

Solving PDEs with Backward Implicit Time Steps

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This package [EconPDEs.jl](#) introduces a fast and robust way to solve systems of PDEs + algebraic equations (i.e. DAEs) associated with economic models. It builds on the method presented in [Achdou et al. \(2016\)](#), but extends it to handle non-linearities.

Consider a PDE of the form

$$0 = f_0(x, V) + f_1(x, \partial_x V) \partial_x V + f_2(x, \partial_x V) \partial_{xx} V + \partial_t V \quad (1)$$

As in [Achdou et al. \(2016\)](#), I first discretize the state x on a grid and I approximate derivatives $\partial_x V_t$ and $\partial_{xx} V_t$ using finite difference approximations. First order derivatives are upwinded, which helps with convergence and ensures that boundary conditions are satisfied.

Given V_{t+1} , I solve for V_t using a backward implicit time step:

$$0 = f_0(x, V_t) + f_1(x, \partial_x V_t) \partial_x V_t + f_2(x, \partial_x V_t) \partial_{xx} V_t + \frac{1}{\Delta}(V_{t+1} - V_t) \quad (2)$$

This non-linear equation is solved using a Newton-Raphson method. Julia packages `ForwardDiff` and `SparseDiffTools` are used to compute the (sparse) Jacobian of (2) automatically.

Difference with [Achdou et al. \(2016\)](#) [Achdou et al. \(2016\)](#) propose to solve the PDE (1) using a backward *semi*-implicit time steps of the form:

$$0 = f_0(x, V_t) + f_1(x, \partial_x V_{t+1}) \partial_x V_t + f_2(x, \partial_x V_{t+1}) \partial_{xx} V_t + \frac{1}{\Delta}(V_{t+1} - V_t) \quad (3)$$

Compared to (2), each time step corresponds to a *linear* equation in V_t . This makes it easier to solve each step, since one only needs to invert a matrix. However, this simplification makes the method less robust: one can show that the associated scheme is typically non monotonous, unless f_1 and f_2 do not depend on the value function.

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Stationary Solution In most cases, one is only interested in the stationary solution of the PDE (1), i.e.,

$$0 = f_0(x, V) + f_1(x, \partial_x V) \partial_x V + f_2(x, \partial_x V) \partial_{xx} V \quad (4)$$

In this case, I use the same method, but I adapt the time step Δ over time. More precisely, if the backward time step (2) is successful (i.e., the Newton-Raphson method converges), Δ is increased; otherwise, it is decreased. This method ensures convergence since, the Newton-Raphson method always converges for Δ small enough. Yet, it does not sacrifice speed: as $\Delta \rightarrow \infty$, the method becomes equivalent to a non-linear solver for the PDE, which ensures quadratic convergence around the solution.

The idea of adapting Δ comes from the Pseudo-Transient Continuation method used in the fluid dynamics literature. Formal conditions for the convergence of the algorithm are given in [Kelley and Keyes \(1998\)](#).

Applications Empirically, I find the method to be fast and robust — the examples folder shows that the algorithm solves a wide range of asset pricing models.

References

- Achdou, Yves, Jiequn Han, Jean-Michel Lasry, Pierre-Louis Lions, and Benjamin Moll**, “Heterogeneous Agent Models in Continuous Time,” 2016. Working Paper.
- Kelley, Carl Timothy and David E Keyes**, “Convergence analysis of pseudo-transient continuation,” *SIAM Journal on Numerical Analysis*, 1998, 35 (2), 508–523.