

**Conduction and Radiation**  
**Prof. C. Balaji**  
**Department of Mechanical Engineering**  
**Indian Institute of Technology, Madras**

**Lecture No. # 43**  
**Numerical Methods in Conduction**

So, good morning in yesterday's class we looked at the method of separation of variables it is an important analysis technique for analytically solving partial differential equations. The idea is to split the p d into two o d's introducing a constant called lambda square and the other inherent assumption is solving the o d is much easier comparing to solving the p d. Then, we got the general solution then we had this boundary conditions, using the boundary conditions we got the co we evaluated the constants a, b, c, d. When we were trying to evaluate the constants several new things entered to picture, like there is a lambda and the solution is satisfied and lambda is equal to  $1 \pi$  by  $l$   $2 \pi$  by  $l$  and so on. These lambdas's are called Eigen values of the problem. And now and then finally, you got the solution in terms of  $\sin n \pi x$  by  $l$  and  $\sin x n \pi y$  by  $l$ , where  $\sin x n \pi$  in yesterday's call was basically because the temperature was exponentially decaying in the y direction.

As far as the  $\sin n \pi x$  by  $l$  is concerned it is basically, a repeating of periodic boundary condition. In fact it is, because of this  $\sin n$  by  $n \pi x$  by  $l$  that you are able to use this orthogonal invoke the orthogonality property. That is why this periodic boundary condition this 0 0 boundary condition is very important, thing is towards the end of the class we also saw that if you have a problem where more than one boundary condition is 0. Then you have to use the super position principle have  $1 0 0$  plus  $0 0 1 0 0$  you split the problem into several sub problem and use the method of separation of variables we individually find the solution, and then get a composite solution which is the additive sum of all the solutions.

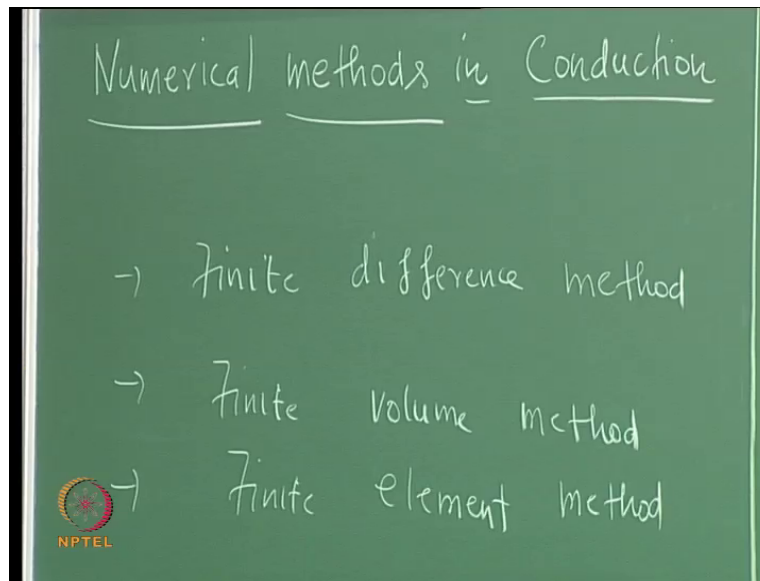
But, now you can see that even for a two dimensional problem two dimensional steady steady state steady conduction in a slab with constant properties a getting an analytical solution was quiet involved. And then finally, your finally, you have to evaluate a series it is not that you have got an expression like e to the power of minus m I so e to the power. So, you have to evaluate a series, and then the accuracy of a solution depends on the number of terms you retain in the series. So, it is not very very convenient though you we we will claim that it is an

analytical solution and so on. You will soon realize that if you evaluate the heat flux using the series, you require many many terms. It is not so easy to get the heat flux with, because again the temperature has to be differentiated and then the  $k \frac{dT}{dx}$  term will contain so many terms in the series.

And for certain other types of boundary condition for example, you are varying one you have  $q$  which is a function of  $x$ , and then you have a convective boundary condition on one side you have two other temperatures on the other two sides and so on. And then it becomes extremely formidable to solve these problems. And the moment you say thermal conductivity is not a constant and so on it is very difficult to solve these problem. Therefore, numerical techniques were developed, and the growth of the numerical techniques was also accelerated by the development of computers. And then this availability of programming languages likes FORTRAN and so on, which enabled the engineers to tackle this problem. Of course, now a days you have got sweets like mat lab which can solve many of these problems, and you have got libraries which solve o d's and p d's and Fortran and all that and then there are commercial software like comsolve is very powerful based on finite element, and fluent which based on finite volume which is based on finite volume which can solve all these problems with considerable ease.

So, the engine behind all these software is some numerical technique. So, these numerical techniques broadly the numerical techniques which are used in conduction in heat transfer basically are either the finite difference method.

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As far as heat transfer and fluid flow are concerned the first two are very very popular. The third is not so popular, but still there are few there are some practitioners would like to use the finite element method for both for both fluid flow and heat transfer. Finite volume most of the commercial software our fluent works on the finite volume method. So, essentially what is the basic difference between finite difference finite volume and finite element method? The finite difference is different from the other two because the finite difference method is based on the Taylor's series approximation of a function, from that you get the first derivative second derivative and then you find out what is the value of the nodal value of temperature or velocity or whatever variable under question, as a function of it is neighboring nodes. And then once you set up a system finally, you set up a system of linear equations which is solved by matrix inversion or gauss seidel iteration and precede that is how a finite difference work.

A finite volume method is basically the control volume technique it is also called a control volume method. Where you a taking control volume and find out what are the fluxes which are entering, what are the fluxes which are leaving, any other heat generation which is in the medium or for any heat storage or heat depletion that is enthalpy increase or enthalpy decrease in the medium. So, it is basically like a you are looking at the law of conservation of energy as far as conduction problem is concerned and writing down all the terms and finding out an expression for the value of the temperature at a particular node. So, there is no Taylor's series expansion in a finite volume method.

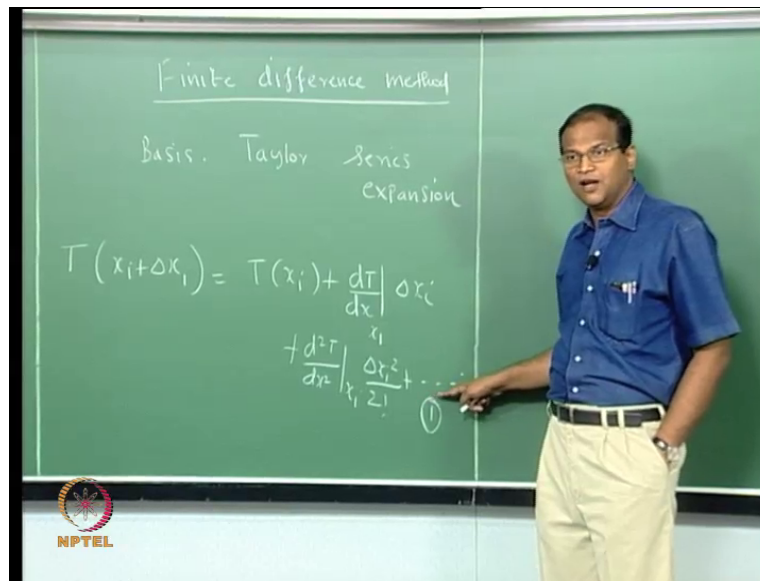
And in a finite in the finite difference method we finally, evaluate finally, we evaluate the derivative at a point, but what is this point we have to be very careful it has a point of area and volume. If the point does not have any area and any volume then you cannot define stress or something if for example, stress cannot be defined at a point and so on. So, finally, we have to have that in mind. In the limit  $\Delta x \Delta y$  tending to 0 that means  $\Delta x$  cannot become 0 then you are in trouble. So, finite volume method is quite natural, because it is based on finite control volume. You do not reduce it to inimitably small size and make it a point. So, naturally the finite volume method can handle certain type of things.

So, the finite element method is actually a cousin of the finite volume method where also the governing equation is integrated once. When the governing equation is integrated once conditions like Nyman condition like  $dT$  by  $dx$  is a constant of for example,  $dT$  by  $dx$  is equal to 0, is very frequently encountered in conduction problems. If you have  $n$  surfaces in a problem one of one or two the surfaces may be insulated.  $DT$  by  $dx$  is equal to 0 is very easily handled by a finite element method, because the governing equation is integrated once, and then you start finding out the equation for a particular element then you find the equations for all the elements assemble all these elements you get a system of linear equation, beyond a certain state it becomes again like a finite volume method. So, you integrate the governing equation once.

But, in finite difference method what you do is you take the governing equation and start discretizing,  $i$  plus 1  $j$   $i$  minus 1  $j$  we do not do that in a finite element method, but anyway that  $I$  1  $i$  minus 1  $j$   $i$   $j$  everything will come in the end. Instead of using the  $i$   $j$  notation in finite volume and finite in finite volume method particularly we use east west north south here we use  $i$  minus 1  $j$   $i$  plus 1  $j$   $i$   $j$  plus 1  $i$   $j$  minus 1.

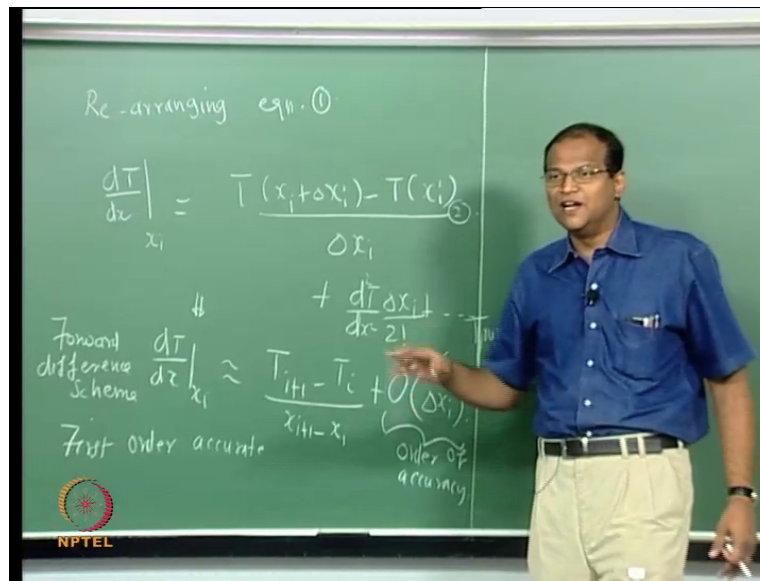
Now in this lectures, we will just we will look at the basic principles underlying the finite difference as well as the finite volume method. The finite element method is beyond the scope of this course.

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So, finite difference what is the basis, the basis is the Taylor's series expansion. So, the basis is the Taylor's series expansion now, let  $T$  be let us say that you want to evaluate  $T$   $T$   $x$   $i$  plus  $T$  is the temperature. I am instead of using  $y$  I want to use  $T$ , because  $T$  is frequently used in conduction,  $x$  is the spatial coordinate so  $T$  is a function of  $x$  it could be a function of  $x$   $y$   $z$  whatever. Now for simplicity we say  $T$  is a function of  $x$   $i$  using Taylor's series we want to expand it as  $T$  is a function of  $x$   $i$ , plus  $dT$  by  $dx$  at  $x$   $i$  into  $\Delta x$   $i$  plus plus higher order terms. The higher order terms will become negligibly small when  $\Delta x$  is  $\Delta x$   $i$  is properly chosen, because it is going to be raised to the highest powers of  $\Delta x$ . so, if  $\Delta x$  is point naught one, this will be point naught one, this will be point naught one squared. This will be point naught one cube and remember at the bottom also it is two factorial three factorial everything is increasing. So, higher order terms can be neglected even though  $d^2 T$  by  $dx$  squared maybe significant and higher order derivatives may be significant.

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Now this equation can be rearranged for dT by dx. So, rearranging equation 1. So, this delta x i is small compared to x i small expansion, that is very very important it is x i plus epsilon if x i is 100 delta x should not be 75. If x is 100 delta x can be 1.5 0.1 like that it is called as small expansion. Rearranging equation one we have yeah plus so the key point here is you are anyway neglecting higher order term, if you say that dT by dx is equal to approximately this, then this will be the order of the error associated with the truncage. So, the order of the error is directly proportional to delta x I, since it is proportional to delta x i to the power of one it is called a first order accurate scheme.

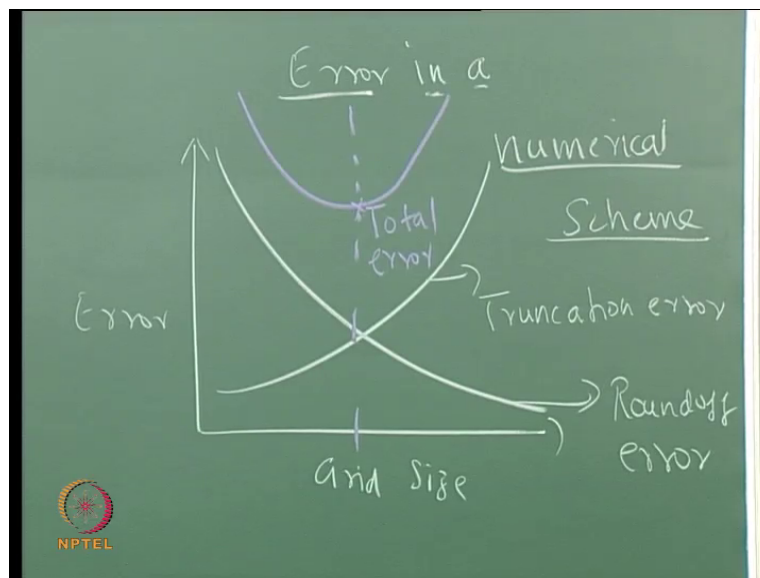
Therefore, now if you write like this dt by da is not 0 it is order of that is the leading order, do not ask me sir what about other terms other terms are there, but other terms are much smaller compared to this fellow. So, the maximum culprit maximum error will be caused by delta x next term is delta x square by 2 factorial. And anyway delta x itself is small. So, this is the gives you the order of accuracy this is called the forward difference. It is first order accurate, because delta x i please be careful, and we are neglecting only upto the second order, we are taking upto second order. We are taking up to second order but, when you get the dT by dx we are dividing by delta x therefore, d square T by dx square gets multiplied by 2 factorial it gets so, one delta x i gets cancelled are you getting the point so it first order accurate.

So, if you want to use this first order accurate scheme for solving a problem, you have to use more number of grids, because there will be a severe what is called the this error is called the

truncation error for the numerical scheme. Now this is for one node in a problem in a two dimensional problem. You may have  $\Delta t$  by  $\Delta x$  by  $\Delta t$  by  $\Delta y$   $\Delta x^2 \Delta y^2$  by  $\Delta x \Delta y \Delta x^2 \Delta y^2$ , then each of this  $\Delta t$  by  $\Delta x$  is represented as a finite difference, instead of a differential you are representing it a finite difference when you assemble all this you will get a set of simultaneous equations for  $T$  of  $i, j$ .  $i$  can vary from 1 to 20  $j$  can vary from 1 to 20 depending on your grid. Then when you are doing these operations the computer can store only a finite number of digits. Repeatedly when you are doing gauss seidel method then there will be a loss, there will be a loss of information because the computer will truncate after a certain number of decimal that is called round off error.

So, the round of error will increase the round of error will increase if you are using more number of grids. When you are using more number of grid  $\Delta x$   $i$  will come down therefore, the truncation error will decrease. So, there is an optimum grid size at which the sum of these two errors is the total error will be a minimum.

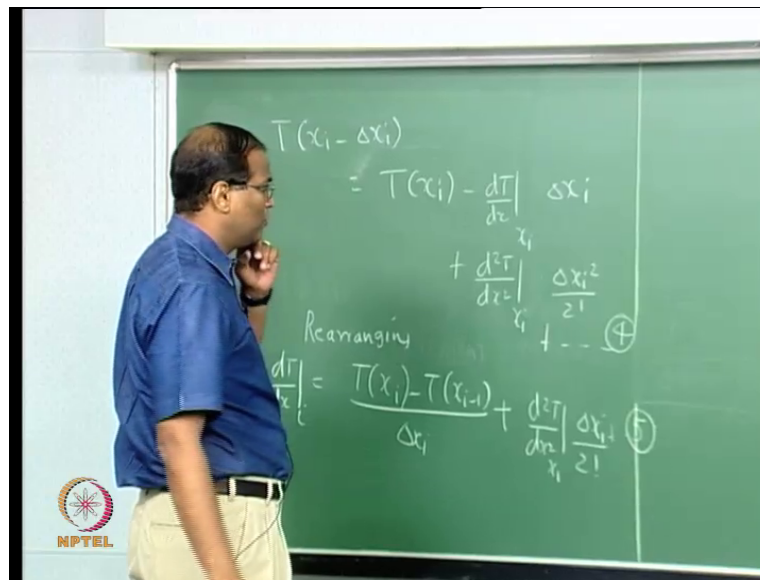
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So, if you plot the error what are these two, a grid size in grid size is size of the grid  $\Delta x$  increases, because number of number of computer operations are reduced, because only you are solving on a 5 by 5. You are solving on a 50 by 50 truncation error increases. What happened tell me is grid size this is not number of grid grid size is the size of  $\Delta x$ . If the size of  $\Delta x$  increases truncation error will increase, there is no doubt about it. I am not saying number of grid points here, it is one by number of grid point it is inversely

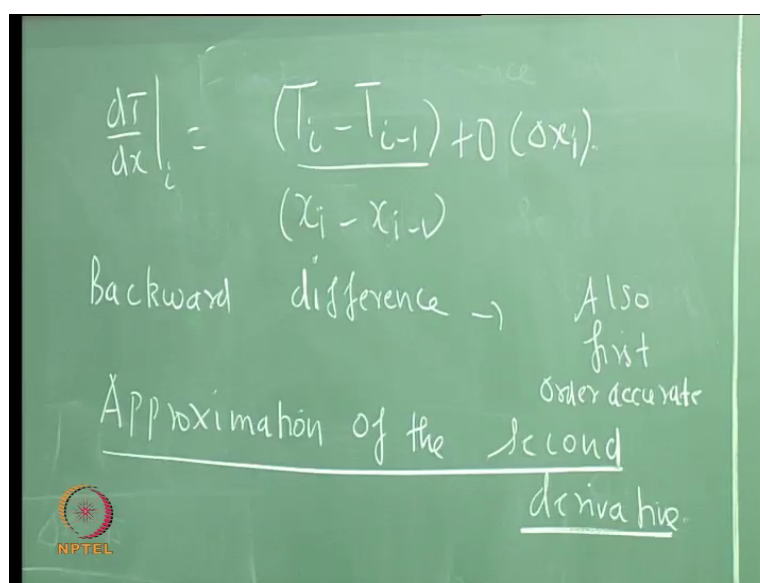
proportional to the number of grid. You can plot the same thing as number of grid points then you have to make it ultra round off and truncate. So, this is about forward difference it is also possible to use a backward difference.

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Now what will be the backward difference, T of next step into minus and plus will alternate plus this what is this now this should be 3 rearranging plus.

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So, I can just write it as the backward difference is also first order accurate. So, it is called backward difference please note we are trying to get expressions for  $dT$  by  $dx$   $d^2T$  by  $dx^2$  and so on, because we are frequently encountering partial differential equation. So, we want expressions for  $\frac{dT}{dt}$  by  $\frac{dx}{dt}$   $\frac{dT}{dy}$  more importantly we want expressions for  $\frac{d^2T}{dx^2}$  by  $\frac{d^2T}{dy^2}$  square into  $\frac{d^2T}{dx^2}$  by  $\frac{d^2T}{dy^2}$  square, because only the mixed derivative first derivative second derivative. First derivative in time second derivative in space, with that all combinations will be there unsteady heat unsteady heat conduction steady state heat conduction everything will be covered. Of course, I am deriving all these only for the Cartesian coordinate, you will be little more involved but, using the same procedure you can derive the finite difference from approximations for the first derivative for both the cylindrical as well as the spherical coordinates.

Now approximation of the second derivative that is very very important, because if you want to solve the Laplace equation. The  $\frac{d^2T}{dx^2}$  and  $\frac{d^2T}{dy^2}$  make an appearance. So, now we will have to look at the approximation of the second derivative. So, please remember that even the backward difference is also first order accurate approximation.

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$$T(x_{i+1}) = T(x_i) + \left. \frac{dT}{dx} \right|_{x_i} \Delta x_i + \left. \frac{d^2T}{dx^2} \right|_{x_i} \frac{\Delta x_i^2}{2!} + \left. \frac{d^3T}{dx^3} \right|_{x_i} \frac{\Delta x_i^3}{3!} + \dots$$

$$T(x_{i-1}) = T(x_i) - \left. \frac{dT}{dx} \right|_{x_i} \Delta x_i + \left. \frac{d^2T}{dx^2} \right|_{x_i} \frac{\Delta x_i^2}{2!} - \left. \frac{d^3T}{dx^3} \right|_{x_i} \frac{\Delta x_i^3}{3!} + \dots$$

Now  $T$  of  $x$  can I write like this itself or  $T$  of  $x$  plus 1 can I write that  $x$  plus 1 or  $x$  of we will keep as  $x$  of  $x$  of  $x$  plus 1, equal to this is based on forward differences.  $T$  of so I am writing the Taylor series expansion for  $T$  of  $x$  plus 1. And I am also writing the Taylor's

series expansion of  $T(x_{i+1}) - T(x_{i-1})$ . No in fact, now there are several ways of doing what we want to do first, instead of taking the second derivative what I want to do first is, with the same thing several things can be done, we can call the title as central difference before going to this.

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Handwritten on a green chalkboard:

$$T(x_{i+1}) - T(x_{i-1}) = 2 \frac{dT}{dx} \Big|_i \Delta x_i + 2 \frac{d^3T}{dx^3} \Big|_i \frac{\Delta x_i^3}{3!} + \dots$$

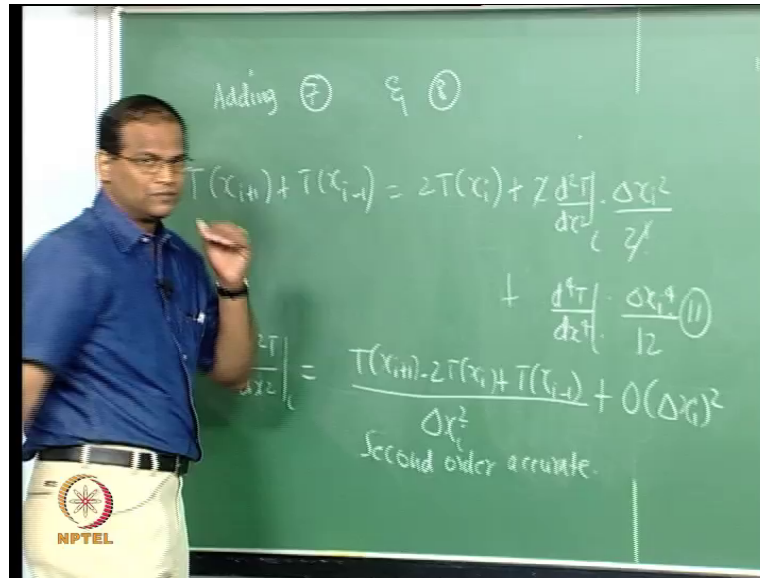
$$\frac{dT}{dx} \Big|_i = \frac{T(x_{i+1}) - T(x_{i-1})}{2\Delta x_i} - \frac{d^3T}{dx^3} \Big|_i \frac{\Delta x_i^2}{2!} + \dots$$

$$\frac{dT}{dx} \Big|_i \approx \frac{T_{i+1} - T_{i-1}}{2\Delta x_i} + O(\Delta x_i^2)$$

Annotations on the board include: "Second order accurate" and "central difference approximation". An NPTEL logo is visible in the bottom left corner.

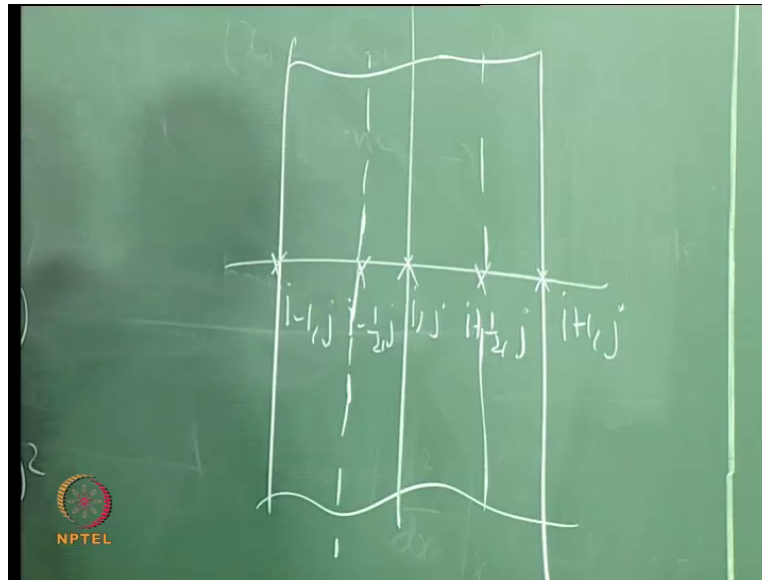
Now subtracting 8 from 7 no no I want to find first derivative. Subtracting subtracting subtracting 8 from 7 we get  $T(x_{i+1}) - T(x_{i-1})$  get cancelled  $2 \frac{dT}{dx}$  correct, second derivative gone third derivative is there plus. Now therefore, minus minus therefore,  $\frac{dT}{dx}$  at  $i$  can be written as  $\frac{T(x_{i+1}) - T(x_{i-1})}{2\Delta x_i}$  but, the beauty now is the order of accuracy is  $\Delta x_i^2$ . So, if then if you are evaluating the first derivative at  $x_i$  you are taking a node ahead  $i+1$  minus  $i-1$  divided by  $2\Delta x_i$ , but you get the second order accurate. So, if you use a central difference only thing is the value of the node at that the value of the variable at that node is not considered? You want to get  $\frac{dT}{dx}$  at  $T(x_i)$  but,  $T(x_i)$  is not used in the expression no problem. So, that is the therefore, this is this is the central difference approximation this is the central difference approximation for first derivative it is clearly seen that it is second order accurate, because it is  $\Delta x_i^2$  square. This is the the second order accurate.

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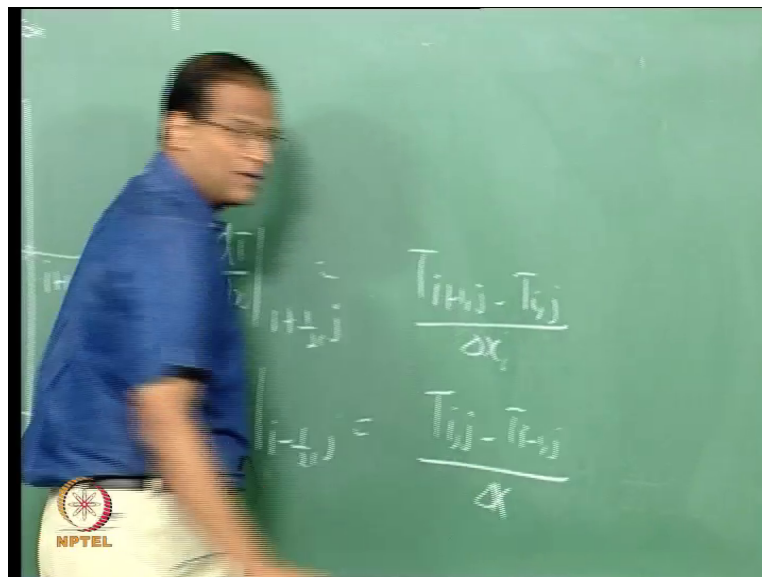
If you add what happens shall we do that, if you add adding 7 and 8 adding 7 and 8 we have t of is equal to dT by dx is gone. And then 2 into then plus now you get get an expression for d square dx by d square. Therefore, I think we need a number did I put a number 9 I will call this 10 I will call this 11. Therefore is equal to divided by 2 by delta x square, but 2 is 2 gets cancelled no no then there is a problem is it correct, no the no no no there is a 2 here. What I have done is T x i plus 1 plus T x i minus 1 is there, minus 2 is so the two remains in the d square dx by d square. 2 factorial 2 factorial will get cancelled this will be 24. So, this will be delta x squared plus the square of square gets cancelled. It is still second order accurate, this is central difference not forward difference this is called the central difference. i plus 1 i minus 1 everything is there what is the forward here, you have got forward and backward it is its one elliptic equation so elliptic. So, this is basically second order accurate.

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So, now we have got an expression for  $d^2x$  by  $dx^2$ , it is called central difference. This can also be alternative derivation  $i + j$   $i - 1$   $j$ . so, I can take an intermediate node  $i + \text{half } j$  I can take here  $i - \text{half } j$ .

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So, I can get so  $dT$  by  $dx$  at half  $j$  is equal to  $T$  of  $i + 1$   $j$  minus  $T$  of  $i$   $j$  divided by  $\Delta x$ , I can get  $dT$  minus  $dx$  at  $i - \text{half } j$ , that is  $T$  at  $i$   $j$  minus  $T$  of  $i - 1$   $j$  so this is. Now I

will say d square T by dx squared here is dT by dT by dx at i plus 1 i plus half minus dT by dx at i minus half divided by delta x.

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$$\frac{d^2\bar{T}}{dx^2}\bigg|_{i,j} = \frac{\frac{d\bar{T}}{dx}\bigg|_{i+\frac{1}{2},j} - \frac{d\bar{T}}{dx}\bigg|_{i-\frac{1}{2},j}}{\Delta x}$$

There is already a delta x here delta x and delta x will give you delta x square and you will again get T of i plus 1 j minus T of i minus 1 j T of i plus 1 j plus T of i minus 1 j now you substitute this so you will get the same thing.

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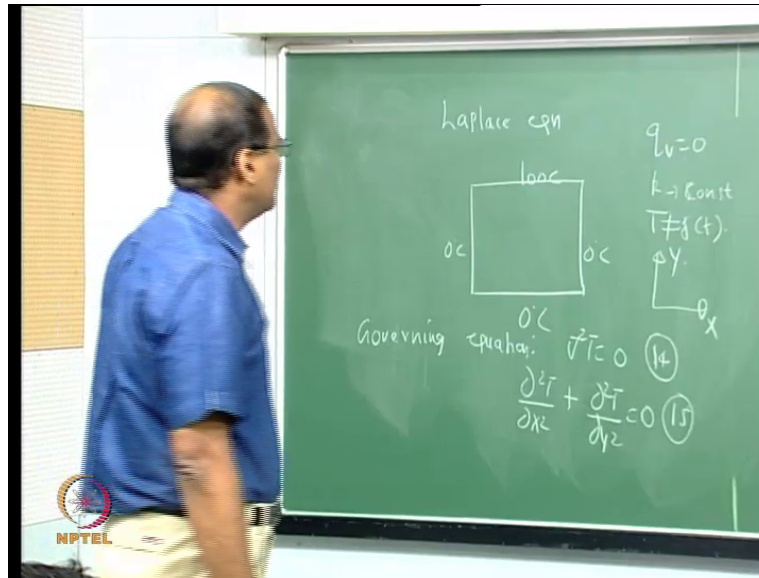
$$= \frac{T_{i+1,j} - 2T_{i,j} + T_{i-1,j}}{\Delta x^2}$$

But, when you do this way it is very difficult the order of accuracy and all that, it is more involved to get it this, but here it was straight forward is also another. So, there are two ways of getting the same thing. Now we will go on to the so we have got the finite difference representation of the first derivative and the second derivative. So, therefore, we can take up a simple partial differential equation, and try to get try to develop a finite difference scheme, for getting the nodal value of a particular node temperature at a particular node, in terms of its neighboring node that is the so for everything for all the the goal of the finite difference approach is to replace all derivatives by algebraic equation. It is a linear combination of its neighbors that linear combination stems from the fact that you can express everything in the form of a Taylor's series.

So, the goal of a finite difference method is finally, to use all the methods you have learnt in linear algebra to solve out this system of equation. Therefore, if you have to be good in finite difference or finite volumes you have to be good in linear algebra. How to invert matrix whether it is worth inverting matrix more than 100 by 100, how to accelerate the convergence of gauss siedl what is successive over relaxation under relaxation. This is for steady for unsteady what is the time step you have to use, number cell Fourier number, grid Fourier number, unconditionally stable, conditionally stable explicit method, implicit method, semi implicit and explicit crank method and so on. So, you have to be good in all this.

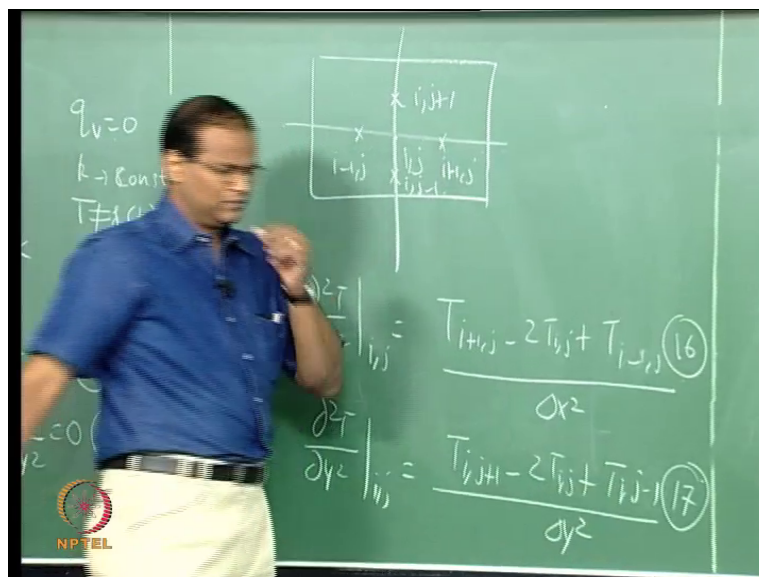
Now now this is the first this is all the preliminaries, mathematically how do you get the finite difference representation. We are heat transfer engineers so our goal is to solve the heat transfer equation. So, we will just take up the Laplace equation and see how we can use this to solve the Laplace equation is it clear up to this stage. d yeah yeah yeah correct is that.

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Let us consider a rectangle slab same same old, steady state 2 dimensional constant properties no heat generation. Regarding equation  $\nabla^2 T = 0$  no  $T$  is naught  $T$  is naught. So, what is the equation number here, so we have to so we will give a number for this this one there is half business we have what is the equation number for this 13. So, we will give a number 14 this, so  $\frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} = 0$ .

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Now, we write the finite difference expressions for both the  $d^2 T$  by  $dx^2$  and  $d^2 T$  by  $Dy^2$ . For that what we will do is we will consider a node  $i, j$  we consider the 4 neighbors'  $i+1, j$  and get the second derivative with respect to  $x$  and the second derivative with respect to  $y$  for  $T$ . any problem. So, I keeping it as  $\Delta x \Delta x$  cannot be changing the node, but let us start with uniform grid so number 16.

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let  $\Delta x = \Delta y$

$$\frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} = 0 =$$

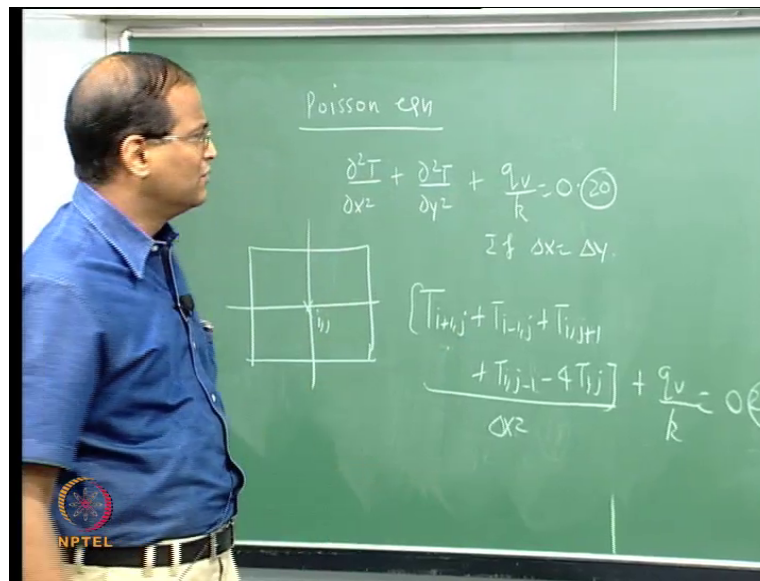
$$\frac{T_{i+1,j} + T_{i-1,j} + T_{i,j+1} + T_{i,j-1} - 4T_{i,j}}{\Delta x^2} = 0 \quad (18)$$

$$\therefore T_{i,j} = \frac{(T_{i+1,j} + T_{i-1,j} + T_{i,j+1} + T_{i,j-1})}{4}$$

Now we can add  $d^2$  by let us say let  $\Delta x$  be equal to  $\Delta y$ , let  $\Delta x$  be equal to  $\Delta y$  therefore,  $d^2 T$  by  $dx^2$  means therefore, therefore, it is very simple. So, the value of the temperature in the particular node is basically the arithmetic average of all the 4 nodes if which are equally which are equispaced from the node  $i, j$ . so, this is the application of the this is the application of the finite difference method to Laplace equation. So, it will be slightly changed if you want to apply this to the Poisson equation where there will be a volumetric heat generation. Volumetric heat generation can take place in a medium if there is nuclear reaction or a chemical reaction which is taking place.

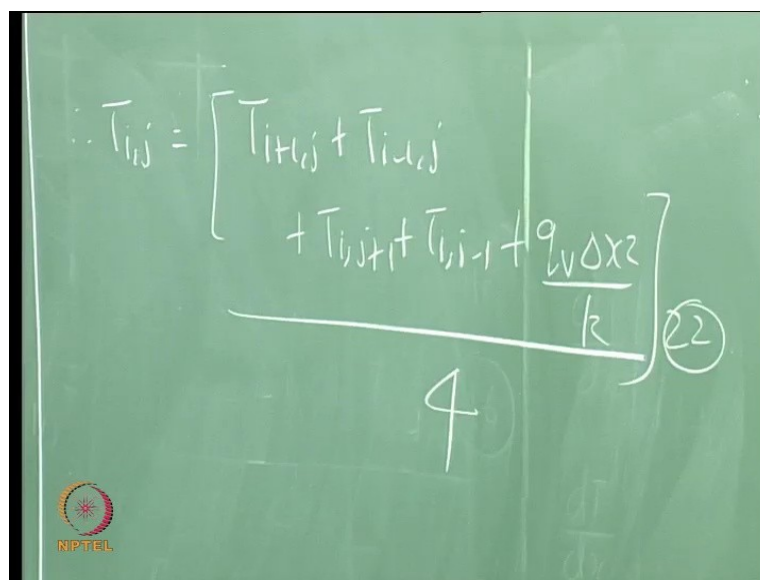


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For example, if you want to incorporate heat conduction I am sorry heat generation plus correct if  $q_v$  is equal to 0 it will reduce to equation 19 equation 22.

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So, that  $q_v \Delta x^2$  by  $k \Delta x^2$  will give an idea of how much will be the contribution from the heat generation term. For example, if the heat generation even if the heat generation 1 into 10 to the power of 6 heat generation 1 into 10 to the power of 6. Suppose if  $\Delta x$  is just 1 millimeter or then  $\Delta x$  will be 1 into 10 to the power of minus

3,  $\Delta x^2$  will be  $1 \times 10^{-6}$ . So,  $1 \times 10^6$  into 1 into the power minus 6, if you have a thermal conductivity value of minimum of 200 then the it will it will not really matter, but if  $k$  it is Bakelite or if it cork and then  $\Delta x$  is small is larger than  $q_v \Delta x^2$  by  $k_v$  will play it is part in the algorithm.

This can be reworked for a non uniform grid space. Non uniform grid size where  $\Delta x_i$  is from  $\Delta y_i$ . so, here you will certain  $T_i$  will get multiplied by  $\Delta y$  by  $\Delta x$  and all that. So, this essentially the formulation. So, you can just keep on formulating for other steady equivalence for the cylindrical as well as the spherical coordinate.

In the next class first 10 minutes we will just take a 100 0 0 problem apply it and see how we can solve it using gauss seidel method we will just start with three four iteration and then I will tell the numerical scheme for a one d unsteady conduction problem and the rudiments of finite volume method also I will explain in the next class.