

Numerical Solutions of Ordinary and Partial Differential Equations
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Lecture - 4
Runge - Kutta Methods for IVPs

Hello. So, we talk about Runge Kutta methods for initial value problems. So, in the last lectures we have talked about Euler method, then modified Euler method. So, we introduced the concept of slope averaging. So, the generalization is if you take an interval, you try to take some intermediate points and then take the weighted average of the slopes. This is the more generalization which leads to Runge Kutta method. So, let us talk about the derivation is little lengthy, because its more technical based on the formula you have to compare with the with the Taylor expansion, and then try to determine the coefficients and weights and all that, so we have to be little patient while understanding the derivation.

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Runge-Kutta Method

$$y_{n+1} = y_n + h \sum_{i=1}^l w_i y'_n(x_n + a_i h)$$

l - slopes $\sum_{i=1}^l w_i = 1$

$x_n + a_i h = x_n \quad (a_i = 0)$

$$y'_n(x_n + a_1 h) = y'_n(x_n) = f(x_n, y_n) = k_1$$

$$y'_n(x_n + a_2 h) = f(x_n + a_2 h, y_n(x_n + a_2 h))$$

$$= f(x_n + a_2 h, y_n + a_2 h f(x_n, y_n))$$

$$= f(x_n + a_2 h, y_n + a_2 h k_1) = k_2$$

So, let us talk about this. So, as I mentioned the concept is within an interval, when you step from x_n you know the data and we try to compute the data at x_{n+1} . So, that is what because this is single step method. So, we have been doing that now what is the idea use several intermediate points $a_i h$. So, on and then take the slopes at each point.

So, that is weighted average that is what I said. So, your approximate solution is y_{n+1} plus h times the weights and slope at these intermediated points.

So, as I mentioned we are taking weights such that and 1 points, 1 intermediate points this 1 denotes 1 slopes, and moreover I started with a 2. So, in $x_{n+1} = x_n + h$. So, that I have taken a 1 equals 0. This is just for convenience. So, that there is a symmetry in that derivation that is all nothing serious about it you can take even this a 1 a 2. There is no harm now when you talk about star. So, what is the first 1 i starts from 1 therefore, x_n is. So, the first entry is y_n of x_n plus a 1 h. So, this is x_n this is, call this k_1 . So, the next entry x_n plus a 2 this is f of x_n plus a 2 h y_n of x_n plus a 2 h. We use Euler so this is y_n plus a 2 h y' of x_n . So, that means in Taylor series, we have neglected the second order onwards that is what we have.

So, this is y_n plus, this y' of x_n is f right. So, this is a 2 h, further this can be generalized in the sense this is equals to f of a 2 h this is k_1 right. So, I introduce k_1 there. So, this let us call k_2 , see this for try to follow we started here and f of x_n plus a 2 h is this. So, then we used Euler for this is just Euler then this is nothing but k_1 as defined. Now, we have defined something called k_2 . So, that means each slope at each point were computing were giving a notation right now we move to the next right. So, what is the next point.

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$$\begin{aligned}
 y'_n(x_n + a_3 h) &= f(x_n + a_3 h, y_n(x_n + a_3 h)) \\
 &= f\left(x_n + a_3 h, y_n + a_3 h \frac{b_{31} y'_n(x_n) + b_{32} y'_n(x_n + a_2 h)}{b_{31} + b_{32}}\right) \\
 &= f(x_n + a_3 h, y_n + h(a_{31} k_1 + a_{32} k_2)) = k_3 \\
 \therefore k_1 &= f(x_n, y_n) \\
 k_2 &= f(x_n + a_2 h, y_n + a_2 h k_1) \\
 k_3 &= f(x_n + a_3 h, y_n + h(a_{31} k_1 + a_{32} k_2))
 \end{aligned}$$

The next point is $y_n + \Delta y$, so this is $f(x_n + \Delta x)$. So, still we have not used the concept of averaging, now the first time we are trying to introduce here because where are we computing, we are computing y' at $x_n + \Delta x$. So, there is already $x_n + \Delta x$ and from here to here. So, there already 2 slopes involved prior to that right. Now, we try to use a averaging concept. So, that is this we have to, this is y_n plus the increment is Δy , then we have y' of x_n right. So, this we tried to get it in weighted average sense.

So, we introduce some notation $w_1 y'$ of x_n plus, $w_2 y'$ of $x_n + \Delta x$. See this is Δy therefore, we have used passed 2 slopes x_n $x_n + \Delta x$. So, this is weighted average. So, this can be simplified this more generalization. So, we are not so much worried about particular constant because ultimately we end up with a generalization in terms of specified constants. So, this is plus, now I take h common here this h look at these constants orbited constants. So, this Δy multiplied by w_1 divide by this. So, this we can call some new constant.

And y' of x_n is just k_1 plus again Δy multiplied by w_2 divide by this, this I call another constant and y' of $x_n + \Delta x$ is k_2 . So, it is more of technical you should try to follow. First one this is at x_n second slope is at $x_n + \Delta x$. So, that is k_2 , then third slope is at $x_n + 2\Delta x$. So, this k_3 , let me define k_1 is this k_2 is, k_1 is involved k_3 both k_1 and k_2 .

So, one can proceed further and try to get, but we do not want to do that because we fix finite number of intermediate points and take the slopes and try to do it. Now, an immediate concern is anybody would ask and who will decide this intermediate points and who will decide this weights and who will tell me how many points to be considered and all the stuff right.

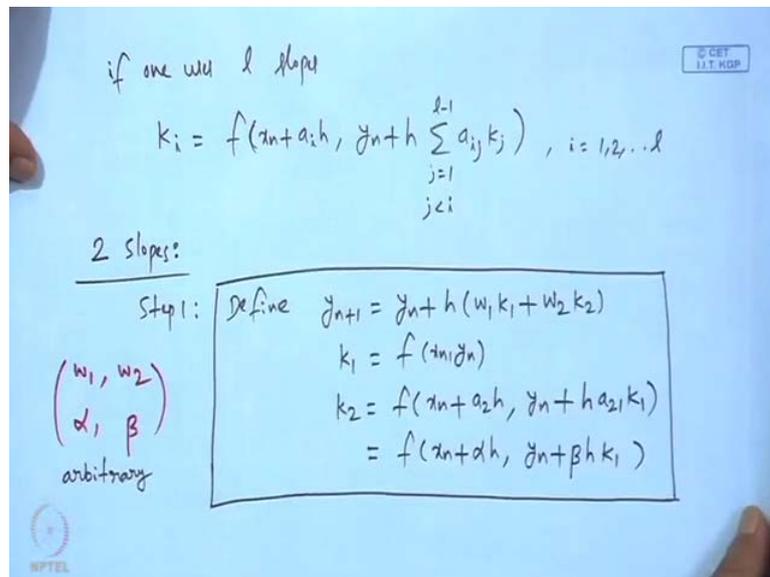
So, we have to really consider this and since already I mentioned this is a bit laborious, may be somebody would start with less number of intermediate points. So, that your weighted average is done with less number of slopes, then you expect a little simpler method first and then straightly complicated. So, let us try to fix the number of intermediate points.

So, if one uses l slopes, we generate k_i of the form $x_n + ih$ $y_n + h$. We have taken common then we get this kind of pattern where i is. So, it is one can identify easily,

see this k_2 is x_n plus $a_2 h$, h is common so $a_2 k_1$ k_3 is x_n plus $a_3 h$, then h comes out then $a_3 k_1$ plus $a_2 k_2$ plus k_2 . So, the generalization is $a_4 k_1$, $a_4 k_2$, $a_4 k_3$, like that for k_4 so this is the generalization right. So, this gets reduced to simpler forms depending on the number of slopes.

So, as I mentioned let us start with just 2 slopes. So, if u consider 2 slopes right, then step one you define the method. So, we are defining the method, y_{n+1} is y_n plus just 2 slopes. So, what will we get, $h w_1 k_1$ plus $w_2 k_2$ i mean instead of the arbitrary constants, earlier notation we can just generalize it. So, this is these are the weights standard I have used. Now, k_1 k_2 contain arbitrary coefficient.

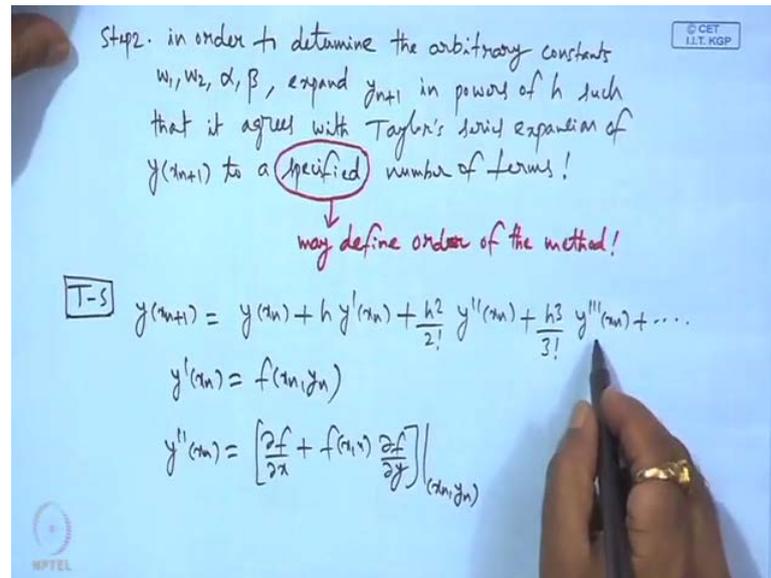
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So, what are they k_1 is f of and k_2 is f of x_n plus $a_2 h$ y_n plus $h a_{21} k_1$. Now, at this stage what is left, yes we have to determine the arbitrary coefficient and the weights. So, to do that I would like to switch to some other notation, just for a convenience because using such indices will be little nasty in the derivation, x_n plus say αh y_n plus $\beta h k_1$. So, I hope you follow right.

So, I am not going with these notations just for convenience head 1 is replaced by beta. Now, what is a next task. So, step 1 we have defined this that means this is going to be our method, k method with 2 slopes provided one determines what are the arbitrary coefficient left, the arbiter coefficient are w_1 w_2 α β . So, these are the orb try coefficient. Now, one has to determine right. So, what is step 2?

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Step 2 is in order to determine the arbitrary constants of coefficient w_1, w_2, α, β what we do. Already I mentioned see what is this is an approximate solution for the given initial problem. Now, approximating what approximating the true solution therefore, who is fellow with whom we have to compare, we have to compare this fellow with the exact solution, but then exact solution is not known to us. So, what we do is your exact solution is if it exist I mean under the sums of existence and uniqueness. Your exact solution will be expanded in terms of l series.

Now once you expand your Taylor Series, compare your Taylor Series with your approximation here. So, let us do that, in order to determine the arbitrary constants w_1, w_2, α, β , expand y_{n+1} that is the approximation in powers of h , such that it agrees with Taylor series expansion of true. That is this and how do you compare how long to a specified number of terms.

Now, when you say specified number of terms, may be this is specified number of terms. This may define may define order of the method. So, comparing up to this order that means that will dictate the order of the method. So, let us do that as I mentioned it is going to be very, very technical so we have to follow the steps. So, let us expand Taylor series first.

So, I am writing the t s that is Taylor series expansion first, y of x_{n+1} is y of x_n plus so on. We have this terms, y dashed is f of x_n, y_n , y double is $\text{dou } f \text{ dou } x \text{ plus } f \times y$

So, f and y evaluated at x_n, y_n , I mean it is better to compute the next time as well before we proceed with comparison. So, let us do it straightly complicated, but then we need this term y^3 .

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$$y^{(4)}(x_n) = \frac{\partial^4 f}{\partial x^4} + 2f \frac{\partial^2 f}{\partial x \partial y} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial f}{\partial y} \left(\frac{\partial f}{\partial x} + f \frac{\partial f}{\partial y} \right) \Big|_{(x_n, y_n)}$$

Expansion of y_{n+1} : h-exp

$$k_1 = f(x_n, y_n)$$

$$k_2 = f(x_n + \alpha h, y_n + \beta h k_1)$$

$$= f(x_n, y_n) + \alpha h \frac{\partial f}{\partial x} + \beta h k_1 \frac{\partial f}{\partial y}$$

$$+ \frac{1}{2!} \left(\alpha^2 h^2 \frac{\partial^2 f}{\partial x^2} + 2\alpha\beta h^2 k_1 \frac{\partial^2 f}{\partial x \partial y} + \beta^2 h^2 k_1^2 \frac{\partial^2 f}{\partial y^2} \right)$$

$$+ \dots$$

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So, y^3 one can do it 2 f of course, all this validated at. So, this as I mentioned this is lightly complicated. So, we have to see, this is supposed to be the exact. Now, we have the approximation this is approximation. So, what is our aim, our aim is to expand this in powers of h so that we compare with this. So, it's slightly complicated, let us try to do that. Now, our immediate concern is to expand this in powers of h that means see for example, k_1 there is nothing to expand, but when you take k_2 , there is you have to expand this right. So, let us do that, y^3 is done. Now, we proceed to expansion of y_{n+1} . So, this I call h exp.

So, k_1 there is nothing, k_2 it is better to write the form of k_2 , f of $x_n + \alpha h, y_n + \beta h k_1$. So, this is f of x, y so it is a 2 variable expansion Taylor series. So, we have to expand this plus αh with respect to this variable. So, I am not writing the evaluation point. So, it is understood at each step I have to write, but one can use generalized notation just plus first order terms I am writing first f expanded with respect to this, then the first order term.

Now I will expand with respect to the first order term with respect to this. So, that will be $\beta h k_1$ f by f because with respect to this variable. So, this the first order

terms, plus second order terms, this is the increment. So, alpha square h square, second order derivative plus the mix term 2 alpha h beta h.

So, that will be 2 alpha beta h square k 1 this is the mix term plus second order term of with respect to y beta square h square k 1 square plus well look at this. So, we have to expand with respect x and y. So, this function about the point x n y n, then first order terms alpha x so f by so x plus this is increment for y therefore, the increment times dou f by dou y. Now, the second order terms, 1 naught 2 factorial. So, this is second order term of this, this is second order term of this and this is the next term. Now, this can be simplified as follows.

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$$k_2 = f(x_n, y_n) + h \left(\alpha \frac{\partial f}{\partial x} + \beta f \frac{\partial f}{\partial y} \right) + \frac{h^2}{2!} \left(\alpha^2 \frac{\partial^2 f}{\partial x^2} + 2\alpha\beta f \frac{\partial^2 f}{\partial x \partial y} + \beta^2 f^2 \frac{\partial^2 f}{\partial y^2} \right) + \dots$$

$$y_{n+1} = y_n + h w_1 k_1 + h w_2 k_2 = y_n + h w_1 f + h w_2 \left[f + h \left(\alpha \frac{\partial f}{\partial x} + \beta f \frac{\partial f}{\partial y} \right) + \frac{h^2}{2!} \left(\alpha^2 \frac{\partial^2 f}{\partial x^2} + 2\alpha\beta f \frac{\partial^2 f}{\partial x \partial y} + \beta^2 f^2 \frac{\partial^2 f}{\partial y^2} \right) + \dots \right]$$

So, k 2 is given by plus h so beta, we have k 1 there. So, that k 1 is nothing but f so that becomes an f there plus because what is our aim, our aim is to compare with the Taylor series expansion. So, our Taylor series expansion has exactly this kind of form h h square by 2 factorial. So, we are trying to put it in that form . So, this is k 2, what did we do, we are trying to get the h expansion of that approximation.

So, k 1 is this k 2 we have expanded, now what we have to do, we have to substitute the expansions in y n plus 1. So, let us try to do that therefore, your y n plus 1 is y n plus h w 1 k 1 plus h, this is our approximation. So, this is given by y n plus h w 1 k 1 is just f plus h w 2 k 2, this entire expression we have to write down f plus h. This is first term then let me put it here.

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$$y_{n+1} = y_n + h(w_1 + w_2)f + w_2 h^2 (\alpha \frac{\partial f}{\partial x} + \beta f \frac{\partial f}{\partial y}) + \frac{h^3}{2} w_2 (\alpha^2 \frac{\partial^2 f}{\partial x^2} + 2\alpha\beta f \frac{\partial^2 f}{\partial x \partial y} + \beta^2 f^2 \frac{\partial^2 f}{\partial y^2}) + \dots$$

(h-exp)

$$y^{(n+1)} = y^{(n)} + h f + \frac{h^2}{2} (\frac{\partial f}{\partial x} + f \frac{\partial f}{\partial y}) + \frac{h^3}{6} (\frac{\partial^2 f}{\partial x^2} + 2f \frac{\partial^2 f}{\partial x \partial y} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial f}{\partial x} (\frac{\partial f}{\partial x} + f \frac{\partial f}{\partial y}))$$

(T-exp)

compare (h-exp) & (T-exp) to match the coefficients of equal power of h

So, this can be further simplified as follows, y_{n+1} plus h . Let me explain the we get this look our aim is to put it in a Taylor series form, so coefficient of h coefficient of h square. So, look at this you have one term here and with $h w_2 f$. So, if you take common we get exactly this term. So, similarly, we get this is the next term. So, h square, how do we get it, look h square w_2 times this now h cube from this term. So, plus $2\alpha\beta$ is a mix term.

So, this is our h expansion and what was our t expansion. So, this is Taylor series expansion this is our Taylor series with these 2 terms. So, maybe its better to write down y of x plus $h f$ plus h square by 2 $\frac{\partial f}{\partial x} + f \frac{\partial f}{\partial y}$ plus h cube by 6 $\frac{\partial^2 f}{\partial x^2} + 2f \frac{\partial^2 f}{\partial x \partial y} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial f}{\partial x} (\frac{\partial f}{\partial x} + f \frac{\partial f}{\partial y})$. So, this our Taylor series expansion. Now, what is our aim, aim is to compare as I mentioned the next task is, to compare the Taylor series expansion and the h expansion up to desired terms.

So, that means you have coefficient of h coefficient of h square so we compare. So, that we get some we except a system of equations let us see. So, again I would like to this is an approximation which were x expand in powers of h and this is our Taylor series expansion, now we try to compare. So, what we do is compare h expansion and t expansion to match the coefficient of equal powers of h . So, let us look at it y_n must be approximating y of x and then look you have $h f$. So, $h f$, w_1 and plus w_2 must be 1.

So, this is approximating this, then this you have $h f$ in this term. So, if you try to compare this two, what happen w_1 plus w_2 must be 1 right. So, let us try to do that. So, if we do that we get from first case, then so let me use again red. So, this is red these 2 then the green, now h^2 upper I am using. So, you have h^2 h^2 by 2. So, there is a half a term for df by dx , if you remove h^2 there is just a factor of half a term for df by dx whereas, here you have αw_2 sitting. Therefore, αw_2 must be half there and here βw_2 is sitting for df by dy and there is another half sitting for df by dy therefore, βw_2 must be half there.

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$$\begin{aligned}
 & \left. \begin{aligned} w_1 + w_2 &= 1 \\ \alpha w_2 &= \frac{1}{2} \\ \beta w_2 &= \frac{1}{2} \end{aligned} \right\} \text{choosing } \alpha \neq 0, \beta = \alpha, \\
 & \qquad \qquad \qquad w_2 = \frac{1}{2\alpha} \\
 & \qquad \qquad \qquad w_1 = 1 - \frac{1}{2\alpha} \\
 & \therefore y_{n+1} = y_n + h \left[\left(1 - \frac{1}{2\alpha}\right) k_1 + \frac{1}{2\alpha} k_2 \right] \\
 & \qquad k_1 = f(x_n, y_n) \\
 & \qquad k_2 = f(x_n + \alpha h, y_n + \alpha h k_1)
 \end{aligned}$$

(RK-2)

We get this is one equation, then αw_2 is half, βw_2 is half. So, this I have done it only up to how many terms up to order of h^2 . Now, compared so y_n is approximate of $y(x_n)$ and by comparing by this we get, w_1 plus w_2 is 1 and comparing this h^2 terms we get αw_2 is half βw_2 is half. So, I am not comparing these three terms. So, that means I have compared up to terms up to h^2 right now from this system. If you stop then we should be able to determine the coefficient, but suppose let us stop at this stage. That means you have compared up to h^2 .

You should be able to solve, but how many unknowns you have, w_1 , w_2 , α and β and you have only three unknowns right. So, that means you have to choose one of them arbitrary right. Let us say choosing $\alpha \neq 0$, we get $\beta = \alpha$ w_2 is $1/(2\alpha)$ w_1 is $1 - 1/(2\alpha)$. So, with this our approximation becomes

$y_n + h$, what was our approximation, these are our approximation h into $w_1 k_1 + w_2 k_2$. So, we have w_1, h into our w_1 is $1 - \frac{1}{2}\alpha$, k_1 plus w_2 is $\frac{1}{2}\alpha$.

So, this is our approximation where k_1 is given by f of $x_n + \alpha h, y_n + \alpha h k_1$. So, this is our RK method, still α is arbitrary. Now, you can talk about specific values, so let me write this as RK2 because I have compared only up to h^2 therefore, maybe we are optimistic just by comparing up to h^2 . We can call this method second order probably right, any way we can discuss that. Now, since α is arbitrary let us take some specific value and try to see right.

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(i) if $\alpha = \beta = \frac{1}{2}$; $w_1 = 0, w_2 = 1$

$$y_{n+1} \approx y_n + h f\left(x_n + \frac{h}{2}, y_n + \frac{h}{2} f\right)$$

(ii) $\alpha = 1; \beta = 1, w_1 = \frac{1}{2} = w_2$

$$y_{n+1} \approx y_n + \frac{h}{2} (k_1 + k_2)$$

$$k_1 = f$$

$$k_2 = f(x_n + \alpha h, y_n + \beta h k_1)$$

$$= f(x_n + h, y_n + h f)$$

If α is half, then β will also half then what will happen to your method α is half right. So, this fellow is zero and this fellow is 1 right, if α is half. So, let us if α is half, β is half then we get $w_1 = 0, w_2 = 1$ and you method reduces to $y_n + h f$ is $y_n + h f$ of $x_n + \frac{h}{2}, y_n + \frac{h}{2} f$. So, this is a method we get. So, $y_n + h f$ is a method α is half. So, this coefficient vanishes and α is half. So, this we get $w_1 = 0, w_2 = 1$ and β equal to α equal to half. Therefore, the method reduces to this one. Suppose, α equals to one then β equals to 1 then w_1 equals to half and which is also w_2 .

So, in this case $y_n + h f$ is $y_n + h f$ of $x_n + \alpha h, y_n + \alpha h k_1$ plus $w_2 k_2$. So, this reduces to h times $k_1 + k_2$ and what is our k_1, k_2 is f and k_2 is f of $x_n + \alpha h, y_n + \alpha h k_1$.

n plus $\beta h k$ 1 this reduces to f of x n plus $h y$ n plus $h k$ 1 k 1 is f . So, this is the particular case when α is 1 α is. So, these are general choices now the story is not left we have compared up to 2 terms and then shown as system of this form and then one is arbitrary. So, of course, taking non zero, one can have different choice right. So, general choices are α is half and α is 1. Now, what happens to the terms beyond h square, obviously they should contribute to the error.

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Since $(T-rn)$ & $(h-rn)$ agree up to $O(h^2)$,
the difference

$$\epsilon_{n+1} = y^{(n+1)} - y_{n+1}$$

$$= h^3 \left[\left(\frac{1}{6} - \frac{\alpha}{4} \right) \left(\frac{d^3f}{dx^3} + 2f \frac{d^2f}{dx^2} + f^2 \frac{d^2f}{dy^2} \right) + \frac{1}{6} \frac{d^2f}{dy} \left(2 \frac{df}{dx} + f \frac{df}{dy} \right) \right] \Big|_{(x_n, y_n)}$$

$$\approx O(h^3)$$

Remark: No choice of α make the leading term vanish for all $f(x,y)$

So, if you consider the difference, since t expansion and h expansion agree up to h square, what will happen to the difference. The difference must be the residual that is error. So, naturally that should from h cube. So, if you can do the long calculation, y 3 we have computed any way this is in our hand. So, this is the residual which is supposed to be the error of course, at a given. Since, we know f and we have determine α . So, what will happen, this entire thing is known to us and hence this is order of h cube. So, the residual is of order h cube and hence the method is of order h square.

So, one remark see for example, here for all f means not for a particular f this is becoming 0 and for all of no choice of α will make the leading term vanish what is the leading term in this is the leading term. So, further you have h 4, no choice of α make the leading term vanish for all f of $x y$. So, this is an important mark to conclude that really the error is coming from the order of h cube. So, we have approximated with two slopes in generalized formula, but we have considered only two slopes.

Then we have taken the weighted average and expanded the approximate formula expand the Taylor series formula. Then determined the arbitrary coefficient up to h square right and then conclude that the residual comes from h cube. Hence this is the second order method so this is RK 2 that is what I have written. Now, suppose one would like to define a higher order method R K method what one has to do, one has to compare a the terms between the Taylor series expansion and the h series expansion of the approximate and try to determine the arbitrary coefficient.

So, this will determine the method because once you determine the coefficient the method is determined and then you get a new method. So, how long we try to determine of course, this is algebraically you try to do you get very complex system. So, the standard methods are first one you arrive at second order method. There is a choice of one arbiter coefficient and more general choice are half and 1.

So, these are R K second order, but they slightly vary. Now, if you really go for more number of terms, then you get let us say you go to 4 terms. Then you get four in the sense you got to compare h optical h power 4, you get fourth order R K method. So, that will be slightly complicated.

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example $y' = (x^2 + 2xy), \quad y(0) = 1$
 $h = 0.1$

$$y_{n+1} \approx y_n + \frac{h}{2} [f(x_n, y_n) + f(x_n+h, y_n+h f)]$$

$$y(0.1) = y_0 + \frac{0.1}{2} [f(x_0, y_0) + f(x_0+0.1, y_0+0.1 f(x_0, y_0))]$$

$$f(x_0, y_0) = x_0^2 + 2x_0 y_0 = 0$$

$$f(x_0+0.1, y_0+0.1 f(x_0, y_0)) = (0.1)^2 + 2(0.1) = 0.01 + 0.2 = 0.21$$

$$y(0.1) = 1 + \frac{0.1}{2} (0.21)$$

$$= 1 + 0.05 (0.21) \approx$$

$$y(0.2) =$$

So, before for that let us just see some example. So, this is y dashed is x square plus 2 x y and y of 0 is 1 and say h is 0.1. So, then whatever the method we have determined, let us say alpha 1 case. So, alpha 1 case we have y n plus 1 is y n plus h by 2 k 1 is just f, there

$plus k^2$ is f of x_n plus $h y_n$ plus $h f$. So, this is the method we have. Suppose, we try to determine y at $t = 0.1$. So, this will be y_0 plus f of x_0 y_0 plus f of x_0 plus $0.1 y_0$ plus 0.1 into f of. So, we had compute each term, what is f of, this is x_0 square. So, x_0 is 0, this is 0. Therefore, so y_0 is $1 + h$ by 2. This is 0 and f of we need this term right and this is 0. So, this reduces to this is 0, this is gone. So, this reduces to f of 0.1 and 1. So, this is 0.1 square.

So, this will be where we are. So, point h by 2 this is 0 and this is reduced to. So, we get the value. Next we compute a good exercise for you is for the same example you try to compute the values at least 2-3 values, y point 1 point 2 point 3, using Euler method and then R K method. Since, they form of f is very simple one can compare with the exact solution to know which gives slightly at the rate. So, in the next class we try to talk about R K fourth order and may be once we define the order of the method, then we have to talk slightly more about the error.

For example, you are computing y at $t = 0.1$, then there is some error introduced at that stage then to compute y at $t = 0.2$ you are using the value of y at $t = 0.1$, so whatever the error introduced at 0.1 that will propagate. So, we have to really think of we have to really think of how these errors will be propagating and try to control. So, will talk on R K fourth order and then how in general error estimates are computed.

Thank you.