

# Ordinary Differential Equations

## Initial Value Problems

*The question of whether computers can think is just like  
the question of whether submarines can swim*

*Edsger W. Dijkstra*

# Topics to Be Discussed

- **This unit requires the knowledge of some very basic ordinary equations.**
- **The following topics will be presented:**
  - **Euler's method**
  - **Heun's method**
  - **Predictor-Corrector methods**
  - **Various Runge-Kutta methods with an emphasis on the second order methods**

# What Is an Initial Value Problem?

- A first order differential equation has a form like this:

$$y' = f(x, y)$$

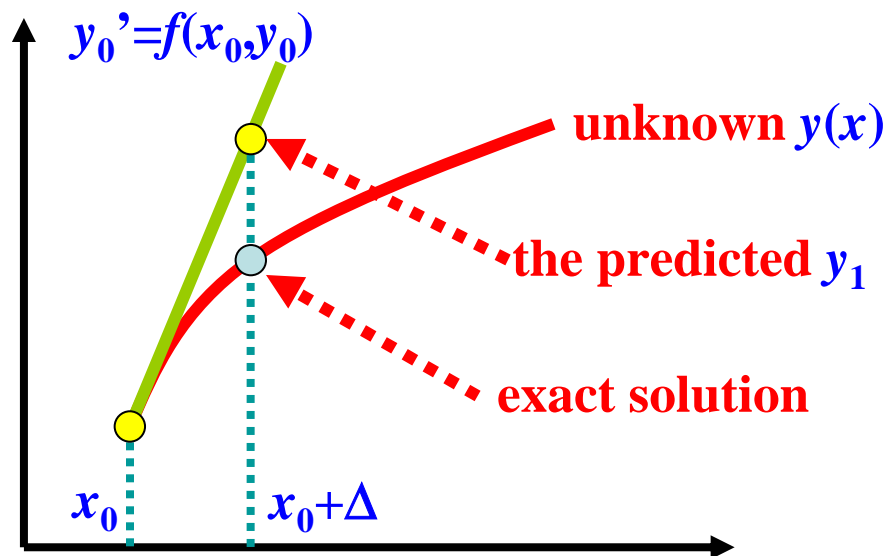
- Here,  $y$  is a function of  $x$ , and  $y'$  is the derivative of  $y$ . Function  $f(x, y)$  gives the relation of  $x$  and  $y$ .
- However, solution to the above ODE is not unique, and an initial condition is required:

$$y' = f(x, y) \quad \text{and} \quad y(x_0) = y_0$$

- The problem is: find  $y$  with  $y(x_0) = y_0$  step-by-step in an interval  $[x_0, x_n]$ .

# Euler's Method: 1/6

- The easiest method is perhaps Euler's.
- Since  $y' = f(x,y)$  and  $y_0 = y(x_0)$ , we know the tangent slope of  $y$  at  $(x_0, y_0)$ , *i.e.*,  $y_0' = f(x_0, y_0)$ .
- If  $x_1 = x_0 + \Delta$  and  $\Delta$  is small enough, we may **predict**  $y(x+\Delta)$  using the tangent!



## Euler's Method: 2/6

- How do we “predict”  $y(x+\Delta)$ ?
- Using the forward difference method for differentiation,  $y'(x_0)$  can be approximated as

$$y'(x_0) \approx \frac{y(x_0 + \Delta) - y(x_0)}{\Delta}$$

- Replacing  $y'$  yields:

$$\frac{y(x_0 + \Delta) - y(x_0)}{\Delta} \approx f(x_0, y(x_0))$$

- Therefore, the prediction is:

$$y(x_0 + \Delta) \approx y(x_0) + \Delta \times f(x_0, y(x_0))$$

# Euler's Method: 3/6

- Recall the following formula:

$$y(x_0 + \Delta) \approx y(x_0) + \Delta \times f(x_0, y(x_0))$$

- We may start with an initial value  $(x_0, y_0)$ .  
Compute  $x_1 = x_0 + \Delta$  and  $y_1 = y_0 + \Delta \times f(x_0, y_0)$ .
- Then, compute  $(x_2, y_2)$  from  $(x_1, y_1)$ , etc.
- The algorithm below traces out the locus of  $y$ .

```
! Initial values
! [a,b]
! x0 = a
! y0 = y(a) = y(x0)
! n: # of subdivisions
```

```
Δ = (b-a) / n
x = a
y = y(a)
DO i = 1, n
    y = y + Δ × f(x, y)
    x = x + Δ
END
```

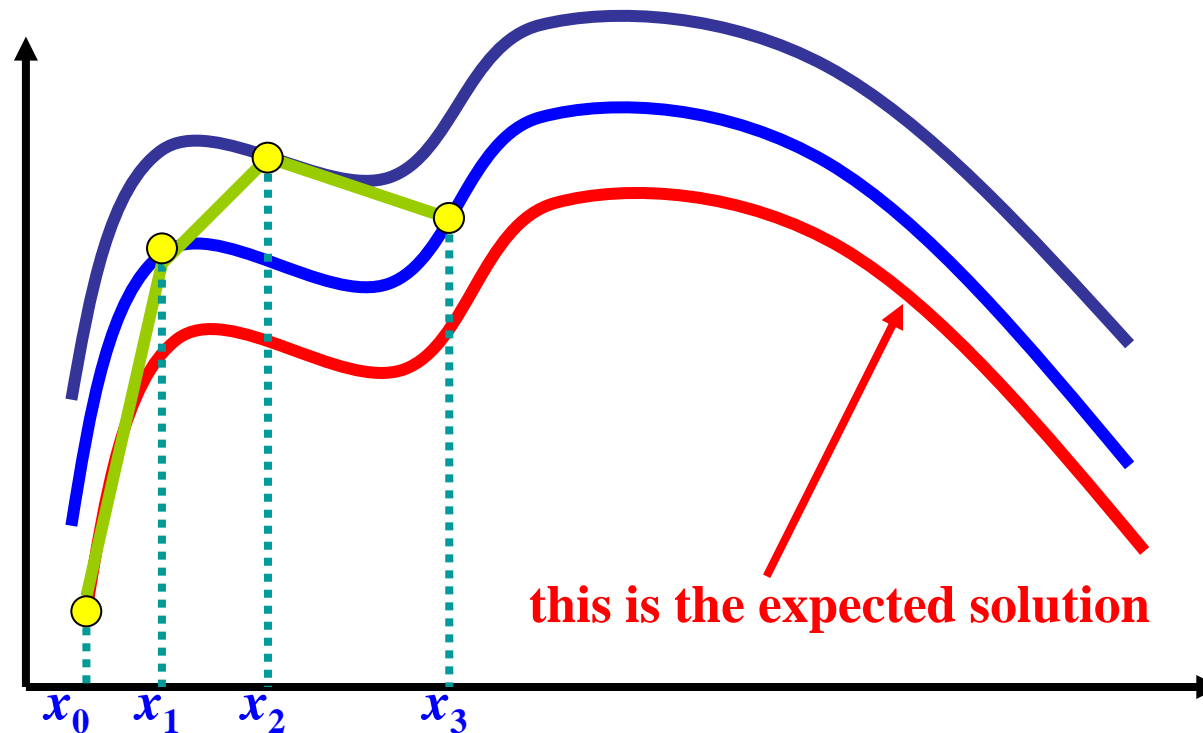
# Euler's Method: 4/6

- Consider  $y' = x + y$  and  $y(0)=2$  on  $[0,1]$ .
- Initial value is  $x_0=0$  and  $y_0=y(x_0)=2$ .
- If  $n = 4$ ,  $\Delta=(1-0)/4=0.25$
- Predicted (next)  $y = y + \Delta(x+y)$ .

	$x$	$y$	$y'=f(x,y)$	Pred. $y$
Init	0	2.0	2.0	$2.5 = 2 + 0.25 \times (0 + 2)$
1	0.25	2.5	2.75	$3.1875 = 2.5 + 0.25 \times 2.75$
2	0.5	3.1875	3.6875	$4.109375 = 3.1875 + 0.25 \times 3.6875$
3	0.75	4.109375	4.859375	$5.3245187 = 4.109375 + 0.25 \times 4.859375$
4	1	5.3245187		

# Euler's Method: 5/6

- If  $\Delta$  is sufficiently small, Euler's method usually works fine.
- However, Euler's method can produce large under- or over- predicted  $y$ .



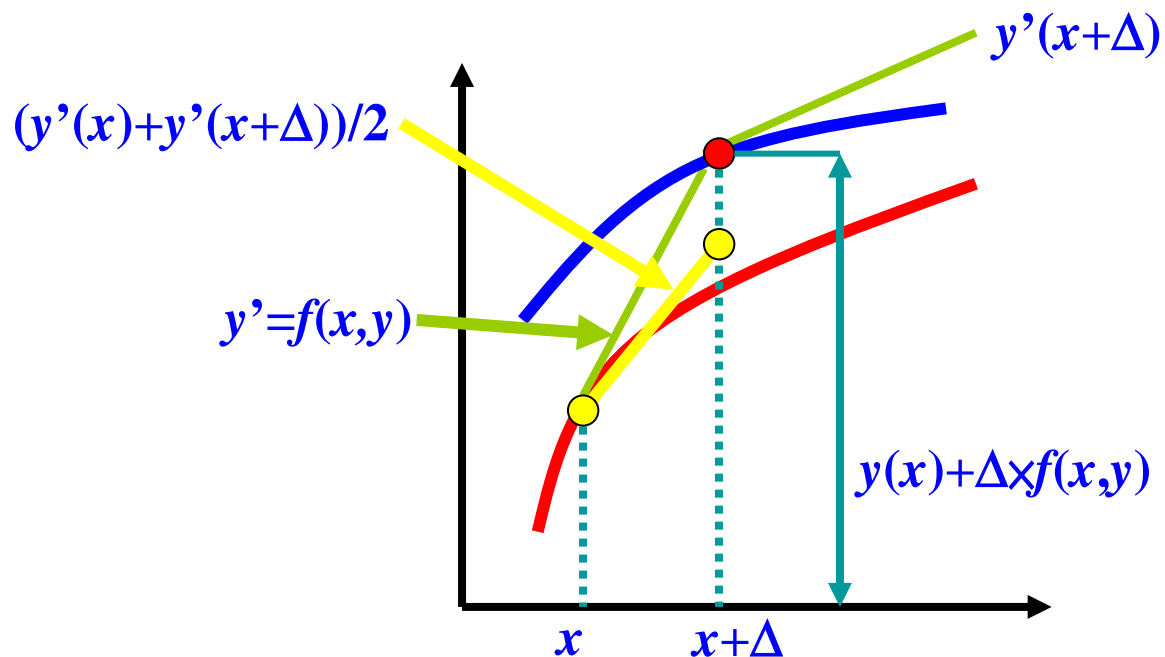
# Euler's Method: 6/6

- Consider  $y' = 1-y$  on  $[0,1]$  with  $y(0)=0$ .
- The solution is  $y=1-e^{-x}$ .
- The absolute error is increasing.

	$x$	$y$	exact $y$	error	% of exact
1	1/6	0.1666667	0.1535182	0.01314841	8.56
2	1/3	0.3055556	0.2834687	0.02208686	7.79
3	1/2	0.4212963	0.3934693	0.02782699	7.07
4	2/3	0.5177469	0.4865829	0.03116405	6.40
5	5/6	0.5981224	0.5654018	0.03272063	5.79
6	1	0.6651020	0.6321206	0.03298146	5.22

# Heun's Method: 1/4

- Heun's method tries to reduce the over- or under- shot of Euler's method.
- Since  $y(x+\Delta)$  may be an overshoot, one may use the average of  $y'(x)$  and  $y'(x+\Delta)$  as a corrector.



# Heun's Method: 2/4

● Heun's method consists of two steps:

□ **Step 1:** Use  $x$  and  $\Delta$  to compute  $y^P(x+\Delta)$  with Euler's method. This is a *predictor*:

$$y^P(x + \Delta) = y(x) + \Delta f(x, y)$$

□ **Step 2:** Use the  $(y'(x) + y'(x+\Delta))/2$  as the new derivative to correct over- or under-shot. This is a *corrector*:

$$y(x + \Delta) = y(x) + \Delta \frac{f(x, y) + f(x + \Delta, y^P(x + \Delta))}{2}$$

# Heun's Method: 3/4

- Heun's method is just a little more complex than Euler's method.

```
! initialization  
! [a,b]: interval  
!  $x_0 = a$   
!  $y_0 = y(a) = y(x_0)$   
! n : intervals
```

```
 $\Delta = (b-a) / n$   
 $x = a$   
 $y = y(a)$   
DO i = 1, n  
   $y^{\text{pred}} = y + \Delta \times f(x, y)$   
   $y = y + \Delta \times (f(x, y) + f(x+\Delta, y^{\text{pred}})) / 2$   
   $x = x + \Delta$   
END DO
```

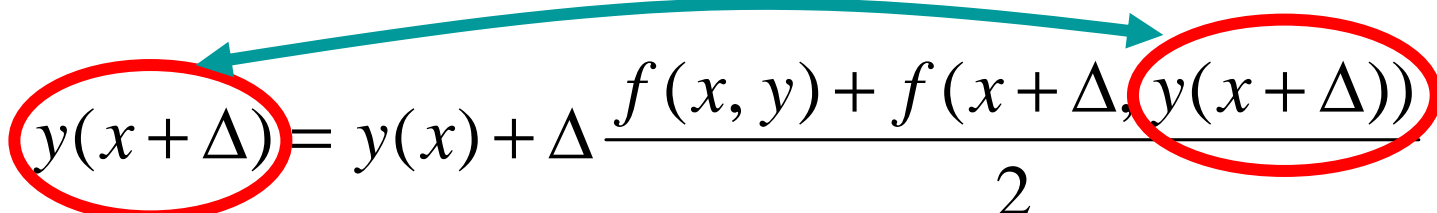
## Heun's Method: 4/4

- Consider  $y' = 1-y$  on  $[0,1]$  with  $y(0)=0$ , again.
- The solution is  $y=1-e^{-x}$ .
- The absolute error is still increasing, but significantly smaller.

	$x$	$y$	exact $y$	error	% of exact
1	1/6	0.15277778	0.15351826	0.00074048	0.48
2	1/3	0.28221452	0.28346872	0.00125420	0.44
3	1/2	0.39187619	0.39346933	0.00159314	0.40
4	2/3	0.48478401	0.48658288	0.00179887	0.37
5	5/6	0.56349754	0.56540179	0.00190425	0.34
6	1	0.63018543	0.63212055	0.00193512	0.31

# Predictor-Corrector Methods: 1/5

- Heun's method is a special and the simplest form of the predictor-corrector methods.
- If we look at the corrector step and remove the superscript  $P$  from the predictor, we have

$$y(x + \Delta) = y(x) + \Delta \frac{f(x, y) + f(x + \Delta, y(x + \Delta))}{2}$$


- $y(x + \Delta)$  is the unknown value that we want to compute, and it appears in *both* sides of the equation.

# Predictor-Corrector Methods: 2/5

- To make this observation clearer, replacing  $y(x+\Delta)$  with a  $z$  yields:

$$z = y(x) + \Delta \frac{f(x, y) + f(x + \Delta, z)}{2}$$

- Since  $z$  is an unknown, the above is an equation of variable  $z$ , and **the desired  $z$  is a root!**
- If such a  $z$  can be found, say  $z^*$ , the equation is perfectly balanced as follows:

$$z^* = y(x) + \Delta \frac{f(x, y) + f(x + \Delta, z^*)}{2}$$

- Therefore, this predictor-corrector method reduces to root finding.

# Predictor-Corrector Methods: 3/5

- How to solve the following equation in  $z$ ?

$$z = y(x) + \Delta \frac{f(x, y) + f(x + \Delta, z)}{2}$$

- Since it is not easy to find the derivative with respect to  $z$ , Newton's method is out.
- However, it is in a perfect form of fixed-point iteration (*i.e.*,  $z = g(z)$ ).
- We may use  $y^P$  as an initial value, followed by a number of iterations of fixed-point iteration to find a better corrector  $y^C$ .
- Note that fixed-point iteration method does not always converge.

# Predictor-Corrector Methods: 4/5

- The following is the iterative computation for a better corrector with the fixed-point iteration.
- **MAX** is the maximum number of iterations, and  $\epsilon$  is a tolerance value.
- There are more powerful predictor-corrector methods.

```
ypred = y + Δx f(x, y)
DO k = 1, MAX
  ycorrect = y + Δx (f(x, y) + f(x+Δ, ypred)) / 2 ! Fixed-point
  IF ( | (ycorrect - ypred) / ycorrect | < ε) EXIT
  ypred = ycorrect
END DO
y = ycorrect
```

# Predictor-Corrector Methods: 5/5

- Consider  $y' = 1-y$  on  $[0,1]$  with  $y(0)=0$ , again.
- The solution is  $y=1-e^{-x}$ .
- The absolute error is still increasing, but better than Heun's method.

$\varepsilon = 0.0005$

	$x$	$y$	exact $y$	error	% of exact	Iters
1	1/6	0.15384677	0.15351826	0.00032851	0.21	4
2	1/3	0.28401792	0.28346872	0.00054920	0.19	3
3	1/2	0.39416370	0.39346933	0.00069436	0.18	3
4	2/3	0.48736477	0.48658288	0.00078189	0.16	3
5	5/6	0.56622791	0.56540179	0.00082612	0.15	3
6	1	0.63295889	0.63212055	0.00083834	0.13	3

# Runge-Kutta Methods: 1/22

- Runge-Kutta methods is **a family** of methods.
- In fact, Runge-Kutta methods are extensions to many popular methods, including Euler's and Heun's methods.
- Euler's method computes  $y(x+\Delta)$  by adding  $\Delta \times f(x,y)$  to the current  $y$ .
- Runge-Kutta methods try to express this correction as a **linear combination** of a number of “correction” steps.

# Runge-Kutta Methods: 2/22

- Runge-Kutta methods express  $y(x+\Delta)$  in the following way:

$$y(x + \Delta) = y(x) + \Delta \times \Omega(x, y, \Delta)$$

- Function  $\Omega(x,y,\Delta)$  is a function of the current point  $(x,y)$  and the step-size  $\Delta$ , and is usually referred to as the *increment function*.

- The increment function is the linear combination of two sets of values,  $c_i$ 's and  $k_i$ 's ( $1 \leq i \leq n$ ):

$$\Omega(x, y, \Delta) = c_1 k_1 + c_2 k_2 + \cdots + c_n k_n = \sum_{i=1}^n c_i k_i$$

# Runge-Kutta Methods: 3/22

- The  $c_i$ 's are values to be determined, and are usually  $c_i \geq 0$  and  $c_1 + c_2 + \dots + c_n = 1$ .
- The  $k_i$ 's are defined by a recurrence relation similar to Heun's method.
- Here are the definitions of the  $k_i$ 's:

$$k_1 \leftarrow = f(x, y)$$

$$k_2 \leftarrow = f(x + p_2 \Delta, y + (a_{2,1} k_1) \Delta)$$

$$k_3 \leftarrow = f(x + p_3 \Delta, y + (a_{3,1} k_1 + a_{3,2} k_2) \Delta)$$

⋮

$$k_n \leftarrow = f(x + p_n \Delta, y + (a_{n,1} k_1 + a_{n,2} k_2 + \dots + a_{n,n-1} k_{n-1}) \Delta)$$

# Runge-Kutta Methods: 4/22

- Let us look at  $p_i$ 's,  $a_{i,j}$ 's and  $k_i$ 's more closely:

$$k_1 = f(x, y)$$

$$k_2 = f(x + p_2\Delta, y + (a_{2,1}k_1)\Delta)$$

$$k_3 = f(x + p_3\Delta, y + (a_{3,1}k_1 + a_{3,2}k_2)\Delta)$$

⋮

$$k_n = f(x + p_n\Delta, y + (a_{n,1}k_1 + a_{n,2}k_2 + \cdots + a_{n,n-1}k_{n-1})\Delta)$$

- Those  $p_i$ 's and  $a_{i,j}$ 's are predetermined values.
- $k_1$  is  $f(x,y)$ ,  $k_2$  uses  $k_1$  in a way like a corrector.
- $k_3$  uses  $k_1$  and  $k_2$  for “correction” purpose, etc.
- A Runge-Kutta method uses up to  $k_j$  is referred to as an *order j* Runge-Kutta method.

# Runge-Kutta Methods: 5/22

- Consider a *first-order* Runge-Kutta method.
- It uses  $k_1 = f(x, y)$  and the increment function is:

$$\Omega(x, y, \Delta) = c_1 k_1$$

- Thus, first order Runge-Kutta method is

$$y(x + \Delta) = y(x) + \Delta \times c_1 k_1 = y(x) + \Delta \times c_1 f(x, y)$$

- If we chose  $c_1 = 1$ , first-order Runge-Kutta method becomes Euler's method.

# Runge-Kutta Methods: 6/22

- Heun's method calculates  $y(x+\Delta)$  as follows:

$$y(x+\Delta) = y(x) + \Delta \frac{f(x, y) + f(x+\Delta, y^P(x+\Delta))}{2}$$

- $y^P(x+\Delta)$  is computed as follows:

$$y^P(x+\Delta) = y(x) + \Delta f(x, y)$$

- Therefore, Heun's method is a second-order Runge-Kutta method:

$$\begin{aligned}
 y(x+\Delta) &= y(x) + \Delta \left[ \frac{1}{2} f(x, y) + \frac{1}{2} f(x+\Delta, y^P(x+\Delta)) \right] \\
 &= y(x) + \Delta \left[ \frac{1}{2} \overset{k_1}{\boxed{f(x, y)}} + \frac{1}{2} \overset{k_2}{\boxed{f(x+\Delta, y + \overset{k_1}{\boxed{f(x, y)\Delta}})}} \right] \\
 &= y(x) + \Delta [c_1 k_1 + c_2 k_2] \quad \begin{matrix} p_2=1 \\ a_{2,1}=1 \end{matrix}
 \end{aligned}$$

# Runge-Kutta Methods: 7/22

- Let us examine general second-order Runge-Kutta methods.
- A second-order Runge-Kutta method is:

$$y(x + \Delta) = y(x) + \Delta \left[ c_1 f(x, y) + c_2 \overset{k_2}{f(x + p_2 \Delta, y + a_{2,1} f(x, y) \Delta)} \right]$$

- We wish to expand  $k_2$ . This will need the two-variable Taylor series as follows:  $h^2, k^2$  and higher terms

$$g(x + h, y + k) = g(x, y) + h \frac{\partial g}{\partial x} + k \frac{\partial g}{\partial y} + (\dots)$$

- The following shows the result of expanding  $k_2$ :

$$\begin{aligned} & f(x + p_2 \Delta, y + a_{2,1} f(x, y) \Delta) \\ &= f(x, y) + (p_2 \Delta) \frac{\partial f}{\partial x} + (a_{2,1} f(x, y) \Delta) \frac{\partial f}{\partial y} + (\dots) \end{aligned}$$

$\Delta^2$  and higher terms

# Runge-Kutta Methods: 8/22

- Plugging the expanded  $k_2$  back into  $y(x+\Delta)$  yields the following important result:

$$\begin{aligned} y(x + \Delta) &= y(x) \\ &+ \Delta(c_1 + c_2)f(x, y) \\ &+ \Delta^2 \left[ (c_2 p_2) \frac{\partial f}{\partial x} + (c_2 a_{2,1} f(x, y)) \frac{\partial f}{\partial y} \right] \\ &+ \boxed{(\dots)} \end{aligned}$$

$\Delta^3$  and higher terms

# Runge-Kutta Methods: 9/22

- Consider the function  $y(x+\Delta)$  itself. It can be expanded using Taylor series as follows:

$$y(x + \Delta) = y(x) + \Delta y'(x) + \frac{\Delta^2}{2!} y''(x) + \boxed{(\dots)}$$

$\Delta^3$  and higher terms

- Note that  $y'(x) = f(x,y)$  and  $y''(x)$  can be computed using the chain-rule:

$$\begin{aligned} y''(x) &= \frac{d(y')}{dx} = f'(x, y) \\ &= \frac{\partial f(x, y)}{\partial x} + \frac{\partial f(x, y)}{\partial y} \boxed{\frac{dy}{dx}} \text{ this is } y'(x) = f(x, y) \\ &= \frac{\partial f(x, y)}{\partial x} + \frac{\partial f(x, y)}{\partial y} f(x, y) \end{aligned}$$

# Runge-Kutta Methods: 10/22

- Plugging  $y'(x)$  and the computed  $y''(x)$  back into  $y(x+\Delta)$ , we have the second important result:

$$\begin{aligned}y(x + \Delta) &= y(x) + \Delta f(x, y) + \frac{\Delta^2}{2} y''(x) + \boxed{\text{.....}} \quad \Delta^3 \text{ and higher terms} \\ &= y(x) + \Delta f(x, y) + \Delta^2 \left[ \frac{1}{2} \frac{\partial f}{\partial x} + \frac{1}{2} f(x, y) \frac{\partial f}{\partial y} \right] + \boxed{\text{.....}} \\ &\quad \Delta^3 \text{ and higher terms}\end{aligned}$$

# Runge-Kutta Methods: 1 1/22

- Now compare the two results of  $y(x+\Delta)$  we obtained. They should be equal!

$$\begin{aligned}
 y(x+\Delta) &= y(x) + \Delta \boxed{(c_1 + c_2) f(x, y)} \\
 &+ \Delta^2 \boxed{\left[ (c_2 p_2) \frac{\partial f}{\partial x} + (c_2 a_{2,1} f(x, y)) \frac{\partial f}{\partial y} \right]} \\
 &+ (\dots) \\
 y(x+\Delta) &= y(x) + \Delta \boxed{f(x, y)} \\
 &+ \Delta^2 \boxed{\left[ \frac{1}{2} \frac{\partial f}{\partial x} + \frac{1}{2} f(x, y) \frac{\partial f}{\partial y} \right]} \\
 &+ (\dots)
 \end{aligned}$$

$c_1 + c_2 = 1$  (indicated by a red arrow pointing from the first box to the second box)  
 $c_2 p_2 = 1/2$  (indicated by a blue double-headed arrow between the second and third boxes)  
 $c_2 a_{2,1} = 1/2$  (indicated by a green arrow pointing from the third box to the second box)

# Runge-Kutta Methods: 12/22

- From comparing the coefficients of  $\Delta$  and  $\Delta^2$ , we have the following:

$$\begin{array}{lcl} c_1 + c_2 & = & 1 \\ c_2 p_2 & = & 1/2 \\ c_2 a_{2,1} & = & 1/2 \end{array} \quad \text{or} \quad \begin{array}{lcl} c_1 & = & 1 - c_2 \\ p_2 & = & 1/(2c_2) \\ a_{2,1} & = & 1/(2c_2) \end{array}$$

- Since four unknowns (*i.e.*,  $c_1, c_2, p_2, a_{2,1}$ ) cannot be solved from three equations, one must fix a variable to some value.
- Varying the value of  $c_2$  yields a series of second-order Runge-Kutta method.

# Runge-Kutta Methods: 13/22

- If  $c_2 = 1/2$ , then  $c_1 = 1/2$ ,  $p_2 = 1$  and  $a_{2,1} = 1$ .
- Second-order Runge-Kutta method becomes:

$$\begin{aligned}y(x + \Delta) &= y(x) + \Delta \left[ \frac{1}{2} f(x, y) + \frac{1}{2} f(x + \Delta, y + f(x, y)\Delta) \right] \\ &= y(x) + \frac{\Delta}{2} [f(x, y) + f(x + \Delta, y + f(x, y)\Delta)]\end{aligned}$$

- This is Heun's method! We saw it earlier.

$$\begin{aligned}c_1 &= 1 - c_2 \\ p_2 &= a_{2,1} = \frac{1}{2c_2}\end{aligned}$$

# Runge-Kutta Methods: 14/22

- If  $c_2=1$ , then  $c_1=0$ ,  $p_2 = a_{2,1} = 1/2$ .
- Second order Runge-Kutta method becomes:

$$\begin{aligned}y(x + \Delta) &= y(x) + \Delta k_2 \\ &= y(x) + \Delta f \left( x + \frac{1}{2} \Delta, y + \frac{1}{2} \Delta f(x, y) \right)\end{aligned}$$

- This is referred to as the *midpoint* method.

$$\begin{aligned}c_1 &= 1 - c_2 \\ p_2 &= a_{2,1} = \frac{1}{2c_2}\end{aligned}$$

# Runge-Kutta Methods: 15/22

- If  $c_2 = 2/3$ , then  $c_1 = 1/3$ ,  $p_2 = a_{2,1} = 3/4$ .
- Second-order Runge-Kutta method becomes:

$$y(x + \Delta) = y(x) + \Delta \left[ \frac{1}{3} f(x, y) + \frac{2}{3} f \left( x + \frac{3}{4} \Delta, y + \frac{3}{4} \Delta f(x, y) \right) \right]$$

- This is the *Ralston* method.

$$\begin{aligned} c_1 &= 1 - c_2 \\ p_2 &= a_{2,1} = \frac{1}{2c_2} \end{aligned}$$

# Runge-Kutta Methods: 16/22

- The following shows a possible second-order Runge-Kutta method program.
- The initialization part, however, is for Ralston method.

```
! Initialization  
! [a,b] : input  
! n : intervals
```

```
c2 = 2.0/3.0  
c1 = 1 - c2  
p2 = 1/(2*c2)  
a21 = 1/(2*c2)
```

for Ralston method

```
 $\Delta = (b-a)/n$   
x = a  
DO i = 1, n  
    k1 = f(x,y)  
    k2 = f(x+p2* $\Delta$ , y+a21*k1* $\Delta$ )  
    y = y +  $\Delta$ *(c1*k1 + c2*k2)  
    x = x +  $\Delta$   
END
```

general second-order Runge-Kutta method

# Runge-Kutta Methods: 17/22

- ODE  $y' = 4e^{0.8x} - 0.5y$  with  $y(0) = 2$  has the following exact solution:

$$y = \frac{4}{1.3} \left( e^{0.8x} - e^{-0.5x} \right) + 2e^{-0.5x}$$

- The following uses  $\Delta = 1$  on  $[0, 4]$ :

predictor-corrector

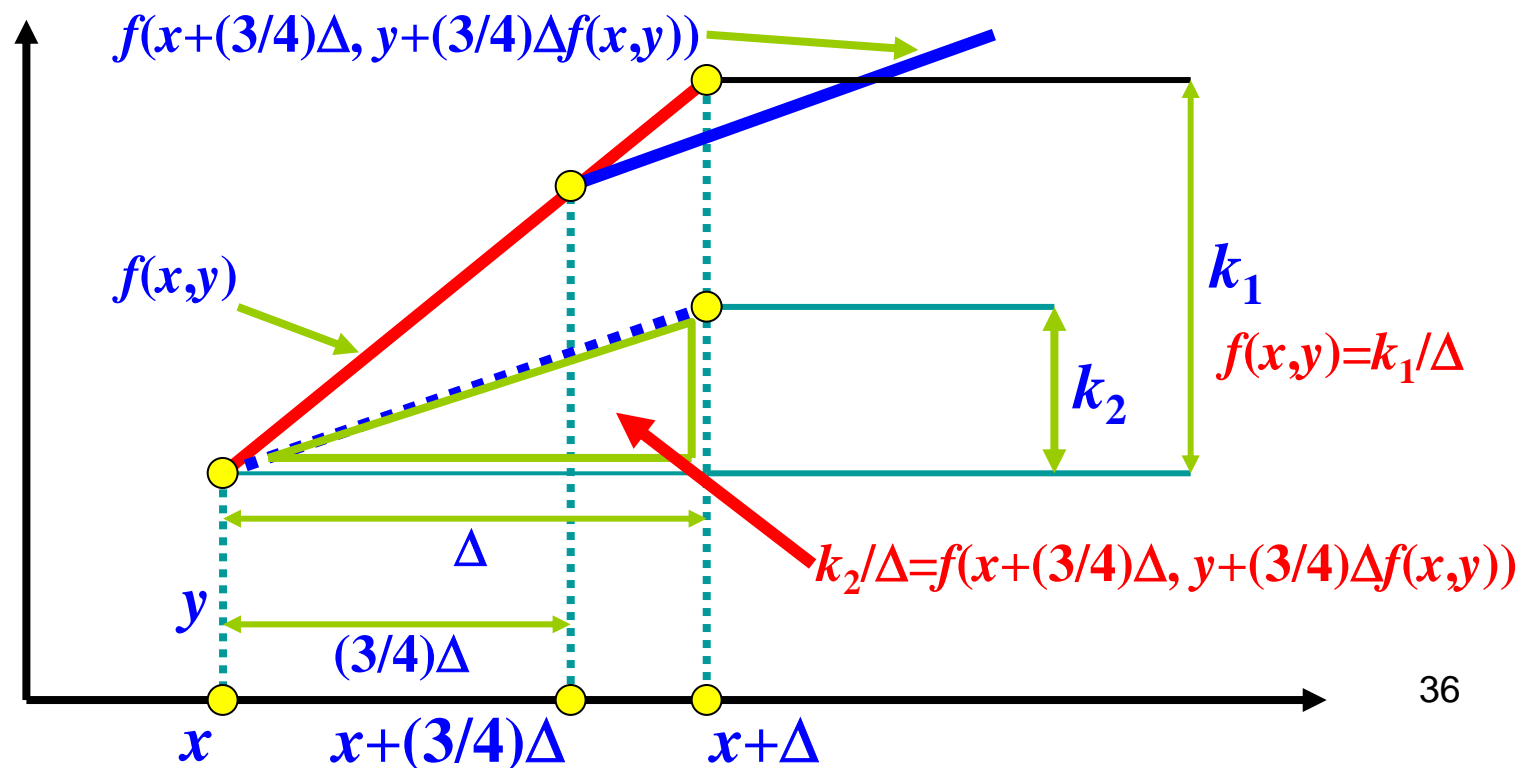
x	Exact y	Euler	%	Heun	%	P-C	%	Ralston	%
0	2	2	0	2	0	2	0	2	0
1	6.19463	5.00000	19.3	6.70108	8.2	6.36053	2.7	6.44232	4.0
2	14.84392	11.40216	23.2	16.31978	9.9	15.30125	3.1	15.58216	5.0
3	33.67717	25.51321	24.2	37.19925	10.5	34.74091	3.2	35.45657	5.3
4	75.33897	56.84931	24.5	83.33777	10.6	77.72971	3.2	79.39618	5.4

6 iterations

# Runge-Kutta Methods: 18/22

- A geometric interpretation of the  $k_i$ 's.
- Take the Ralston method as an example.

$$y(x + \Delta) = y(x) + \Delta \left[ \frac{1}{3} f(x, y) + \frac{2}{3} f\left(x + \frac{3}{4}\Delta, y + \frac{3}{4}\Delta f(x, y)\right) \right]$$



# Runge-Kutta Methods: 19/22

- **Higher order Runge-Kutta methods are more accurate but more time consuming.**
- **They are derived based on the same principle used for second-order methods.**
- **The following is a commonly used third-order Runge-Kutta method:**

$$k_1 = f(x, y)$$

$$k_2 = f\left(x + \frac{1}{2}\Delta, y + \frac{1}{2}\Delta k_1\right)$$

$$k_3 = f(x + \Delta, y - \Delta k_1 + 2\Delta k_2)$$

$$y(x + \Delta) = y(x) + \frac{1}{6}\Delta(k_1 + 4k_2 + k_3)$$

# Runge-Kutta Methods: 20/22

- **The following is the most commonly used fourth-order method, usually referred to as the *classical fourth-order Runge-Kutta method*:**

$$k_1 = f(x, y)$$

$$k_2 = f\left(x + \frac{1}{2}\Delta, y + \frac{1}{2}\Delta k_1\right)$$

$$k_3 = f\left(x + \frac{1}{2}\Delta, y + \frac{1}{2}\Delta k_2\right)$$

$$k_4 = f(x + \Delta, y + \Delta k_3)$$

$$y(x + \Delta) = y(x) + \frac{1}{6}\Delta(k_1 + 2k_2 + 2k_3 + k_4)$$

# Runge-Kutta Methods: 21/22

- **If more accurate results are required, use Butcher's sixth-order Runge-Kutta method:**

$$k_1 = f(x, y)$$

$$k_2 = f\left(x + \frac{1}{4}\Delta, y + \frac{1}{4}\Delta k_1\right)$$

$$k_3 = f\left(x + \frac{1}{4}\Delta, y + \frac{1}{8}\Delta k_1 + \frac{1}{8}\Delta k_2\right)$$

$$k_4 = f\left(x + \frac{1}{2}\Delta, y - \frac{1}{2}\Delta k_2 + \Delta k_3\right)$$

$$k_5 = f\left(x + \frac{3}{4}\Delta, y + \frac{3}{16}\Delta k_1 + \frac{9}{16}\Delta k_4\right)$$

$$k_6 = f\left(x + \Delta, y - \frac{1}{3}\Delta k_1 + \frac{2}{7}\Delta k_2 + \frac{12}{7}\Delta k_3 - \frac{12}{7}\Delta k_4 + \frac{8}{7}\Delta k_5\right)$$

$$y(x + \Delta) = y(x) + \frac{1}{90}\Delta(7k_1 + 32k_3 + 12k_4 + 32k_5 + 7k_6)$$

# Runge-Kutta Methods: 22/22

- ODE  $y' = 4e^{0.8x} - 0.5y$  with  $y(0) = 2$  has the following exact solution:

$$y = \frac{4}{1.3} \left( e^{0.8x} - e^{-0.5x} \right) + 2e^{-0.5x}$$

- The following uses  $\Delta = 1$  on  $[0, 4]$ :

x	Exact y	Ralston	%	3 <sup>rd</sup>	%	4 <sup>th</sup>	%	6 <sup>th</sup>	%
0	2	2	0	2	0	2	0	2	0
1	6.19463	6.44232	4.0	6.17568	0.31	6.20104	0.10	6.19469	0.00
2	14.84392	15.58216	5.0	14.78616	0.39	14.86248	0.13	14.84410	0.00
3	33.67717	35.45657	5.3	33.53672	0.42	33.72135	0.13	33.67760	0.00
4	75.33897	79.39618	5.4	75.01767	0.43	75.43918	0.13	75.33994	0.00

**The End**