

Project #2: Newton-Krylov method for solving nonlinear systems

Advanced numerical analysis

Behzad Baghapour

1 Introduction

We examine the Newton-Krylov algorithm for solving the steady-state nonlinear Burger's equation using `scipy.optimize` package. The effect of line search and preconditioning are studied in convergence rate. The simulations are performed in a one-dimensional domain with uniform mesh of different sizes. Please provide your answers in a PDF file format including the explanations, algorithm, graphs, and codes developed in your report. For your report, consider *problem statement*, *methodology*, *results and discussions*, and *conclusion* sections. Submit your answers to the Courses according to the prescribed due-date.

2 Problem statement

Burger's equation has a steady-state solution with the prescribed boundary conditions as follows:

$$\begin{cases} u \frac{du}{dx} - \nu \frac{d^2u}{dx^2} = 0, \\ u(0) = +1, \\ u(1) = -1 \end{cases} \quad (1)$$

In Eq. (1), u is the scalar property. The first term, udu/dx is the convection and $\nu d^2u/dx^2$ is the diffusion of the property in the one-dimensional domain and ν is the diffusivity coefficient. The problem can be seen as a nonlinear system which the steady-state solution $u(x)$ is seeking to find as a root of nonlinear equation $F(u) = 0$ in the conservative form:

$$F(u) = 0, \quad F(u) := \frac{d(u^2/2)}{dx} - \nu \frac{d^2u}{dx^2} \quad (2)$$

The domain is discretized into uniform mesh $\Delta x = 1.0/(n - 1)$ where n is the number of grid points. The nonlinear operator in Eq. (1) is also discretized by the second-order central finite difference method. Therefore, for each node of domain we have:

$$F(i) = \frac{1}{4\Delta x}(u_{i+1}^2 - u_{i-1}^2) - \frac{\nu}{\Delta x^2}(u_{i+1} - 2u_i + u_{i-1}), \quad i = 0, 1, \dots, n - 1 \quad (3)$$

The norm $\|F\|$ is going to be minimized using the Newton-Krylov algorithm. Accordingly, the nonlinear problem is linearized as follows:

$$J(x)\delta x = R(x) \quad (4)$$

where $J(x) = F'(x)$ is the Jacobian of F and $R(x) = -F(x)$ is the nonlinear residual. For a discretized domain with n grid points, the Jacobian, $J_{n,n}$ has the dimension of $n \times n$ and the residual R_n has the dimension of n . The linearized system of equations (4) can be solved using fast Krylov methods (e.g. GMRES). The overall solution steps can be summarized as follows:

Newton-Krylov algorithm

1. Linearizing the nonlinear system $F(\mathbf{x})=0$
2. Solving the linearized system with Krylov methods:
 - 2.1. Preconditioning the linearized system
3. Global convergence check by stabilization techniques:
 - 3.1. Line search, Trust zone, ...
4. Update the solution for the next iteration

Figure 1(a) compares the field with different diffusivity coefficients and Fig. 1(b) shows the effect of preconditioning on the convergence with *Armijo* line search.

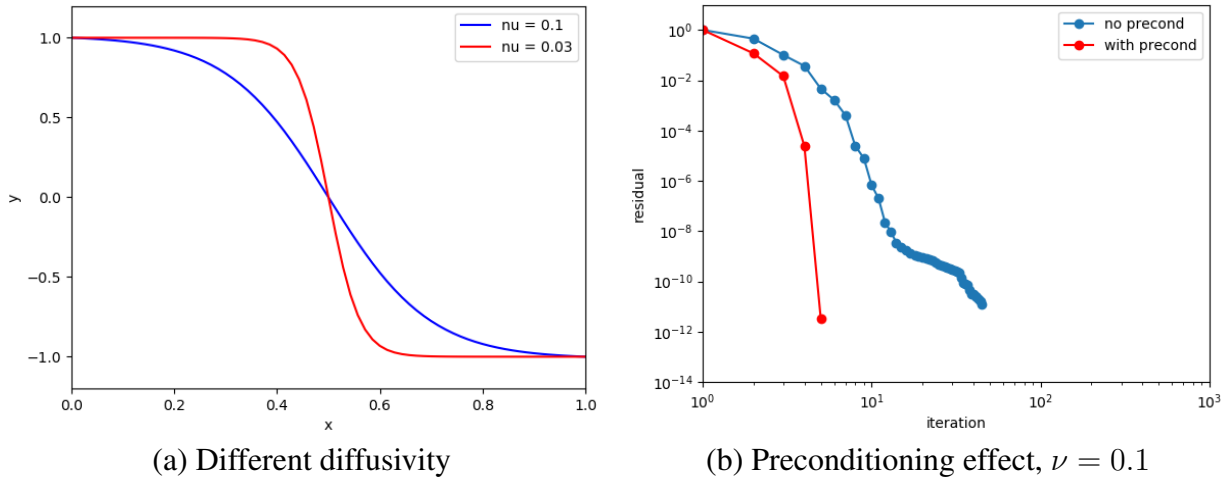


Figure 1: Solution of steady-state Burger's equation for $n = 71$.

3 Solution method

The system of nonlinear equations (3) is going to be solved using `newton_krylov` method of `scipy.optimize` module. The mesh dimensions considered for the domain $-1 \leq x \leq +1$ is $nx = \{51, 101, 301\}$ with the nonlinear residual norm of `f_tol=1e-10` and initial guess $\mathbf{u}_0 = 0$.

1. Compare the convergence rate (log scale), without line search and preconditioning, for the mesh size $n = 301$ with $\nu = 0.1, 0.01, 0.001$. Plot the scalar field for different diffusivity and analyze the effect of dissipation on the field and convergence.
2. For $\nu = 0.1$, and different mesh sizes with *Armijo* line search, study the effect of preconditioning M on the convergence rate of `newton_krylov`. Accordingly, the viscous (diffusion) part of Eq. (1) with sparse ILU `spilu` and `fill_factor=10` is considered:

$$F_v(u) = \nu \frac{\partial^2 u}{\partial x^2} \approx \frac{\nu}{\Delta x^2} (u_{i+1} - 2u_i + u_{i-1}) \rightarrow M_{ij} = \frac{\partial F_{v,i}}{\partial u_j}, M_{i,i\pm 1} = \frac{\nu}{\Delta x^2}, M_{i,i} = \frac{-2\nu}{\Delta x^2}$$

To set the preconditioner, use `LinearOperator` and introduce it to the method by `inner_M`.