



The Halfway Way: A fresh look at the midpoint method

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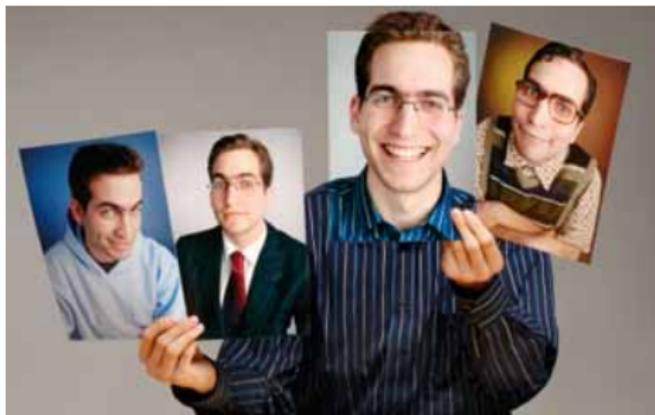
The Executive Summary



- The midpoint method is second order and absolutely stable;
- It is B-stable;
- It preserves linear and quadratic conservation quantities;
- It produces reliable error estimates;
- Safe time-steps are calculated accurately, efficiently, and adaptively;
- Existing backward Euler codes upgrade with one line of new code;
- C, C++, Fortran, FreeFem, MATLAB, Octave, Python, R versions.



Multiple Identities



Methods are for ODE's, rules for numerical quadrature.

| | | |
|-----------|---|---|
| | $f(\text{mid})$ | $\frac{1}{2}(f(\text{left})+f(\text{right}))$ |
| explicit: | explicit midpoint method | Heun's method or Improved Euler method |
| implicit: | implicit Runge-Kutta 2 or implicit midpoint method | trapezoidal method |

From now on, "midpoint method" refers to *implicit midpoint method*.



Two Ways to Look at It



The implicit midpoint method can be seen as:

$$y_{n+1} = y_n + \tau_n f(t_{n+1/2}, y_{n+1/2})$$

or as Backward and Forward Euler steps of size $\frac{\tau_n}{2}$:

$$y_{n+1/2} = y_n + \frac{\tau_n}{2} f(t_{n+1/2}, y_{n+1/2}) \quad (\text{BE: Backward Euler})$$

$$y_{n+1} = y_{n+1/2} + \frac{\tau_n}{2} f(t_{n+1/2}, y_{n+1/2}) \quad (\text{FE: Forward Euler})$$

The second step can be rewritten simply as:

$$y_{n+1} = 2y_{n+1/2} - y_n \quad (\text{FE: Forward Euler})$$

so the implicit problem only needs to be solved once, in BE.



Cauchy's One Leg θ method



The resulting method can be designated as (BEFE):

$$y_{n+1/2} = y_n + \frac{\tau_n}{2} f(t_{n+1/2}, y_{n+1/2}) \quad (\text{BE: Backward Euler})$$

$$y_{n+1} = 2y_{n+1/2} - y_n \quad (\text{FE: Forward Euler})$$

and admits a generalization to Cauchy's one-leg θ method:

$$y_{n+\theta_n} = y_n + \theta_n \tau_n f(t_{n+\theta_n}, y_{n+\theta_n})$$

$$y_{n+1} = \frac{1}{\theta_n} y_{n+\theta_n} - \left(\frac{1}{\theta_n} - 1\right) y_n$$





The θ -method for $\frac{1}{2} \leq \theta_n \leq 1$, and the BEFE special case are unconditionally stable, A-stable, and B-stable.

We say a method is **B-stable** if, for all u, v elements of a Banach or Hilbert space, and $\forall f()$ for which $\langle f(u) - f(v), u - v \rangle \leq 0$, we have $\|y_{n+1} - z_{n+1}\| \leq \|y_n - z_n\|$, for any two sequences y and z of approximations computed with the method, and any index n .

B-stability implies A-stability.



Error Estimates for Adaptive Stepsize



For a smooth exact solution $y(x)$, the local truncation error for BEFE is

$$T_{n+1} \equiv y(t_{n+1}) - y_{n+1} = \frac{1}{24} \tau_n^3 y'''(t_n + 1/2) + \mathcal{O}(\tau_n^5)$$

For a given local error tolerance tol , propose the next time step as

$$\tau_{n+1} = \kappa \tau_n \left(\frac{\text{tol}}{\|T_{n+1}\|} \right)^{\frac{1}{3}}$$

where the safety factor $\kappa \leq 1$.



How to go MAD: Midpoint Adaptive method:



$t_0, y_0, \text{tol}, T, \kappa$ given.

t_1, y_1, τ_0 from one step second-order method in convergence range

$t^{\text{new}} = t_1, \tau^{\text{new}} = \tau_0, n = 1$

while $t_n \leq T$ **do**

$\tau_n \leftarrow \tau^{\text{new}};$

 evaluate y_{n+1} with the midpoint rule;

 evaluate \widehat{T}_{n+1} ;

$\tau^{\text{new}} \leftarrow \kappa \tau_n \left| \text{tol} / \|\widehat{T}_{n+1}\| \right|^{\frac{1}{3}};$

if $\|\widehat{T}_{n+1}\| \leq \text{tol}$ **then**

$t_{n+1} \leftarrow t_n + \tau^{\text{new}},$

$n \leftarrow n + 1$

end

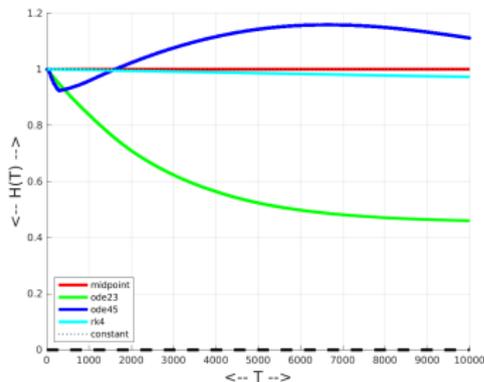
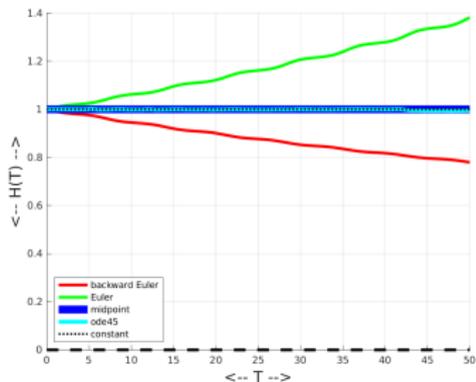
end



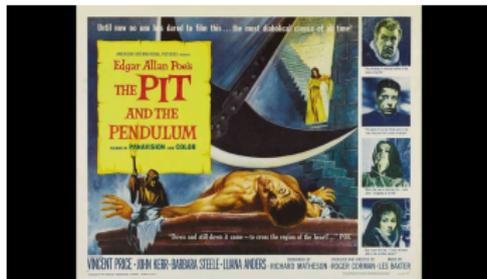
Test: Rigid Body Rotation



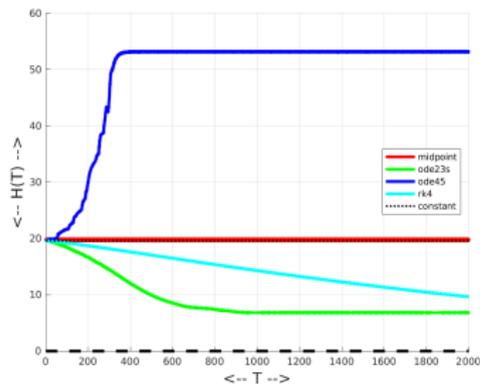
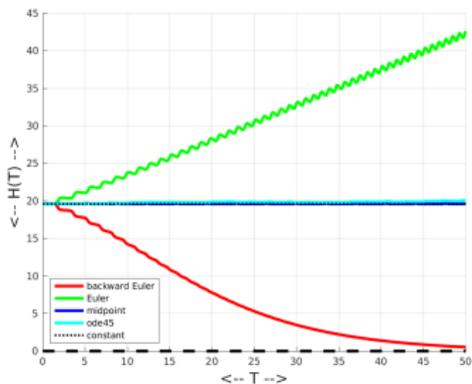
$$\text{Conservation: } H(t) = u^2 + v^2 + w^2$$



Test: Nonlinear Pendulum



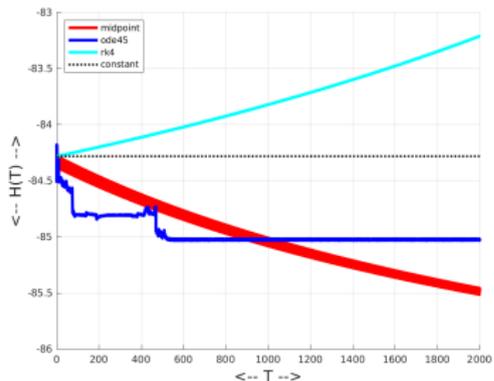
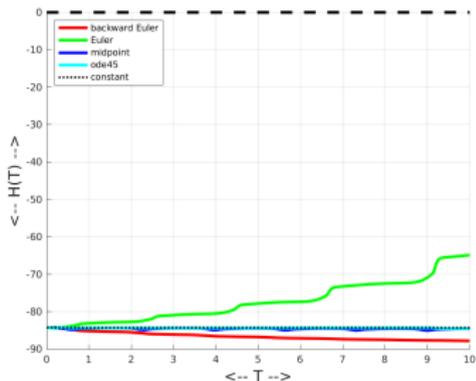
$$\text{Conservation: } H(t) = \frac{mg}{l}(1 - \cos(u)) + \frac{1}{2}mv^2$$



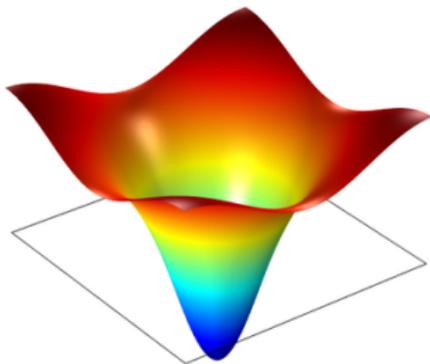
Test: Predator Prey



Conservation: $H(t) = \delta u - \gamma \log(u) + \beta v - \alpha \log(v)$
(not quadratic!)



Implicit Solvers Must Handle Nonlinear Equations



Any implicit ODE method must reliably solve a sequence of systems of nonlinear equations. Given the small stepsizes of a typical ODE method, the previous ODE solution is often a good first approximation to the solution at a new (but very close) time.

A simple method that usually works is to apply a fixed point iteration.

The developers of MINPACK provided the function `hybrd()` for solving general systems of nonlinear equations, and versions of this code are available in MATLAB, Python, and R under the name `fsolve()`.



Fixed Point Iteration



Ada Lovelace: *"The calculation will eat its own tail."*

```
tm = t0 + 0.5 * dt
ym = y0 + 0.5 * dt * f(t0, y0)

for j = 1 : it_max
    ym = y0 + 0.5 * dt * f(tm, ym)
end

t1 = t0 + dt
y1 = 2 * ym - y0
```



Using fsolve()



If you're close enough, you can't miss!

```
th = to + 0.5 * dt;  
yh = yo + 0.5 * dt * ( f ( to , yo ) )';  
yh = fsolve ( @(yh)residual(f,to,yo,th,yh), yh );  
...  
function value = residual ( f , to , yo , th , yh )  
    value = yh - yo - ( th - to ) * ( f ( th , yh ) )';  
    return  
end
```





Enter your ODE, and crank out the result!

| Language | fixed point | fsolve | adaptive |
|-----------|--------------------|--------------|----------------------|
| C | midpoint_fixed.c | midpoint.c | |
| C++ | midpoint_fixed.cpp | midpoint.cpp | |
| Fortran77 | midpoint_fixed.f | midpoint.f | |
| Fortran90 | midpoint_fixed.f90 | midpoint.f90 | |
| FreeFem | | midpoint.edp | |
| MATLAB | midpoint_fixed.m | midpoint.m | midpoint_adaptive.m |
| Octave | midpoint_fixed.m | midpoint.m | midpoint_adaptive.m |
| Python | midpoint_fixed.py | midpoint.py | midpoint_adaptive.py |
| R | midpoint_fixed.R | midpoint.R | |



Professional Codes available



Or you may prefer your ODE to be handled by a professional!

| Language | library | code |
|----------|------------------------|---------------------------------------|
| C | Gnu Scientific Library | <code>gsl_odeiv2_step_rk2imp()</code> |
| C++ | Gnu Scientific Library | <code>gsl_odeiv2_step_rk2imp()</code> |
| Julia | ODE | Midpoint |



References



Catalin Trenchea, John Burkardt,
Refactorization of the midpoint rule,
Applied Mathematics Letters, Volume 107, September 2020,

Catalin Trenchea, John Burkardt,
Refactorization of the midpoint rule,
Technical Report TR-MATH 20-02,
https://www.mathematics.pitt.edu/sites/default/files/midpoint3_technicalreport.pdf,

John Burkardt, Wenlong Pei, Catalin Trenchea,
A stress test for the midpoint time-stepping method,
International Journal of Numerical Analysis and Modeling, Volume 19, Number
2-3, pages 299-314, 2022.





Links to source code in any language:

[https://people.sc.fsu.edu/~jburkardt/...](https://people.sc.fsu.edu/~jburkardt/)

| Language | subdirectory | Sample directory |
|-----------|--------------|------------------|
| C | c_src/ | midpoint.html |
| C++ | cpp_src/ | midpoint.html |
| Fortran77 | f77_src/ | midpoint.html |
| Fortran90 | f_src/ | midpoint.html |
| FreeFem++ | freefem_src/ | midpoint.html |
| MATLAB | m_src/ | midpoint.html |
| Octave | octave_src/ | midpoint.html |
| Python | py_src/ | midpoint.html |
| R | r_src/ | midpoint.html |

For most languages, there are actually several implementations: fixed point/fsolve/adaptive.





- The midpoint method is powerful, accurate, and stable.
- The method is A-stable, B-stable, linearly and nonlinearly stable.
- It is a symplectic method for general Hamiltonian systems.
- The correct estimator for local truncation error only involves the differentiation defect, but not the interpolation defect.
- Implementations are provided in a variety of computing languages.

