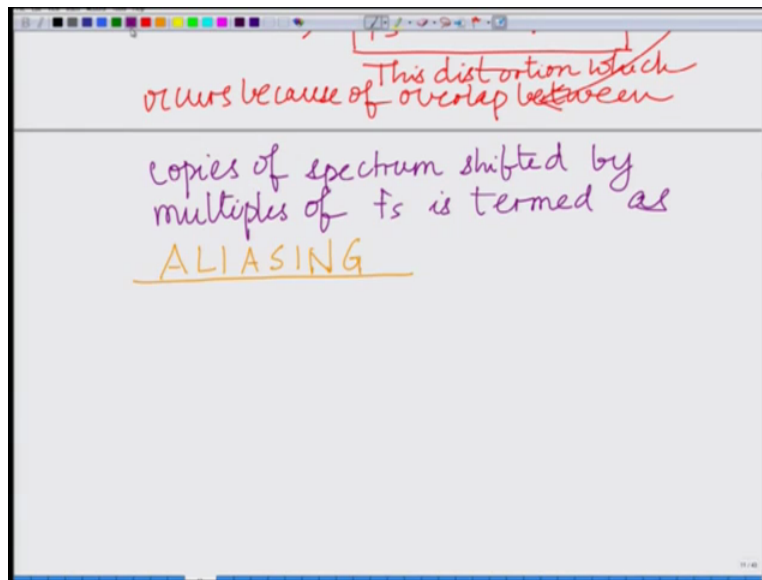


So there can be there can be an overlap between these 2 copies, so MF now if you can see this is interesting so there can be an overlap and the overlap the overlap leads to distortion, so you have a copy of the original spectrum that is another you have the original spectrum its 0 there is another copy shifted by F_s when you take the sum these 2 copies an overlap and therefore once you sum that can lead to distortion, alright.

And this can cause distortion of the sampled that is the (si) the spectrum of the sampled signal is not similar to the (spect) the spectrum of the sampled signal that is $m \Delta t$ that is $m \Delta F$ is not similar to the spectrum MF of the original signal it is a distortion version of the (spect) distorted version of the spectrum of the original signal MF, alright. And when does this arise this distortion arises if there is an overlap, now realize that this distortion arises when you can see when will there be an overlap? There will be an overlap when F_s minus F_m is less than F_m F_s minus F_m is less than F_m which implies F_s is less than twice F_m .

So these spectral the spectral overlap so this is a very important and very interesting condition the spectral overlap, alright. The spectral overlap between these different copies of the spectrum shifted by the various integer multiples of F_s will occur if the sampling frequency F_s is less than twice the maximum frequency F_m of the original analog signal, alright and therefore at this distortion which occurs because of the spectral overlap is termed as Aliasing, okay.

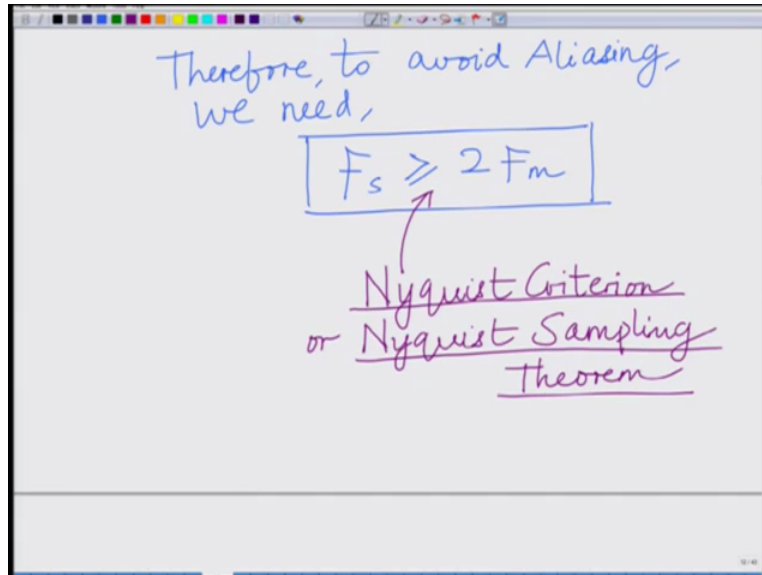
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So this distortion this distortion which occurs because of overlap between copies of spectrum shifted by multiples of F_s is termed as is termed as Aliasing, this is termed as Aliasing. So this distortion which occurs because of this overlap this distortion is termed as Aliasing and aliasing occurs if F_s less than twice F_m that is aliasing occurs this is a very important condition aliasing occurs if F_s less than twice F_m that is another way of stating this is basically sampling frequency is less than twice the frequency of the your sampling frequency is less than twice the maximum,

sampling frequency is less than twice the maximum frequency of the signal being sampled. That is twice the maximum frequency of your $m(t)$.

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Therefore to avoid aliasing we need F_s greater than F_m therefore to avoid aliasing to avoid aliasing or aliasing distortion we need F_s greater than or equal to twice F_m and this is termed as a this is termed as a Nyquist this is termed as a Nyquist criterion for sampling to avoid aliasing this is termed as the this has many names this is termed as the Nyquist criterion this is termed as a Nyquist criterion or your Nyquist, this is termed as a Nyquist criterion or the Nyquist sampling theorem which basically says that for to avoid aliasing which arises from the overlap of the spectral copies of the shifted spectral copies at the different multiples of the sampling frequency F_s one has to sample at a sampling frequency F_s greater than twice F_m where F_m is the maximum message frequency, alright.

Maximum message frequency F_m you can think of it also as the bandwidth of the original signal you have to sample at great sampling frequency greater than twice the bandwidth this is termed as the Nyquist criterion or the Nyquist sampling theorem which states that to avoid aliasing one has to sample at F_s greater than twice F_m , alright. And therefore what we have seen is something very interesting that the original sample the original the original sampled signal can be a distorted version of the original signal and that distortion occurs because these different spectral copies that is the spectrum of the sampled signal which comprises of the sum of the spectral

copies which are shifted by different multiples of the sampling frequency various shifted by all possible multiples of the all possible integer multiples of the sampling frequency F_s , alright.

And because that can be possible overlap between these spectral copies that results in distortion which is termed as aliasing distortion occurs if F_s is less than twice F_m therefore to avoid this distortion we need sampling frequency greater than or equal to twice F_m this is termed as the this condition, alright this result is basically termed as a Nyquist (cri) this criterion or this condition required that is the minimum sampling frequency required to avoid aliasing is termed as a Nyquist criterion and this result is termed as a Nyquist sampling theorem, alright. So we will stop here and look at other aspects in the subsequent modules, thank you.