

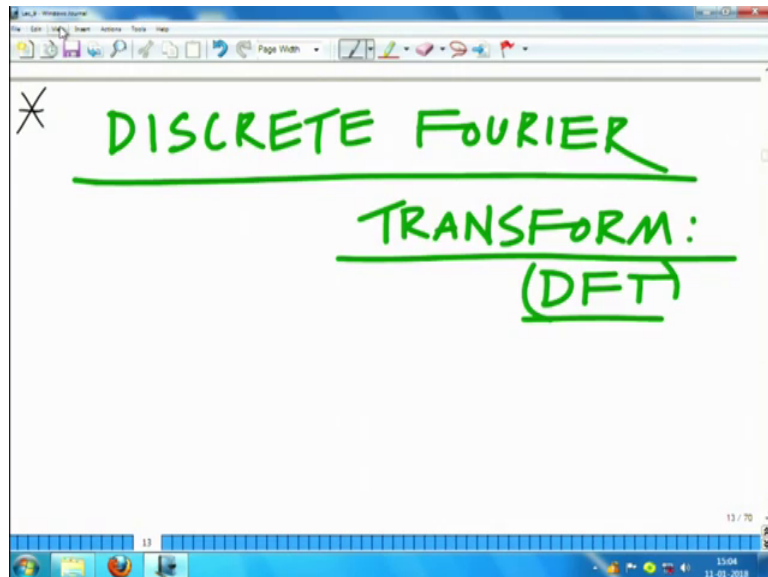
Principles of Signals and Systems
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Lecture - 61

Discrete Fourier Transform: Properties – Conjugation, Frequency Shifts, Duality, Circular Convolution, Multiplication, Parseval's Relation, Example Problems for Fourier Analysis of Discrete Time Signals

Hello, welcome to another module in this massive open online course. So, we are looking at the DFT; that is the discrete Fourier transform all right, and its application for finite length in discrete time sequence.

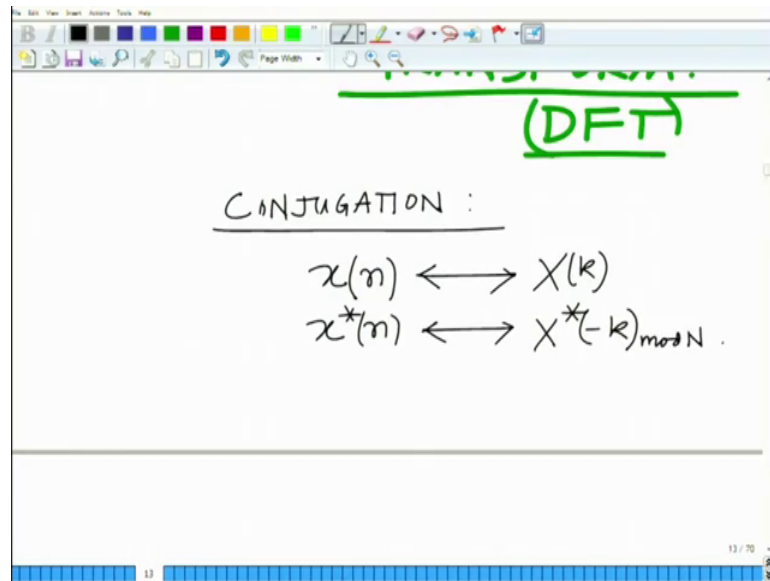
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So, we are looking at the discrete Fourier transform which as I have already alluded to, is one of the most important tools which has revolutionized signal process.

So, this is the discrete Fourier transform. So, we have already seen and this is the DFT and in the DFT what we have is, we have the DFT of the finite length sequence and we are looking at the properties of the DFT.

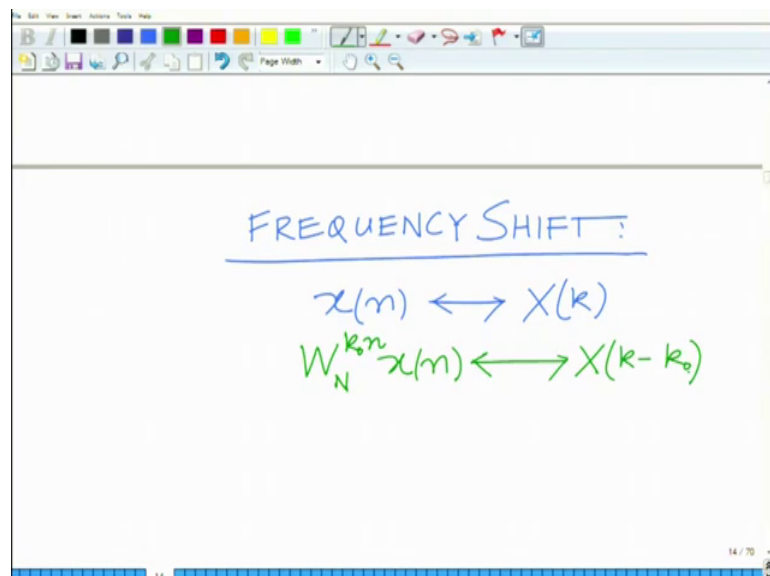
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Let us continue our discussion with the properties. So, this is the next property that is conjugation. So, x ; so we have a sequence x_n which has the DFT

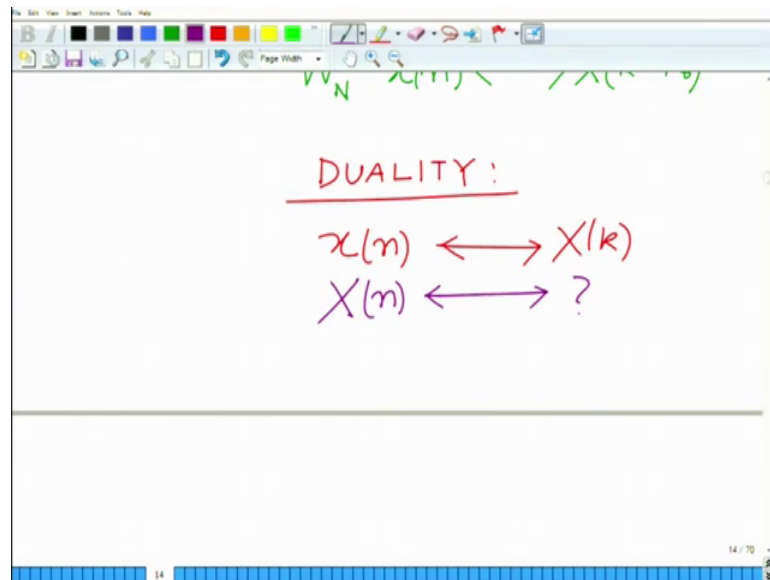
So, let us say X of k . Then you can show that x conjugate of n , it is not very difficult to see x conjugate of n has the DFT that is X conjugate of minus k , but remember the minus k has to be, has to be interpreted as modulo N all right. So, that is X conjugate of minus k modulo N in the next property. So, again this is a very simple property which you can verify yourself.

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There is a frequency shift property, the frequency shift property, shift in time results in modulation and frequency naturally one can expect, as expected modulation in time results in a shift in frequency. So, x_n as the DFT X of k then W_N^k naught of n x_n as the DFT $X_{k \text{ minus } k \text{ naught } 0k}$. Then you have the duality all right.

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So, x_n has the DFT X of k . Now what can we say about the sequence, remember duality is the time domain becomes frequency domain and what happens to the DFT of capital X of n then a small x of n has the DFT capital coefficients given by capital X of k . What can we say about the DFT coefficients, what can we say about the DFT of the time domain coefficients given by capital X of n . And that is also fairly easy to see we have you just write it as capital small x of n .

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The image shows a whiteboard with a toolbar at the top. The main content is handwritten in purple and yellow. The first equation is $x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-kn}$. Below it, the text "Interchange n, k" is written in yellow. The second equation is $x(k) = \frac{1}{N} \sum_{n=0}^{N-1} X(n) W_N^{-nk}$. The page number "15 / 70" is visible in the bottom right corner.

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) W_N^{-kn}$$

Interchange n, k

$$x(k) = \frac{1}{N} \sum_{n=0}^{N-1} X(n) W_N^{-nk}$$

Remember you can write it is in the inverse DFT as 1 over N summation k equal to 0 to N minus 1 X k W N of minus k n, which is.

Now, we change. So, now, if you look at x of, now if change interchange n and k replace n by k, interchange the rules of n and k and then what you will have is, I can simply write x of k equals 1 over N summation k equal to 0 to N minus 1, or in fact, we are interchanging xn n and k. So, this is n equal to 0 capital N minus 1 X of n W N and this remains minus nk, which is the same as minus k n.

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The image shows a whiteboard with a toolbar at the top. The main content is handwritten in yellow and blue. The first equation is $x(k) = \frac{1}{N} \sum_{n=0}^{N-1} X(n) W_N^{-nk}$. The second equation is $x(-k) = \frac{1}{N} \sum_{n=0}^{N-1} X(n) W_N^{nk}$. The third equation is $\Rightarrow N x(-k) = \sum_{n=0}^{N-1} X(n) W_N^{nk}$, with "DFT of Xn" written below the summation. The page number "15 / 70" is visible in the bottom right corner.

$$x(k) = \frac{1}{N} \sum_{n=0}^{N-1} X(n) W_N^{-nk}$$
$$x(-k) = \frac{1}{N} \sum_{n=0}^{N-1} X(n) W_N^{nk}$$
$$\Rightarrow N x(-k) = \sum_{n=0}^{N-1} X(n) W_N^{nk}$$

DFT of Xn

And now what I can do is if I consider x of minus k that becomes 1 over, that becomes 1 over N summation n equal to 0 to N minus 1 $X_n W_N$ raise to $n k$, whether replacing k by, replacing it by minus k ok, W_N by star n .

Of course the minus k is interpolated modulo N and the final operation is to get the N to the other side and that implies $N x$ of minus k is basically summation n equal to 0 to N minus 1 capital $X_n W_N$ to the power of $n k$, which is basically DFT of you can see X_n .

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The image shows a whiteboard with handwritten mathematical equations. The top equation is:

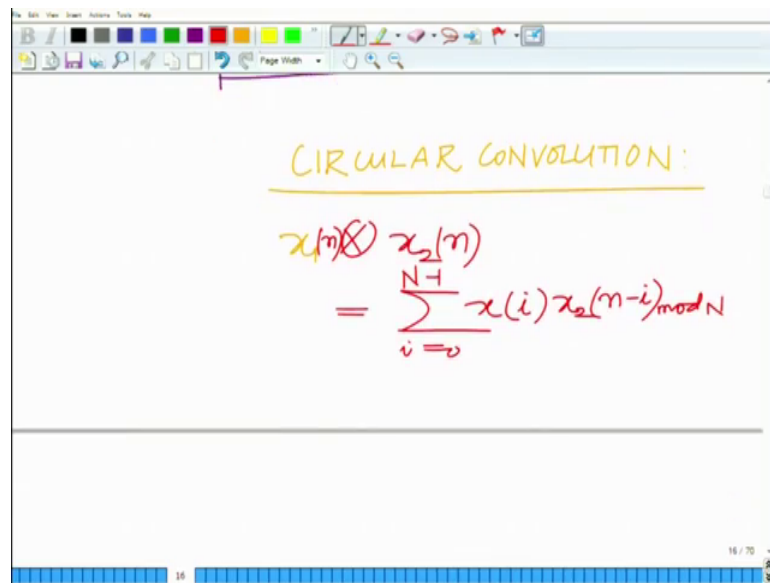
$$\Rightarrow N x(-k) = \sum_{n=0}^{N-1} X(n) W_N^{nk}$$

Below this equation, it is noted that the sum is the DFT of $X(n)$. The bottom part of the whiteboard shows a boxed relationship:

$$X(n) \longleftrightarrow N x(-k) \text{ mod } N$$

So, X_n , the capital X_n sequence you can see has the DFT coefficients which are $N x$ of minus k . So, And of course, the minus k is remember everything is always interpreted modulo N . So, a capital X of n has the DFT discrete time Fourier transform coefficients given by capital $N x$ of minus k all right. And then the next property is also fairly important, which is a circular convolution

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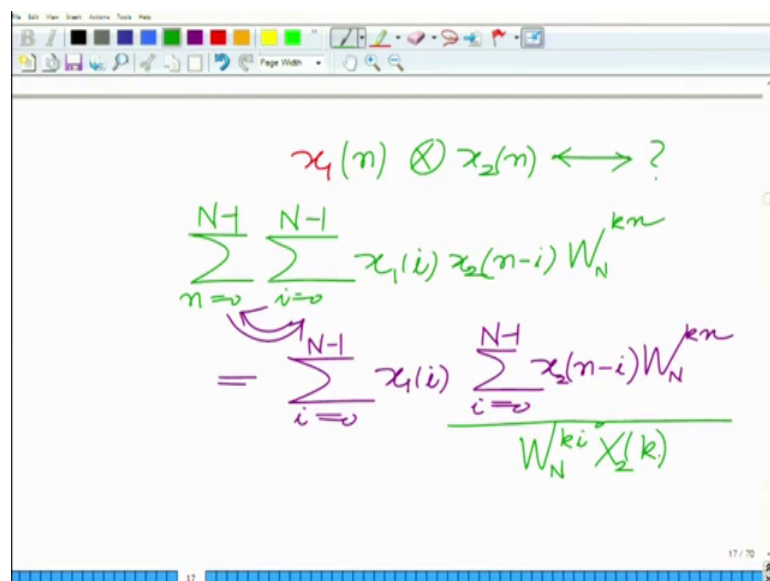
The slide shows a whiteboard with a toolbar at the top. The title "CIRCULAR CONVOLUTION:" is written in yellow and underlined. Below it, the equation $x_1(n) \otimes x_2(n) = \sum_{i=0}^{N-1} x_1(i) x_2((n-i) \bmod N)$ is written in red ink.

$$\text{CIRCULAR CONVOLUTION:}$$
$$x_1(n) \otimes x_2(n) = \sum_{i=0}^{N-1} x_1(i) x_2((n-i) \bmod N)$$

Which is basically x_1 circularly convolved with x_2 or x_1 circularly convolved with x_2 of n . This is basically equal to summation i equal to 0 to N minus 1 $x_1(i) x_2(n-i)$.

But the $n-i$ has to be interpreted modulo N uh, which is basically x_2 for summation n equal to, so basically which is summation $x_1(i) x_2(n-i \bmod N)$. So, summation $x_1(i) x_2(n-i \bmod N)$. And now if you look at the DFT of this you want to ask the question.

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The slide shows a whiteboard with a toolbar at the top. The title $x_1(n) \otimes x_2(n) \leftrightarrow ?$ is written in green. Below it, the equation $\sum_{n=0}^{N-1} \sum_{i=0}^{N-1} x_1(i) x_2((n-i) \bmod N) W_N^{kn} = \sum_{i=0}^{N-1} x_1(i) \sum_{l=0}^{N-1} x_2(l) W_N^{ki} = W_N^{ki} X_2(k)$ is written in green and purple ink. A purple arrow points from the $(n-i) \bmod N$ term in the first sum to the l term in the second sum.

$$x_1(n) \otimes x_2(n) \leftrightarrow ?$$
$$\sum_{n=0}^{N-1} \sum_{i=0}^{N-1} x_1(i) x_2((n-i) \bmod N) W_N^{kn}$$
$$= \sum_{i=0}^{N-1} x_1(i) \sum_{l=0}^{N-1} x_2(l) W_N^{ki}$$
$$= W_N^{ki} X_2(k)$$

The question that we want to ask is x_1 of n circularly convolved with x_2 of n . What is the DFT of this quantity, and the DFT of this quantity can be found as follows, you have

summation i equal to 0 to n minus 1, remember this is $x_1(i)$, $x_1(i) x_2(n-i) W_N^{ki}$ summation n equal to 0 to N minus 1.

That is basically your or the DFT create, the DFT coefficient of the convolution circular convolution of $x_1(n)$, spherically convolved with $x_2(n)$ all right. And now again as usual we interchange the order of summation. So, that gives us summation i equal to 0 to N minus 1 summation n equal to 0 to capital N minus 1.

We have x_1 , now the term with respect to i that comes outside. So, that is $x_1(i)$ summation i equal to 0 to N minus 1 $x_2(n-i) W_N^{ki}$, and remember this is the shifted sequence, DFT of the shifted sequence of the x_2 of $n-i$. So, this is simply W_N^{ki} times $x_2(k)$.

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$$= \sum_{i=0}^{N-1} x_1(i) W_N^{ki} X_2(k)$$

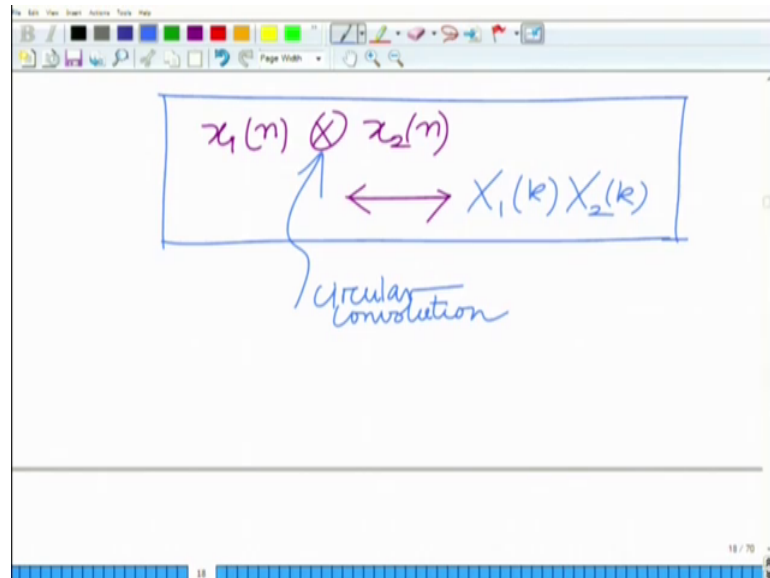
$$= X_1(k) X_2(k)$$

So, now we have this basically reduces to summation i equal to 0 to N minus 1 $x_1(i)$ into W_N^{ki} into $X_2(k)$ and $X_2(k)$ is constant depending only on k and you can see this is nothing, but $X_1(k)$. So, that reduces to $X_1(k)$ times $X_2(k)$.

And this is something that is very important. Circular convolution of two finite length sequences right of equal length capital length, when you circularly convolved them the DFT domain they respect, the respective DFT coefficients get multiplied and this is a very important property, it is used by practical wireless communication system.

In fact, most of the four, all of the four g phones that you use are based on the LTI standard which uses OFDM orthogonal frequency division multiplication multiplexing, which is based on this phenomena correct.

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So, we have to finite time domain sequences x_1 of length n circularly convolved with x_2 of length n and that basically has the DFT coefficients $X_1(k)$ into multiplication in the DFT domain ok. So, these are the coefficients $X_1(k)$ into $X_2(k)$ all right; so circular converter.

So, this is, remember this is your circular convolution, just so that you are not confused; one has to use the appropriate sense of convolution correct, for periodic sequence for periodic signals or finite length signals its always circular convolution, for infinite signals its always the regular in the linear convolution all right, and on the other hand when you have multiplication in the time domain again as a dual property multiplication in time, when you have multiplication in time.

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MULTIPLICATION:

$$x_1(n) \cdot x_2(n) \longleftrightarrow \frac{1}{N} \sum_{k=0}^{N-1} X_1(k) \otimes X_2(k)$$

19 / 70

So, $x_1(n)$ into $x_2(n)$ that has the DFT that is 1 over N and you can show this $X_1(k)$ circularly convolved with $X_2(k)$, because remember capital X_1 capital X_2 are also finite length, DFT sequences so its circular convolution between $X_1(k)$ and $X_2(k)$ and divided by N ; that is the rule for multiplication of 2 time domain signals. And now we can derive the Parseval's relation from this.

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PARSEVAL'S RELATION:

$$x_2(n) = x_1^*(n)$$
$$|x_1(n)|^2 \longleftrightarrow \frac{1}{N} \sum_{k=0}^{N-1} X_1(k) \otimes X_1^*(-k)$$
$$= \frac{1}{N} \sum_{i=0}^{N-1} X_1(i) X_1^*(i-k)$$

19 / 70

The Parseval's relation, we can derive the Parseval's relation, Parseval's relation we can set $x_2[n]$ equals $x_1^*[n]$. So, if you multiply $x_1[n]$ with $x_1^*[n]$, then we have magnitude $x_1[n]$ square that has the Fourier transform 1 over N $X_1[k]$.

And the Fourier transform of $x_1^*[n]$; that is $X_1^*[k]$ which is if you look at this; that is going to be summation 1 over i equal to 0 to N minus 1 $X_1[i]$ $X_1^*[i - k]$; of course, the $i - k$ is modulo of N .

So, this is basically the Fourier transform of magnitude X_1 square, which means if you look at summation of that k -th DFT coefficient, summation of n equal to 0 to N minus 1 ; that is if you look at the k -th DFT coefficient of this magnitude $x_1[n]$ square W_N to the k n .

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The image shows a whiteboard with the following handwritten content:

$$\sum_{n=0}^{N-1} |x_1[n]|^2 W_N^{kn}$$

Below this, a horizontal line is drawn. To the left of the line, an arrow points to the expression with the text: k^{th} DFT coefficient of $|x_1[n]|^2$.

$$= \frac{1}{N} \sum_{i=0}^{N-1} X_1[i] X_1^*[i-k]$$

To the right of the second equation, an arrow points to the text: set $k=0$.

This is the k -th DFT coefficient of magnitude $x_1[n]$ square. Remember this is the k -th DFT coefficient. And this is equal to remember 1 over N summation i equal to 0 to N minus 1 $X_1[i]$ $X_1^*[i - k]$. Now in this both sides what we do now is you set k equal to 0 . When you set k equal to 0 what I have is this reduces to W_N raise to 0 which is 1 .

So, on the left hand side you have summation n equal to 0 to capital N minus 1 magnitude $x_1[n]$ square. On the right hand side you have 1 over N summation i equal

to 0 to N minus 1 X_i into X_i conjugate i minus k , but k is 0. So, X_i into X_i conjugate i , which is magnitude X_i square summation i equal to 0 to N minus 1 divided by N. So, that is your Parseval's relation.

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$$\sum_{n=0}^{N-1} |x(n)|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |X(k)|^2$$

PARSEVAL'S RELATION FOR DFT

So, set k equal to 0 that gives you summation, that gives us something very interesting, just a very handy relation that gives you summation n equal to 0 to capital N minus 1 magnitude x_n square its 1 over N summation i equal to 0 to N minus 1 magnitude X_i square and this is the Parseval's relation for, this is the Parseval's relation for DFT.

This is a Parseval's relation for DFT discrete Fourier transform all right, and that basically completes our discussion for the DFT that is the discrete Fourier transform. All right. So, what we can do now is, we can start looking at some example problems for the Fourier analysis of discrete time

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EXAMPLE PROBLEMS FOR
FOURIER ANALYSIS OF DISCRETE
TIME SIGNALS:

#1. Consider periodic signal.
 $N_0 = \text{even}$

$$x(n) = \begin{cases} 1 & 0 \leq n \leq \frac{N_0}{2} - 1 \\ 0 & \frac{N_0}{2} \leq n \leq N_0 - 1 \end{cases}$$

So, we have example problems. In fact, several example problems for Fourier analysis of discrete time signals. The Fourier analysis of discrete time signals, Let us start with the first problems, first similar to our discussion here, we are going to start with the discrete Fourier series. First we are going to the discrete time Fourier transform and then ultimately the discrete Fourier transform, just the way we have covered this in the various modules.

Let us look at our first problem that is considered a periodic signal period N_0 equal to N_0 , which is even period N_0 is even. So, $x(n)$ equals 1, in a single period is given as $x(n)$ equal to 1 for $0 \leq n \leq \frac{N_0}{2} - 1$, and it is 0 for $\frac{N_0}{2} \leq n \leq N_0 - 1$.

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$$x(n) = \begin{cases} 1 & 0 \leq n \leq N_0 - 1 \\ 0 & N_0 \leq n \leq N_0 - 1 \end{cases}$$

DESCRIPTION FOR a single period.
 DFT of $x(n) = ?$

$$\omega_0 = \frac{2\pi}{N_0}$$

$$C_k = \frac{1}{N} \sum_{n=0}^{N-1} x(n) e^{-j\omega_0 kn}$$

So, this is basically description of it in a single period and then it repeats ok. This is for 1 period a naught and then again it repeats ok, description for a single period, this is the description for a single period. Now what is that discrete DF, the DFS, what is that discrete Fourier series coefficients of x_n . Now we know.

That N is the period. So, the fundamental frequency ω_0 can be defined as 2π by N and the DFS coefficient is C_k equals summation 1 over. Remember the DFS coefficients are given as 1 over N summation n equal to 0 to capital N minus 1 to $x_n e$ raise to minus j ω_0 kn equals 1 over N summation n equal to 0 to.

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$$= \frac{1}{N_0} \sum_{n=0}^{N_0-1} e^{-j\omega_0 kn}$$

$$= \frac{1}{N_0} \frac{1 - e^{-j\omega_0 k N_0}}{1 - e^{-j\omega_0 k}}$$

$$C_k = \frac{1}{N_0} \frac{1 - e^{-j\omega_0 k N_0}}{1 - e^{-j\omega_0 k}}$$

Well, now it is only 1 from 0 to N naught by 2 minus 1. So, N naught by 2 minus 1 xn is 1 e raise to minus j omega naught kn, which is basically 1 over N naught 1 minus e raise to minus j omega naught k e raise to minus j omega naught k into N naught by 2 into over 1 over 1 minus e raise to minus j omega naught k.

And now here in the numerator you can see omega naught equals 2 pi by N naught. So, omega naught N naught over 2 is basically simply pi. So, in the numerator I have basically, this is your Ck, left Ck right as 1 over the right hand side is 1 over N naught times 1 minus e raise to minus j omega naught N naught over 2 is pi 1 minus 1 1 minus e raise to minus j pi k 1 minus e raise to minus j omega naught k.

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$$= \frac{1}{N_0} \cdot \frac{e^{j\pi k/2}}{e^{-j\pi k/2}} \cdot \frac{e^{j\pi k/2} - e^{-j\pi k/2}}{e^{-j\pi k/2} - e^{j\pi k/2}}$$

$$C_k = \frac{1}{N_0} \cdot e^{j\frac{k}{2}(\pi - \omega_0)} \cdot \frac{\sin(k\pi/2)}{\sin(\omega_0 k/2)}$$

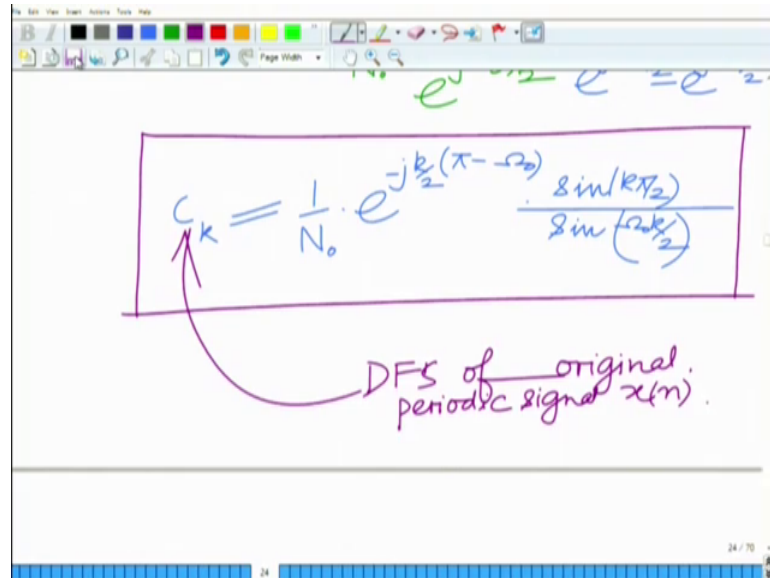
Which is basically 1 over N naught e raise to minus j pi k divided by 2 e raise to minus j omega naught k divided by 2 e raise to minus j pi k over 2 minus e raise to.

I am sorry e raise 2 j pi k over 2 minus e raise to minus j pi k over 2, denominator i raised to minus j omega naught k over 2 or 1 e raise to 0 omega naught 0 2 minus e raise to minus j omega naught k over 2.

Which is basically now you can simplified the this as 1 over N naught e raise to minus j k over 2 times pi minus omega naught times sin k pi over 2 divided by sin omega naught k over; that is our expression for.

So, this is your expression for C_k ; that is the DFS discrete Fourier series coefficients original periodic signal x_n oh all right.

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The image shows a handwritten equation for the DFS coefficient C_k enclosed in a purple box. The equation is $C_k = \frac{1}{N_0} \cdot e^{-j\frac{k}{2}(\pi - \Omega_0)} \frac{\sin(kT_0/2)}{\sin(\frac{\Omega_0 k}{2})}$. Above the box, the expression $e^{-j\frac{k}{2}(\pi - \Omega_0)} = e^{-j\frac{k}{2}\pi} e^{j\frac{k}{2}\Omega_0}$ is written in green. Below the box, a purple arrow points from the text "DFS of original periodic signal $x(n)$ " to the C_k term. The background is a whiteboard with a toolbar at the top and a blue bar at the bottom.

So, we have basically what you can see or basically what we have done in this module, is we have completed or discussion of the DFT, the various properties of the DFT as, such as the duality the convolution, the circular convolution and also the Parseval's relation followed by an example for the discrete Fourier series.

So, we will stop here and look at other examples in the subsequent modules.

Thank you very much.