

Solving ODE's with the theta or the trapezoidal rule

The theta method uses a parameter $0 < \theta < 1$
for solve ODE's. Special values of θ are:

$\theta = 0$: Backward Euler
 $\theta = 1/2$: Trapezoid method
 $\theta = 1$: Forward Euler

DSolRes

θ rule + Newton Raphson ODE Solver

```
DSolRes (Ds, toe, xi,  $\theta$ , N) :=
:= [ m := length(xi) r := [1..m] c := r hId := eval(h.identity(m)) ]
[ X := xi_r T := eval([min(toe)]) dt := eval( $\frac{\max(toe) - T_1}{N}$ ) J := 0 ]
for k ∈ [1..N]
  [ to := T_k xo := col(X, k) ]
  [ t := to + dt x := xo + dt.D(to, xo) ]
  res(x) := x - xo - (t - to).( $\theta$ .feval(Ds, to, xo) + (1 -  $\theta$ ).feval(Ds, t, x))
  R := res(x)
  for iter ∈ [1..MaxI]
    if normi(R) ≤  $\epsilon_y$ 
      break
    else
      J_r_c := eval( $\frac{1}{h} \cdot \left( \text{res}(x + \text{col}(hId, c))_r - R_r \right)$ )
       $\Delta x$  := eval( $-J^{-1} \cdot R$ )
      if normi( $\Delta x$ ) ≤  $\epsilon_x$ 
        break
      else
        [ x := eval(x +  $\Delta x$ ) R := eval(res(x)) ]
  [ T_{k+1} := t X_{c k+1} := x_c ]
augment(T, X^T)
```

The same, but using alg lib solver: faster.

θ rule + Newton Raphson ODE Solver

```
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for k ∈ [1..N]
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  res(x) := x - xo - (t - to).( $\theta$ .feval(Ds, to, xo) + (1 -  $\theta$ ).feval(Ds, t, x))
  [ T_{k+1} := t X_{c k+1} := al_nleqsolve(x, res)_c ]
augment(T, X^T)
```

Utils

```
feval (f#, x#):=|str2num(concat(num2str(f#), "(", num2str(x#), ")"))

feval (f#, x#, y#):=|str2num(concat(num2str(f#), "(", num2str(x#), ",", num2str(y#), ")"))

MCols (M):=| C:=0
for k ∈ [1..cols(M)]
  C_1 k := col(M, k)
C
Defaults
h:=10-12      εy:=10-14      εx:=10-14
0:=10-7      MaxI:=100
```

DSolRes Example

Example: Exact linear ODE $x'' = A' \cdot x + A \cdot x' + B'$

```
Given
to:=0      te:=3      xo:=[ 0 1]      N:=15

A(t):=0.1.t      B(t):=e-0.2.t.t
```

```
let
  A'(t):=d/dt A(t)=1/10      IA(t):=∫tot A(τ2) dτ2
  B'(t):=d/dt B(t)= $\frac{5-t}{5 \cdot \sqrt[5]{e}}$ 
```

```
Convert the ODE into
a system
D(t, x):=[ $\begin{matrix} x_2 \\ A'(t) \cdot x_1 + A(t) \cdot x_2 + B'(t) \end{matrix}$ ]
```

```
Exact solution,
numerically evaluated
xs:=∫tot (1-B(0)+B(τ)).e-IA(τ) dτ.eIA(t)      xs(t):=xs
```

The numerical evaluation for the integrals could be very slow. Enable this for speed up the calculations, only if Maple can find a close solution and SMath recognize all functions.

```
xs(t):=feval("maple", xs)=√5 · (2 · (√5 · (1 - exp(- $\frac{t \cdot (4+t)}{20}$ ))) + √π · (-erf( $\frac{\sqrt{5} \cdot (2+t)}{10}$ ) + erf( $\frac{1}{\sqrt{5}}$ ))) · exp( $\frac{1}{5}$ )) + √π ·
```

```
Numerical solutions
xoT:=xoT
```

```
Method a: Trapezoidal rule      [T Xa Xa']:=MCols(DSolRes("D", [to te], xo, 0.5, N))
Method b: Backward Euler      [Tb Xb Xb']:=MCols(DSolRes("D", [to te], xo, 1-0, N))
Method c: Forward Euler      [Tc Xc Xc']:=MCols(DSolRes("D", [to te], xo, 0, N))
Method d: rkfixed from uni      [Td Xd Xd']:=MCols(rkfixed(xoT, to, te, N, D))
Method e: Rkadapt from uni      [Te Xe Xe']:=MCols(Rkadapt(xoT, to, te, N, D))
```

(T's are the same) $\text{normi} \left(\left[\text{normi} \left(T - T_d \right) \text{normi} \left(T - T_e \right) \right] \right) = 0$

Compare with the symbolic solution $k := [1..(N+1)]$ As exècted, Rkadapt it's better.

$$Xs_k := x_s \left(T_k \right)$$

$$Xs'_k := \frac{x_s \left(T_k + h \right) - Xs_k}{h}$$

$$\text{normi} \left(Xa - Xs \right) = 1.42 \cdot 10^{-2}$$

$$\text{normi} \left(Xb - Xs \right) = 5.33 \cdot 10^{-1}$$

$$\text{normi} \left(Xc - Xs \right) = 6.4 \cdot 10^{-1}$$

$$\text{normi} \left(Xd - Xs \right) = 6.19 \cdot 10^{-2}$$

$$\text{normi} \left(Xe - Xs \right) = 1.67 \cdot 10^{-10}$$

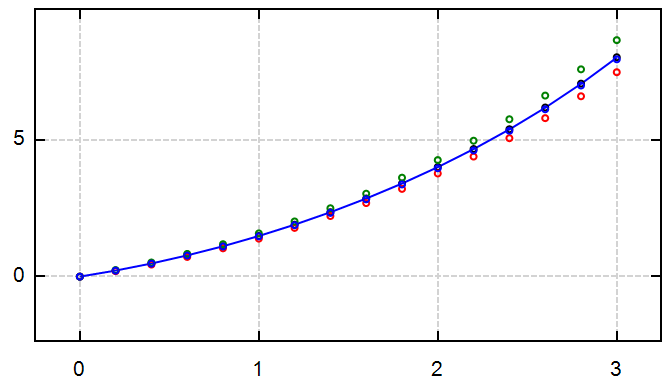
$$\text{normi} \left(Xa' - Xs' \right) = 3.38 \cdot 10^{-2}$$

$$\text{normi} \left(Xb' - Xs' \right) = 2.22 \cdot 10^{-1}$$

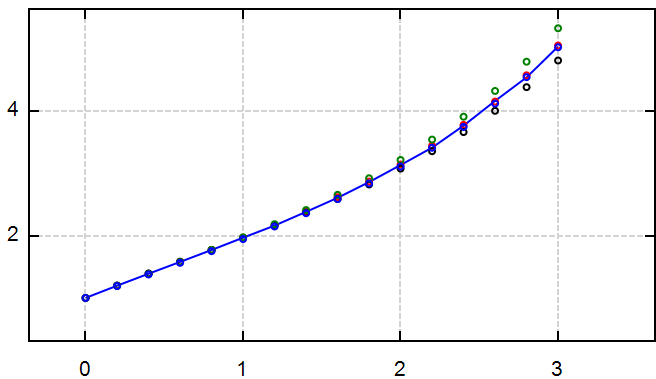
$$\text{normi} \left(Xc' - Xs' \right) = 2.96 \cdot 10^{-1}$$

$$\text{normi} \left(Xd' - Xs' \right) = 4.04 \cdot 10^{-2}$$

$$\text{normi} \left(Xe' - Xs' \right) = 2.66 \cdot 10^{-2}$$



```
{augment (T, Xs)
augment (T, Xa, "o", 3, "black")
augment (T, Xb, "o", 3, "red")
augment (T, Xc, "o", 3, "green")
augment (T, Xd, "o", 3, "blue")}
```



```
{augment (T, Xs')
augment (T, Xa', "o", 3, "red")
augment (T, Xb', "o", 3, "black")
augment (T, Xc', "o", 3, "green")
augment (T, Xd', "o", 3, "blue")}
```

DSolRes Example

Example: How to build an example

Obviously, start from the solution:

Thus

$$n := [1..4]$$

$$S(t) := \begin{bmatrix} t \cdot \sin(t^2) + 2 \\ \ln(1+t) - \ln(4) \\ e^{-t} \\ \cos(t^2) \end{bmatrix}$$

$$D(t, x) := \left(dx_n := \frac{d}{dt} S(t)_n \right) = \begin{bmatrix} 2 \cdot t^2 \cdot \cos(t^2) + \sin(t^2) \\ \frac{1}{1+t} \\ -\frac{1}{e^t} \\ -2 \cdot t \cdot \sin(t^2) \end{bmatrix}$$

You can use this D for, but just complicate the things

$$D(t, x) := \begin{bmatrix} 2 \cdot t^2 \cdot x_4 - \frac{x_1 - 2}{\ln(x_3)} \\ \frac{1}{x_2} \\ 4 \cdot e^{-x_3} \\ -2 \cdot (x_1 - 2) \end{bmatrix}$$

Try to disable this for check that both D's are the same.

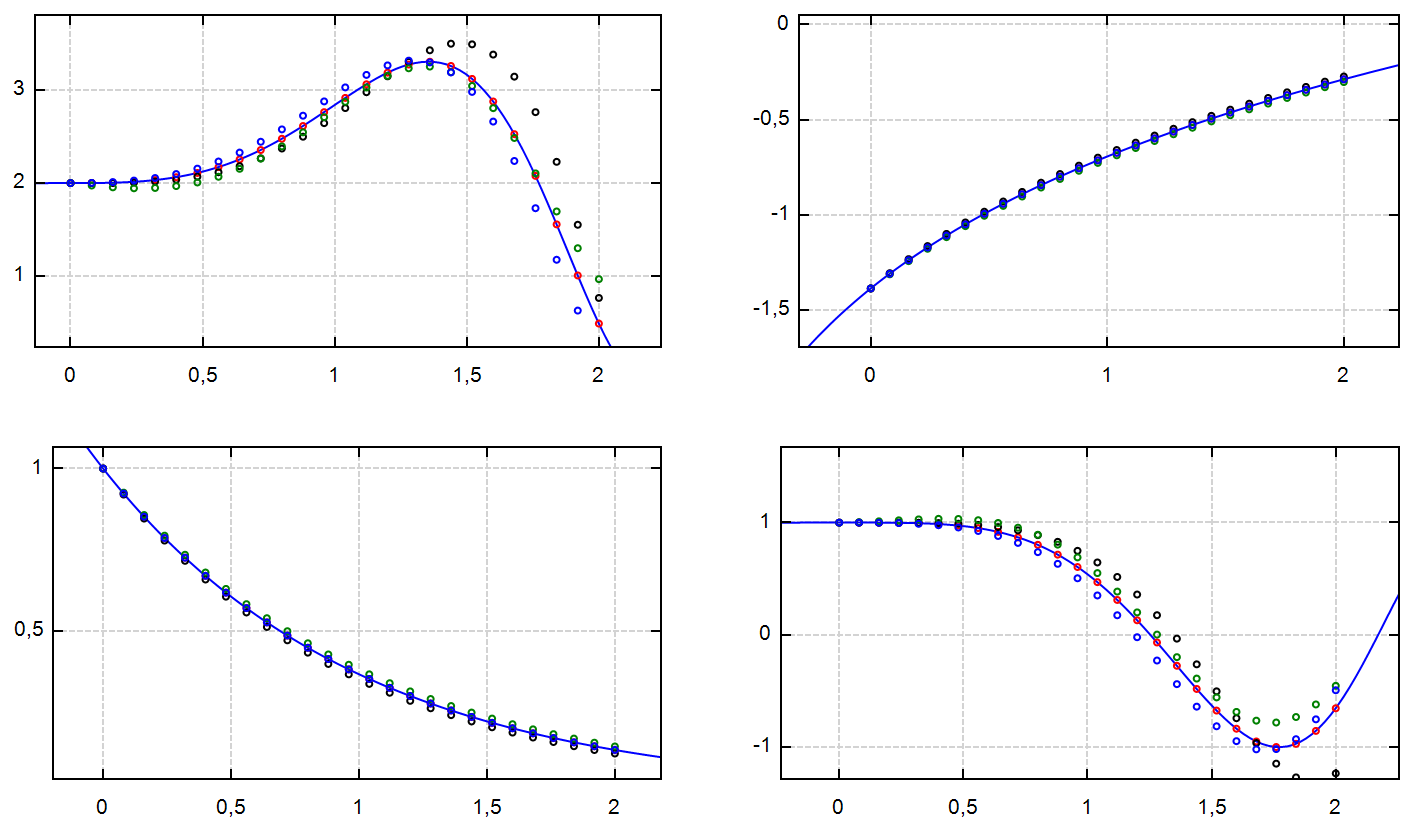
Given to := 0 te := 2 N := 25 Avoid t = 0

Now initials conditios
are just $xo := S (to)^T = [2 \ -1.3863 \ 1 \ 1]$ $xoT := xo^T$

Numerical solutions

Method a: Trapezoidal rule $[T \ Xa \ Ya \ Za \ Ua] := MCols (DSolRes ("D", [to \ te], xo, 0.5, N))$
Method b: Backward Euler $[Tb \ Xb \ Yb \ Zb \ Ub] := MCols (DSolRes ("D", [to \ te], xo, 1 - 0, N))$
Method c: Forward Euler $[Tc \ Xc \ Yc \ Zc \ Uc] := MCols (DSolRes ("D", [to \ te], xo, 0, N))$
Method d: rkfixed from uni $[Td \ Xd \ Yd \ Zd \ Ud] := MCols (rkfixed(xoT, to, te, N, D))$
Method e: Rkadapt from uni $[Te \ Xa \ Ya \ Za \ Ua] := MCols (Rkadapt(xoT, to, te, N, D))$

Compare with the (given) symbolic solution S(t) Hard to decide wich's better.



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