

Ecuaciones Diferenciales Ordinarias

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Ecuaciones Diferenciales Ordinarias (EDO)



Ecuaciones que contienen una o más derivadas de una función que depende solamente de una variable. Estas ecuaciones sirven para modelar fenómenos en física, ingeniería, economía, etc.

Modelos de evolución población o PBI:

$$\frac{d}{dt}p(t) = rp(t)$$

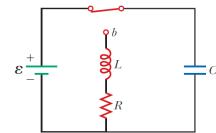
$$\frac{d}{dt}PBI(t) = rPBI(t)$$

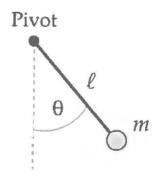
Circuitos eléctricos:

$$L\frac{d^2Q}{dt^2} + R\frac{dQ}{dt} + \frac{Q}{C} = 0$$

Péndulo simple:

$$\frac{d^2\theta}{dt^2} = -\frac{g}{l}Sen(\theta)$$





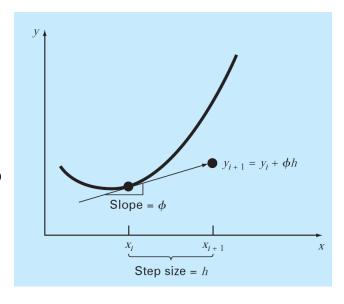
Clasificación de Métodos para EDO's



Métodos de un solo paso Runge-Kutta

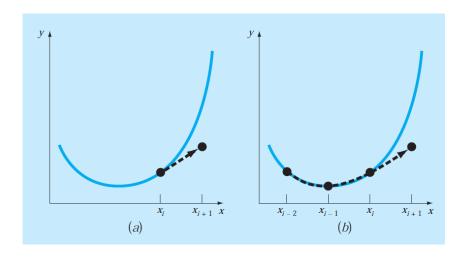
$$\frac{dy}{dx} = f(x, y) \quad y_{i+1} = y_i + \phi h$$

- Euler.
- Heun.
- Punto medio.
- RK2,RK4,RK5,RK6



Métodos multipasos

Se utiliza información precedente para estimar y_{t+1} .



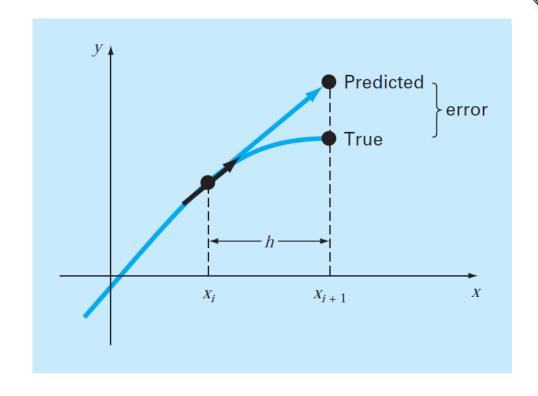
- Heun modificado.
- Adam-Bashforth.
- Adams-Multon.

$$\frac{dy}{dx} = f(x, y)$$

Tomamos a ϕ como la primera derivada

$$\phi = f(x_i, y_i)$$

$$y_{i+1} = y_i + f(x_i, y_i)h$$



 $f(x_i, y_i)$ corresponde ala derivada en el punto inicial.

Nota: f(x, y) depende de x e y y puede ser no lineal en general.

$y_{i+1} = y_i + f(x_i, y_i)h$



Expanción de Taylor

$$y_{i+1} = y_i + y'_i h + \frac{y''_i}{2!} h^2 + \dots + \frac{y_i^{(n)}}{n!} h^n + R_n$$

$$y_{i+1} = y_i + f(x_i, y_i)h + \frac{f'(x_i, y_i)}{2!}h^2 + \dots + \frac{f^{(n-1)}(x_i, y_i)}{n!}h^n + O(h^{n+1})$$

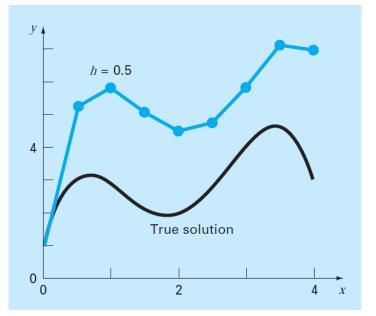
$$R_n = \frac{y^{(n+1)}(\xi)}{(n+1)!} h^{n+1} \qquad E_a = \frac{f'(x_i, y_i)}{2!} h^2$$

Ejemplo

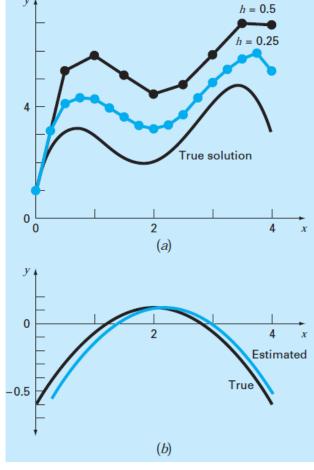
$$\frac{dy}{dx} = -2x^3 + 12x^2 - 20x + 8.5$$

$$y = -0.5x^4 + 4x^3 - 10x^2 + 8.5x + 1$$

x	Y true	Y Euler	Local
0.0	1.00000	1.00000	
0.5	3.21875	5.25000	-63.1
1.0	3.00000	5.87500	-28.0
1.5	2.21875	5.12500	-1.41
2.0	2.00000	4.50000	20.5
2.5	2.71875	4.75000	17.3
3.0	4.00000	5.87500	4.0
3.5	4.71875	7.12500	-11.3
4.0	3.0000	7.00000	-53.0









```
'set integration range
xi = 0
xf = 4
'initialize variables
x = xi
y = 1
'set step size and determine
'number of calculation steps
dx = 0.5
nc = (xf - xi)/dx
'output initial condition
PRINT x. y
'loop to implement Euler's method
'and display results
DOFOR i = 1, nc
  dydx = -2x^3 + 12x^2 - 20x + 8.5
  y = y + dydx \cdot dx
 x = x + dx
 PRINT x, y
END DO
```

```
Assign values for
y = initial value dependent variable
xi = initial value independent variable
xf = final value independent variable
dx = calculation step size
xout = output interval
x = xi
m = 0
Xp_m = X
yp_m = y
  xend = x + xout
 IF (xend > xf) THEN xend = xf
 h = dx
 CALL Integrator (x, y, h, xend)
  m = m + 1
  XD_m = X
 yp_m = y
 IF (x \ge xf) EXIT
END DO
DISPLAY RESULTS
END
```

```
SUB Integrator (x, y, h, xend)
IF (xend - x < h) THEN h = xend - x
    CALL Euler (x, y, h, ynew)
    y = ynew
    IF (x \ge xend) EXIT
  FND DO
 END SUB
 SUB Euler (x, y, h, ynew)
  CALL Derivs(x, y, dydx)
   ynew = y + dydx * h
  x = x + h
 END SUB
 SUB Derivs (x, y, dydx)
  dydx = \dots
 END SUB
```

La idea es mejorar la aproximación de la derivada

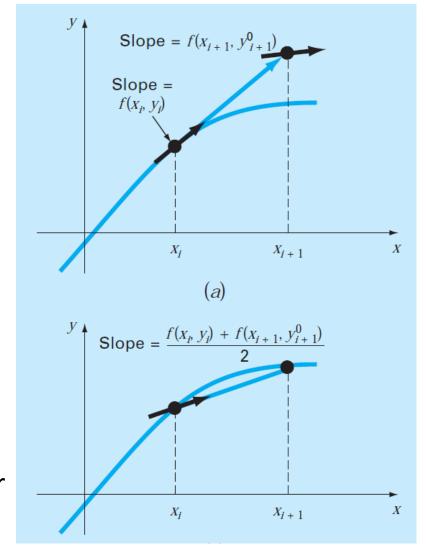
$$y'_i = f(x_i, y_i)$$
 $y'_{i+1} = f(x_{i+1}, y^0_{i+1})$

$$\bar{y}' = \frac{y'_i + y'_{i+1}}{2} = \frac{f(x_i, y_i) + f(x_{i+1}, y^0_{i+1})}{2}$$

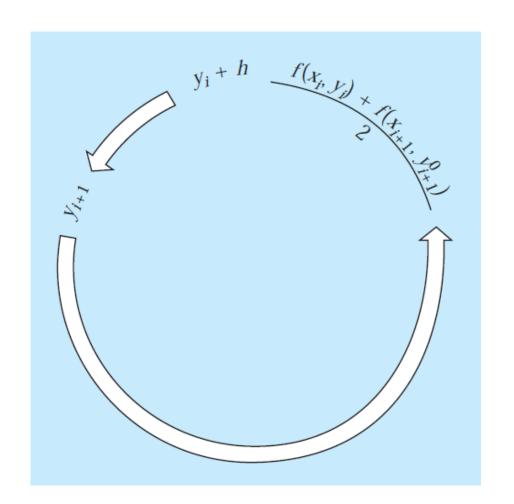
$$y_{i+1}^0 = y_i + f(x_i, y_i)h$$

Predictor

$$y_{i+1} = y_i + \frac{f(x_i, y_i) + f(x_{i+1}, y_{i+1}^0)}{2}h$$
 Corrector







$$y_{i+1} = y_i + \frac{f(x_i, y_i) + f(x_{i+1}, y_{i+1}^0)}{2}h$$

Criterio de convergencia

$$|\varepsilon_a| = \left| \frac{y_{i+1}^j - y_{i+1}^{j-1}}{y_{i+1}^j} \right| 100\%$$



Ejemplo: Obtener $y_1 = y(1)$ Integrar la ecuación con un paso h = 1.

$$y' = 4e^{0.8x} - 0.5y \qquad x = 0, y = 2$$
$$y = \frac{4}{1.3}(e^{0.8x} - e^{-0.5x}) + 2e^{-0.5x}$$

Valor real y(1) = 6.1946314

Calculamos $y_1(i = 0)$, necesitamos

$$f(x_0, y_0) = y'_0 = 4e^0 - 0.5(2) = 3$$

 $y_1^0 = 2 + 3(1) = 5$

Método

$$y_{i+1}^{0} = y_i + f(x_i, y_i)h$$

$$y_{i+1} = y_i + \frac{f(x_i, y_i) + f(x_{i+1}, y_{i+1}^{0})}{2}h$$

$$y_1' = f(x_1, y_1^0) = 4e^{0.8(1)} - 0.5(5) = 6.402164$$

$$y_1 = 2 + 4.701082(1) = 6.701082$$



La aproximación mejora $y_1 \rightarrow y_1^0$

$$y_1 \rightarrow y_1^0$$

Mejora aún más

$$y_1 \rightarrow y_1^0$$

$$y_1 = 2 + \frac{\left[3 + 4e^{0.8(1)} - 0.5(6.701082)\right]}{2}1 = 6.275811$$

$$y_1 = 2 + \frac{\left[3 + 4e^{0.8(1)} - 0.5(6.701082)\right]}{2}1 = 6.275811$$
 $y_1 = 2 + \frac{\left[3 + 4e^{0.8(1)} - 0.5(6.275811)\right]}{2}1 = 6.382129$

		Iterations of Heun's Method				
		1		15		
x	y true	y Heun	lε _t l (%)	y Heun	Ιε _τ Ι (%)	
0	2.0000000	2.0000000	0.00	2.0000000	0.00	
1	6.1946314	6.7010819	8.18	6.3608655	2.68	
2	14.8439219	16.3197819	9.94	15.3022367	3.09	
3	33.6771718	37.1992489	10.46	34.7432761	3.17	
4	75.3389626	83.3377674	10.62	77.7350962	3.18	



En el caso que
$$y' = f(x) \Rightarrow$$

$$y_{i+1} = y_i + \frac{f(x_i) + f(x_{i+1})}{2}h$$

$$\frac{dy}{dx} = f(x) \implies \int_{y_i}^{y_{i+1}} dy = \int_{x_i}^{x_{i+1}} f(x) dx$$

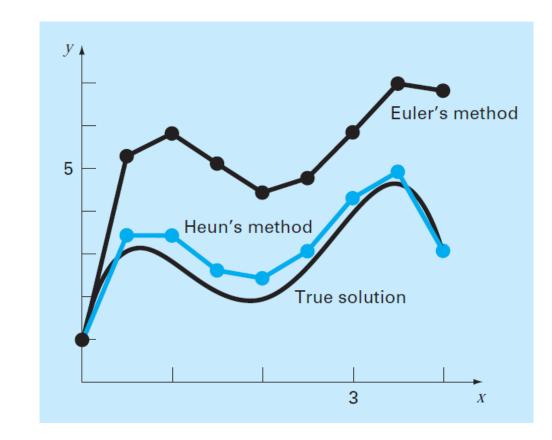
$$y_{i+1} = y_i + \int_{x_i}^{x_{i+1}} f(x) dx$$

$$\int_{x_i}^{x_{i+1}} f(x) \, dx \cong \frac{f(x_i) + f(x_{i+1})}{2} h$$

Regla del trapecio

$$y_{i+1} = y_i + \frac{f(x_i) + f(x_{i+1})}{2}h$$
 $E_t = -\frac{f''(\xi)}{12}h^3$

$$E_t = -\frac{f''(\xi)}{12}h^3$$



Método punto medio

La idea es utilizar el método de Euler evaluando la derivada en $y_{i+1/2}$.

$$y_{i+1/2} = y_i + f(x_i, y_i) \frac{h}{2}$$
 $y'_{i+1/2} = f(x_{i+1/2}, y_{i+1/2})$

$$y_{i+1} = y_i + f(x_{i+1/2}, y_{i+1/2})h$$

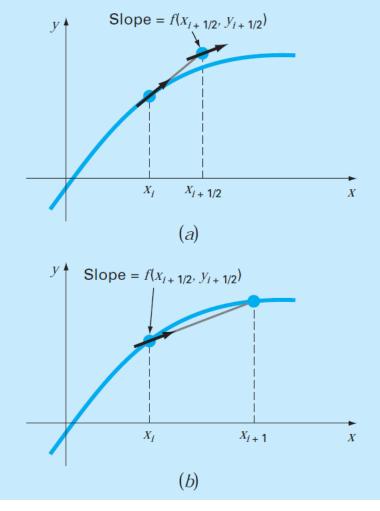
$$\frac{dy}{dx} = f(x) \implies \int_{y_i}^{y_{i+1}} dy = \int_{x_i}^{x_{i+1}} f(x) dx$$

$$y_{i+1} = y_i + \int_{x_i}^{x_{i+1}} f(x) dx$$

$$\int_{x_i}^{x_{i+1}} f(x) dx \cong hf(x_{i+1/2})$$

$$\int_{a}^{b} f(x) dx \cong (b-a) f(x_1) \qquad x_1 = \text{punto medio}$$









(a) Simple Heun without Corrector

```
SUB Heun (x, y, h, ynew)

CALL Derivs (x, y, dy1dx)

ye = y + dy1dx \cdot h

CALL Derivs (x + h, ye, dy2dx)

Slope = (dy1dx + dy2dx)/2

ynew = y + Slope \cdot h

x = x + h

END SUB
```

(b) Midpoint Method

```
SUB Midpoint (x, y, h, ynew)

CALL Derivs(x, y, dydx)

ym = y + dydx \cdot h/2

CALL Derivs (x + h/2, ym, dymdx)

ynew = y + dymdx \cdot h

x = x + h

END SUB
```

(c) Heun with Corrector

```
SUB HeunIter (x, y, h, ynew)
 es = 0.01
 maxit = 20
 CALL Derivs(x, y, dy1dx)
 ye = y + dy1dx \cdot h
 iter = 0
  DO
   yeold = ye
   CALL Derivs(x + h, ye, dy2dx)
   slope = (dy1dx + dy2dx)/2
   ye = y + slope \cdot h
   iter = iter + 1
   IF (ea ≤ es OR iter > maxit) EXIT
  FND DO
  ynew = ye
  x = x + h
END SUB
```

Métodos Runge-Kutta



Métodos que alcanzan la precisión de un desarrollo de Taylor sin calcular derivadas de orden superior

$$y_{i+1} = y_i + \phi(x_i, y_i, h)h$$

$$\phi = a_1k_1 + a_2k_2 + \cdots + a_nk_n$$

```
k_1 = f(x_i, y_i)
k_2 = f(x_i + p_1 h, y_i + q_{11} k_1 h)
k_3 = f(x_i + p_2 h, y_i + q_{21} k_1 h + q_{22} k_2 h)
\cdot
```

Nota:

- Los valores p y q son constantes.
- El valor k_n se obtiene por recurrencia.
- Eficiente para cálculos.
- El método RK para n=1 es el método de Euler.

Método RK2

$$y_{i+1} = y_i + (a_1k_1 + a_2k_2)h$$

$$k_1 = f(x_i, y_i)$$

$$k_2 = f(x_i + p_1 h, y_i + q_{11} k_1 h)$$

$$f(x_i + p_1 h, y_i + q_{11} k_1 h) = f(x_i, y_i) + p_1 h \frac{\partial f}{\partial x}$$

$$+ q_{11}k_1h\frac{\partial f}{\partial y} + O(h^2)$$

$$y_{i+1} = y_i + a_1 h f(x_i, y_i) + a_2 h f(x_i, y_i) + a_2 p_1 h^2 \frac{\partial f}{\partial x} + a_2 q_{11} h^2 f(x_i, y_i) \frac{\partial f}{\partial y} + O(h^3)$$

$$y_{i+1} = y_i + f(x_i, y_i)h + \frac{f'(x_i, y_i)}{2!}h^2$$

$$f'(x_i, y_i) = \frac{\partial f(x, y)}{\partial x} + \frac{\partial f(x, y)}{\partial y} \frac{dy}{dx}$$

$$y_{i+1} = y_i + f(x_i, y_i)h + \left(\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y}\frac{dy}{dx}\right)\frac{h^2}{2!}$$

$$y_{i+1} = y_i + [a_1 f(x_i, y_i) + a_2 f(x_i, y_i)]h$$
$$+ \left[a_2 p_1 \frac{\partial f}{\partial x} + a_2 q_{11} f(x_i, y_i) \frac{\partial f}{\partial y} \right] h^2 + O(h^3)$$

$$a_1 + a_2 = 1$$

$$a_2 p_1 = \frac{1}{2}$$

$$a_2q_{11}=\frac{1}{2}$$

$$a_1 = 1 - a_2$$

$$p_1 = q_{11} = \frac{1}{2a_2}$$

Método RK2

Método de Heund ($a_2=1/2$)

$$y_{i+1} = y_i + \left(\frac{1}{2}k_1 + \frac{1}{2}k_2\right)h$$

$$k_1 = f(x_i, y_i)$$

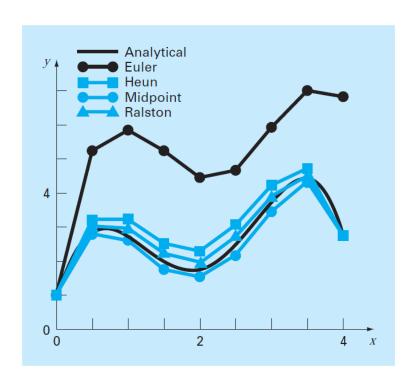
 $k_2 = f(x_i + h, y_i + k_1 h)$

Método punto medio $(a_2=1)$

$$y_{i+1} = y_i + k_2 h$$

$$k_1 = f(x_i, y_i)$$

 $k_2 = f\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k_1h\right)$





Método Ralston ($a_2=2/3$)

$$y_{i+1} = y_i + \left(\frac{1}{3}k_1 + \frac{2}{3}k_2\right)h$$

$$k_1 = f(x_i, y_i)$$

 $k_2 = f\left(x_i + \frac{3}{4}h, y_i + \frac{3}{4}k_1h\right)$

Método RK3

$y_{i+1} = y_i + \frac{1}{6} (k_1 + 4k_2 + k_3)h$

$$k_1 = f(x_i, y_i)$$

$$k_2 = f\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k_1h\right)$$

$$k_3 = f(x_i + h, y_i - k_1h + 2k_2h)$$

Se obtienen seis ecuaciones y ocho incognitas

Método RK4



$$y_{i+1} = y_i + \frac{1}{6}(k_1 + 2k_2 + 2k_3 + k_4)h$$

$$k_1 = f(x_i, y_i)$$

$$k_2 = f\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k_1h\right)$$

$$k_3 = f\left(x_i + \frac{1}{2}h, y_i + \frac{1}{2}k_2h\right)$$

$$k_4 = f(x_i + h, y_i + k_3h)$$

Método RK5 (Butcher)

$$y_{i+1} = y_i + \frac{1}{90}(7k_1 + 32k_3 + 12k_4 + 32k_5 + 7k_6)h$$

$$k_{1} = f(x_{i}, y_{i})$$

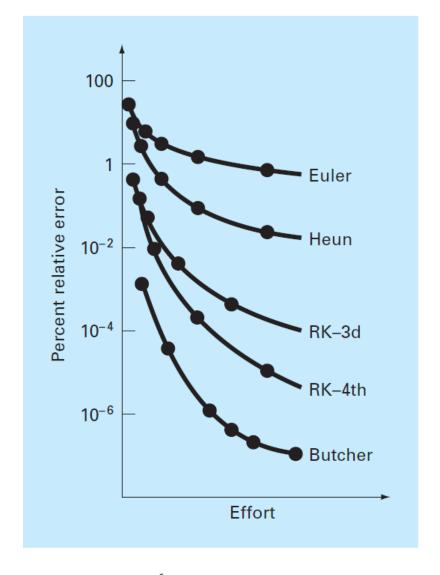
$$k_{2} = f\left(x_{i} + \frac{1}{4}h, y_{i} + \frac{1}{4}k_{1}h\right)$$

$$k_{3} = f\left(x_{i} + \frac{1}{4}h, y_{i} + \frac{1}{8}k_{1}h + \frac{1}{8}k_{2}h\right)$$

$$k_{4} = f\left(x_{i} + \frac{1}{2}h, y_{i} - \frac{1}{2}k_{2}h + k_{3}h\right)$$

$$k_{5} = f\left(x_{i} + \frac{3}{4}h, y_{i} + \frac{3}{16}k_{1}h + \frac{9}{16}k_{4}h\right)$$

$$k_{6} = f\left(x_{i} + h, y_{i} - \frac{3}{7}k_{1}h + \frac{2}{7}k_{2}h + \frac{12}{7}k_{3}h - \frac{12}{7}k_{4}h + \frac{8}{7}k_{5}h\right)$$



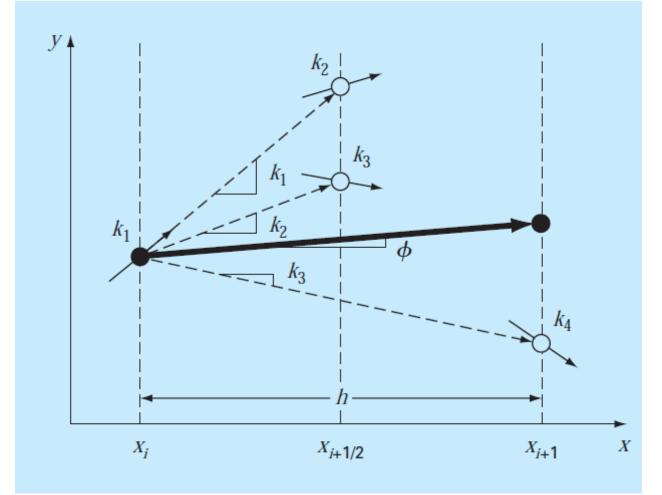
Effort =
$$n_f \frac{b-a}{h}$$

n_f :numero de evaluaciones

Pseudocódigo RK4



```
SUB RK4 (x, y, h, ynew)
 CALL Derivs(x, y, k1)
 ym = y + k1 \cdot h/2
 CALL Derivs(x + h/2, ym, k2)
 ym = y + k2 \cdot h/2
 CALL Derivs(x + h/2, ym, k3)
 ye = y + k3 \cdot h
 CALL Derivs(x + h, ye, k4)
 slope = (k1 + 2(k2 + k3) + k4)/6
 ynew = y + slope \cdot h
 x = x + h
END SUB
```



Aplicación



Resolver la siguiente ecuación diferencial

$$\frac{dy}{dt} = yt^3 - 1.5y$$

- Analíticamente.
- Método de Euler.
- Método de Heund.
- Método de Ralston.
- RK4.

Sistemas de ecuaciones

$$\frac{dy_1}{dx} = f_1(x, y_1, y_2, \dots, y_n)$$

$$\frac{dy_2}{dx} = f_2(x, y_1, y_2, \dots, y_n)$$

$$\frac{dy_n}{dx} = f_n(x, y_1, y_2, \dots, y_n)$$

Se necesitan n condiciones iniciales

Definimos
$$\mathbf{r} = \begin{pmatrix} y_1(x) \\ \vdots \\ y_n(x) \end{pmatrix}$$
 y $\mathbf{F} = \begin{pmatrix} f_1(x, \mathbf{r}) \\ \vdots \\ f_n(x, \mathbf{r}) \end{pmatrix}$

RK4

$$r_{i+1} = r_i + \frac{1}{6} (K_1 + 2K_2 + 2K_3 + K_4)h$$

$$K_1 = F(x, r)$$

$$\mathbf{K_2} = \mathbf{F}\left(x + \frac{1}{2}h, \mathbf{r} + \frac{1}{2}\mathbf{K_1}h\right)$$

$$\mathbf{K_3} = \mathbf{F}\left(x + \frac{1}{2}h, \mathbf{r} + \frac{1}{2}\mathbf{K_2}h\right) \qquad \mathbf{K_4} = \mathbf{F}\left(x + h, \mathbf{r} + \frac{1}{2}\mathbf{K_3}h\right)$$

$$\boldsymbol{K_4} = \boldsymbol{F}\left(x + h, \boldsymbol{r} + \frac{1}{2}\boldsymbol{K_3} h\right)$$

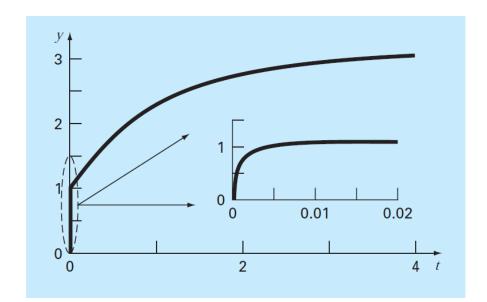


Stiffness:



Son problemas donde la solución tiene periodos de evolución lentos y rapidos.

$$\frac{dy}{dt} = -1000y + 3000 - 2000e^{-t} \qquad y = 3 - 0.998e^{-1000t} - 2.002e^{-t}$$



$$y = 3 - 0.998e^{-1000t} - 2.002e^{-t}$$

Solución parte homogénea

$$\frac{dy}{dt} = -ay \qquad y = y_0 e^{-at}$$

$$y_{i+1} = y_i + \frac{iy_i}{dt}h \qquad \text{Estabilidad} \qquad |1 - ah| < 1$$

$$y_{i+1} = y_i - ay_i h$$

$$y_{i+1} = y_i(1 - ah)$$

Si
$$h>2/a$$
, entonces $y_i\to\infty$ cuando $i\to\infty$ Condición de estabilidad $h<0.002$

Euler implícito



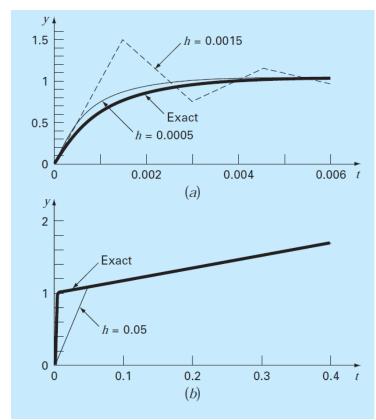
$$y_{i+1} = y_i + \frac{dy_{i+1}}{dt}h$$
 $y_{i+1} = y_i + (-1000y_{i+1} + 3000 - 2000e^{-t_{i+1}})h$

$$y_{i+1} = y_i - ay_{i+1}h$$
 $y_{i+1} = \frac{y_i + 3000h - 2000he^{-t_{i+1}}}{1 + 1000h}$

$$y_{i+1} = \frac{y_i}{1 + ah}$$

Incondiconalmente stable

$$\frac{1}{|1+ah|} < 1$$



Euler implícito (sistema ODE)



$$\frac{dy_1}{dt} = -5y_1 + 3y_2$$

$$\frac{dy_2}{dt} = 100y_1 - 301y_2$$

$$y_1 = 52.96e^{-3.9899t} - 0.67e^{-302.0101t}$$

$$y_2 = 17.83e^{-3.9899t} + 65.99e^{-302.0101t}$$

$$y_{1,i+1} = y_{1,i} + (-5y_{1,i+1} + 3y_{2,i+1})h$$

$$y_{2,i+1} = y_{2,i} + (100y_{1,i+1} - 301y_{2,i+1})h$$

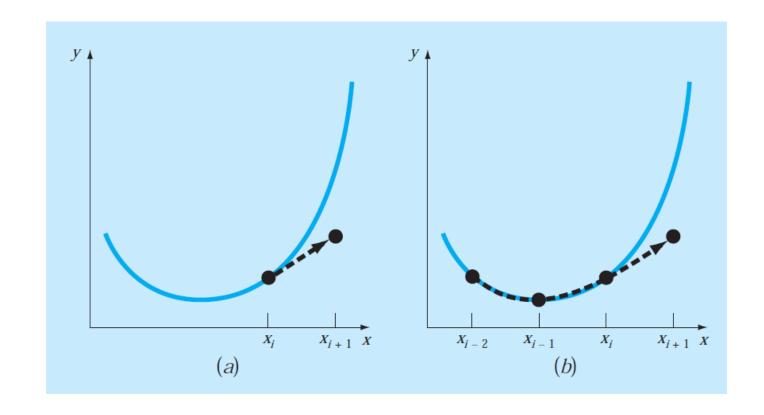
$$(1+5h)y_{1,i+1} - 3hy_{2,i+1} = y_{1,i}$$
$$-100hy_{1,i+1} + (1+301h)y_{2,i+1} = y_{2,i}$$

← Sistema Lineal a resolver

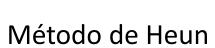
Métodos multipasos



Le objetivo es utilizar información precedente sobre la curvatura de la solución.



Método de Heun modificado



$$y_{i+1}^0 = y_i + f(x_i, y_i)h$$
 $O(h^2)$

$$y_{i+1} = y_i + \frac{f(x_i, y_i) + f(x_{i+1}, y_{i+1}^0)}{2} h \quad O(h^3)$$

Modificación

$$y_{i+1}^0 = y_{i-1} + f(x_i, y_i)2h$$
 $O(h^3)$

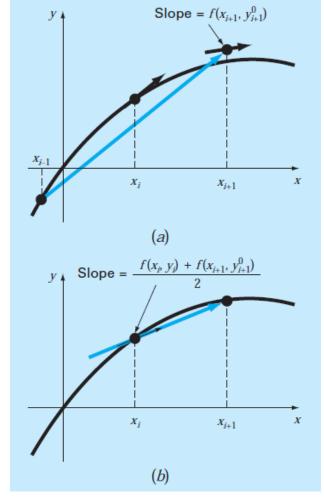
$$y_{i+1}^{0} = y_{i-1}^{m} + f(x_i, y_i^{m}) 2h$$

$$y_{i+1}^{j} = y_i^{m} + \frac{f(x_i, y_i^{m}) + f(x_{i+1}, y_{i+1}^{j-1})}{2} h$$
(for $j = 1, 2, ..., m$)

Corrector con proceso iterativo

$$|\varepsilon_a| = \left| \frac{y_{i+1}^j - y_{i+1}^{j-1}}{y_{i+1}^j} \right| 100\%$$
 Criterio de convergencia





Ejemplo método de Heun modificado

Resolver para h = 1.0 asumiendo que

$$y(x=-1)=0.3929953, y(x=0)=2.0$$

$$y' = 4e^{0.8x} - 0.5y$$
 $x = 0 \text{ to } x = 4$

$$y_{i+1}^{0} = y_{i-1}^{m} + f(x_i, y_i^{m}) 2h$$

$$y_{i+1}^{j} = y_i^{m} + \frac{f(x_i, y_i^{m}) + f(x_{i+1}, y_{i+1}^{j-1})}{2} h$$
(for $j = 1, 2, ..., m$)

$$y_1^0 = -0.3929953 + [4e^{0.8(0)} - 0.5(2)]2 = 5.60700$$

$$y_1^1 = 2 + \frac{4e^{0.8(0)} - 0.5(2) + 4e^{0.8(1)} - 0.5(5.607005)}{2}1 = 6.549331$$

$$y_1^2 = 2 + \frac{3 + 4e^{0.8(1)} - 0.5(6.549331)}{2}1 = 6.313749$$

$$|\varepsilon_a| = \left| \frac{6.313749 - 6.549331}{6.313749} \right| 100\% = 3.7\%$$

$$y_2^0 = 2 + [4e^{0.8(1)} - 0.5(6.360865)]2 = 13.44346$$

Error en las formulas predictor-corrector



Deseamos resolver $\frac{dy}{dx} = f(x, y)$

$$\frac{dy}{dx} = f(x, y)$$

$$\int_{y_{l}}^{y_{l+1}} dy = \int_{x_{l}}^{x_{l+1}} f(x, y) dx$$

$$y_{i+1} = y_i + \int_{x_i}^{x_{i+1}} f(x, y) dx$$

Método del trapecio $h = X_{i+1} - X_i$

$$\int_{x_i}^{x_{i+1}} f(x, y) \ dx = \frac{f(x_i, y_i) + f(x_{i+1}, y_{i+1})}{2} h$$

$$y_{i+1} = y_i + \frac{f(x_i, y_i) + f(x_{i+1}, y_{i+1})}{2}h$$

Error del corrector

$$E_c = -\frac{1}{12}h^3 y^{(3)}(\xi_c) = -\frac{1}{12}h^3 f''(\xi_c)$$

Error en las formulas predictor-corrector



Deseamos resolver
$$\frac{dy}{dx} = f(x, y)$$

$$\int_{y_l}^{y_{l+1}} dy = \int_{x_l}^{x_{l+1}} f(x, y) \ dx$$

$$y_{i+1} = y_i + \int_{x_i}^{x_{i+1}} f(x, y) dx$$

Método del punto medio

$$\int_{x_{i-1}}^{x_{i+1}} f(x, y) \ dx = 2h f(x_i, y_i)$$

$$y_{i+1} = y_{i-1} + 2hf(x_i, y_i)$$

Error del predictor

$$E_p = \frac{1}{3}h^3 y^{(3)}(\xi_p) = \frac{1}{3}h^3 f''(\xi_p)$$

Formula de Adam-Bashforth



$$y_{i+1} = y_i + f_i h + \frac{f_i'}{2} h^2 + \frac{f_i''}{6} h^3 + \cdots$$

$$y_{i+1} = y_i + h \left(f_i + \frac{h}{2} f_i' + \frac{h^2}{3!} f_i'' + \cdots \right)$$

$$f'_i = \frac{f_i - f_{i-1}}{h} + \frac{f''_i}{2}h + O(h^2)$$

$$y_{i+1} = y_i + h \left\{ f_i + \frac{h}{2} \left[\frac{f_i - f_{i-1}}{h} + \frac{f_i''}{2} h + O(h^2) \right] + \frac{h^2}{6} f_i'' + \cdots \right\}$$

$$y_{i+1} = y_i + h\left(\frac{3}{2}f_i - \frac{1}{2}f_{i-1}\right) + \frac{5}{12}h^3f_i'' + O(h^4)$$

Formula de segundo orden

Formulas de Adam-Bashforth



$$y_{i+1} = y_i + h \sum_{k=0}^{n-1} \beta_k f_{i-k} + O(h^{n+1})$$

Order	βο	$oldsymbol{eta}_1$	eta_2	β3	β4	β5	Local Truncation Error
1	1						$\frac{1}{2}h^2f'(\xi)$
2	3/2	-1/2					$\frac{5}{12}h^3f''(\xi)$
3	23/12	-16/12	5/12				$\frac{9}{24}h^4f^{(3)}(\xi)$
4	55/24	-59/24	37/24	-9/24			$\frac{251}{720}h^5f^{(4)}(\xi)$
5	1901/720	-2774/720	2616/720	-1274/720	251/720		$\frac{475}{1440}h^6f^{(5)}(\xi)$
6	4277/720	-7923/720	9982/720	-7298/720	2877/720	-475/720	$\frac{19,087}{60,480}h^7f^{(6)}(\xi)$

Formula de Adams-Multon



Aproximación de Taylor alrededor de y_{i+1}

$$y_i = y_{i+1} - f_{i+1}h + \frac{f'_{i+1}}{2}h^2 - \frac{f''_{i+1}}{3!}h^3 + \cdots$$

$$y_{i+1} = y_i + h \left(f_{i+1} - \frac{h}{2} f'_{i+1} + \frac{h^2}{6} f''_{i+1} + \cdots \right)$$

$$f'_{i+1} = \frac{f_{i+1} - f_i}{h} + \frac{f''_{i+1}}{2}h + O(h^2)$$

$$y_{i+1} = y_i + h\left(\frac{1}{2}f_{i+1} + \frac{1}{2}f_i\right) - \frac{1}{12}h^3f_{i+1}'' - O(h^4)$$

$$y_{i+1} = y_i + h \sum_{k=0}^{n-1} \beta_k f_{i+1-k} + O(h^{n+1})$$

Formula de Adams-Multon



Order	βο	$oldsymbol{eta_1}$	eta_2	β3	β4	eta_5	Local Truncation Error
2	1/2	1/2					$-\frac{1}{12}h^3f''(\xi)$
3	5/12	8/12	-1/12				$-\frac{1}{24}h^4f^{(3)}(\xi)$
4	9/24	19/24	-5/24	1/24			$-\frac{19}{720}h^5f^{(4)}(\xi)$
5	251/720	646/720	-264/720	106/720	-19/720		$-\frac{27}{1440}h^6f^{(5)}(\xi)$
6	475/1440	1427/1440	-798/1440	482/1440	-173/1440	27/1440	$-\frac{863}{60,480}h^7f^{(6)}(\xi)$

Ejercicio

- Resolver con el método
 Euler implícito y explícito para
- $x_1(0)=x_2(0)=1$
- t [0,0.2]
- h=0.05



$$\frac{dx_1}{dt} = 999x_1 + 1999x_2$$

$$\frac{dx_2}{dt} = -1000x_1 - 2000x_2$$

Ejercicio



- y(1.5)=5.222138
- y (2.0)=4.143883
- t [2,3]
- h=0.5
- $\epsilon_{\rm S}$ =0.01%



$$\frac{dy}{dt} = -0.5y + e^{-t}$$