

Implementation of GMRES with Preconditioning in Python

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Convection-diffusion problem

$$\begin{aligned} -au''(x) + bu'(x) &= 0, \quad 0 < x < 1 \\ u(0) &= 0, u(1) = 1 \end{aligned}$$

Right-hand side of differential is 0

Dirichlet boundary conditions

Difference equation for u' (backward)

$$u' = \frac{u_i - u_{i-1}}{h}$$

backward difference

$$u'' = \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2}$$

$$-a \left(\frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} \right) + b \left(\frac{u_i - u_{i-1}}{h} \right) = 0$$

Forming Tri-diagonal Matrix A

$$-u_{i+1} + (2 + c)u_i - (1 + c)u_{i-1} = 0$$

where the Péclet number $R = b/a$ and $c = Rh$

Let $m = 5$ (visualization)

$$A_{backward} = \frac{a}{h^2} \begin{bmatrix} 2+c & -1 & & & \\ -(1+c) & 2+c & -1 & & \\ & -(1+c) & 2+c & -1 & \\ & & -(1+c) & 2+c & -1 \\ & & & -(1+c) & 2+c \end{bmatrix}$$
$$y = \frac{a}{h^2} \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ -\gamma \end{bmatrix} = \frac{a}{h^2} \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \quad \gamma = -1 \text{ for backward}$$

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Objective

$x \in x_0 + \mathcal{K}_m$ – a linear combination of basis vectors in \mathcal{K}_m

$$x = x_0 + V_m y$$

With results from Arnoldi iteration

Skip Arnoldi iteration (psuedocode)

$$\min_y \|b - A(x_0 + V_m y)\|_2 = \min_y \|\beta e_1 - \bar{H}_m y\|_2$$

GMRES Plane Rotations

Introduce plane rotation matrix

Reduce Hessenberg matrix (upper triangular)

Progressive manner

Residual at each iteration

$$\Omega_i = \begin{bmatrix} 1 & & & & & & \\ & \ddots & & & & & \\ & & 1 & & & & \\ & & & c_i & s_i & & \\ & & & -s_i & c_i & & \\ & & & & & 1 & \\ & & & & & & \ddots \\ & & & & & & & 1 \end{bmatrix}$$

GMRES Plane Rotations Example

Let $m = 3$ (visualization)

$$\bar{H}_3 = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ & h_{22} & h_{23} \\ & & h_{33} \\ & & & h_{43} \end{bmatrix}, \quad \bar{g}_3 = \begin{bmatrix} \gamma_1 \\ \gamma_2 \\ \gamma_3 \\ 0 \end{bmatrix}$$

$$\bar{R}_3 = \begin{bmatrix} h_{11} & h_{12} & h_{13}^* \\ & h_{22} & h_{23}^* \\ & & h_{33}^* \\ & & & 0 \end{bmatrix}, \quad \bar{g}_3 = \begin{bmatrix} \gamma_1 \\ \gamma_2 \\ c_6 \gamma_3 \\ -s_6 \gamma_3 \end{bmatrix}$$

$$\min_y \left\| \beta e_1 - \bar{H}_m y \right\|_2 = \min_y \left\| \bar{g}_m - \bar{R}_m y \right\|_2$$

Preconditioner (Left)

Preconditioners:

Jacobi (JA)

Gauss-Seidel (GS)

Symmetric Gauss-Seidel (SGS), $\omega = 1$ in SSOR

$$M^{-1}Ax = M^{-1}y$$

$$M_{JA} = D$$

$$M_{GS} = D + L$$

$$\begin{aligned} M_{SGS} &= (D + L)D^{-1}(D + U) \\ &= \hat{L}\hat{U} \end{aligned}$$

$$\hat{L} = (D + L)D^{-1} = I - LD^{-1}$$

$$\hat{U} = D + U$$

SGS - forward and backward solve (approximate LU factorization)

Pseudocode

Terminate early (threshold)

Algorithm GMRES (plane rotations) with preconditioning

- 1: Choose an initial guess x_0
 - 2: Compute $r_0 = M^{-1}(b - Ax_0)$, $\beta = \|r_0\|_2$, $v_1 = r_0/\beta$
 - 3: **for** $j = 1, 2, \dots, m$ **do**
 - 4: START Arnoldi iteration step
 - 5: Compute $w_j = M^{-1}(Av_j)$
 - 6: **for** $i = 1, 2, \dots, j$ **do**
 - 7: $h_{ij} = (w_j, v_i)$
 - 8: $w_j = w_j - h_{ij}v_i$
 - 9: **end for**
 - 10: $h_{j+1,j} = \|w_j\|_2$
 - 11: **if** $h_{j+1,j} == 0$ **then** ▷ encountered with *preconditioning*
 - 12: Set $m = j$ and goto 22
 - 13: **end if**
 - 14: $v_{j+1} = w_j/h_{j+1,j}$
 - 15: END Arnoldi iteration step
 - 16: START plane rotation step
 - 17: Apply rotation matrix to rows $i, i + 1$ of h along columns $1, 2, \dots, j$
 - 18: Update \bar{g}_m on indices $i, i + 1$ of \bar{g}_m according to equations given by [1]
 - 19: If the norm of the residual $|\gamma_{m+1}|$ is less than some threshold, goto 22
 - 20: END plane rotation step
 - 21: **end for**
 - 22: Define the $(m + 1) \times m$ Hessenberg matrix $\bar{H}_m = \{h_{ij}\}$ for $1 \leq i \leq m + 1, 1 \leq j \leq m$
 - 23: With plane rotations, let $\bar{R}_m = \bar{H}_m$ where \bar{R}_m is the upper triangular form of \bar{H}_m and removing the zero'd row as an $(m + 1) \times m$ matrix
 - 24: Compute y_m the minimizer of $\|\bar{g}_m - \bar{R}_m y\|_2$
 - 25: Compute $x_m = x_0 + V_m y_m$
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Methods to Compare

Residual and Runtime plots

LU factorization (using the built-in `scipy.sparse.linalg.splu`)

Conjugate Gradient (CG) (***added to slides***)

Preconditioned CG (***added to slides***)

GMRES

Preconditioned GMRES

Sparse format

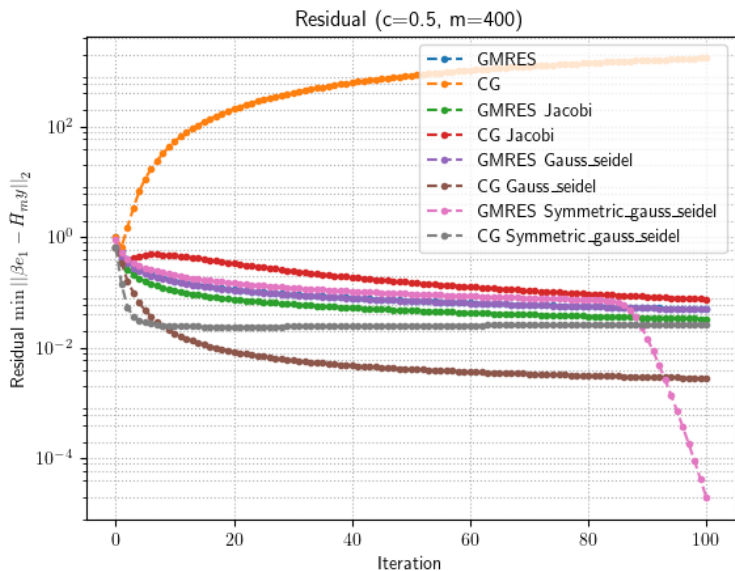
$$m = \{400, 800, 1600, 3200, 6400, 12800\}$$

$$c = \{0.5, 1, 10, 100, 1000\}$$

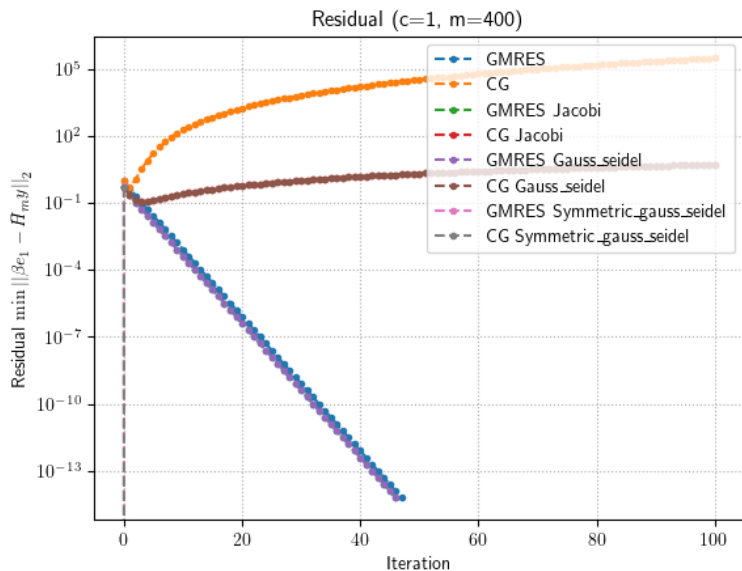
max iterations: 100

tolerance: $1e-14$

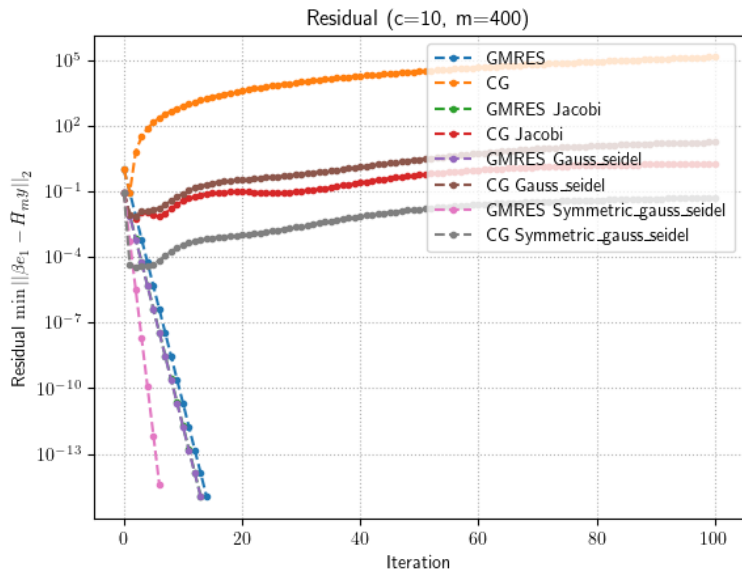
Residual Plots I



Residual Plots II

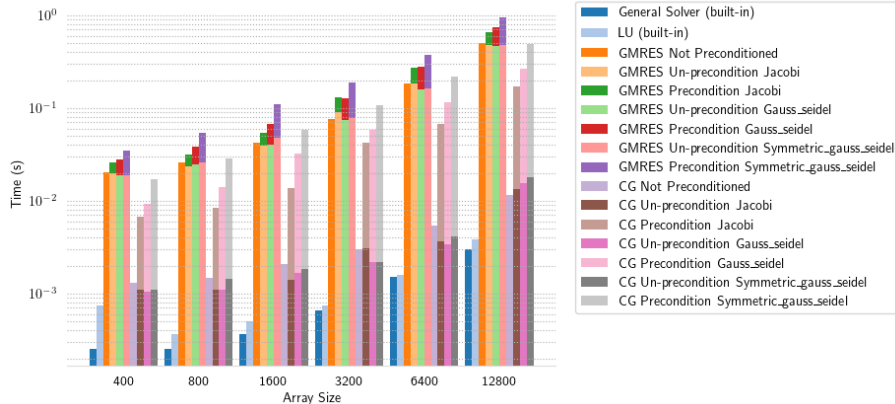


Residual Plots III



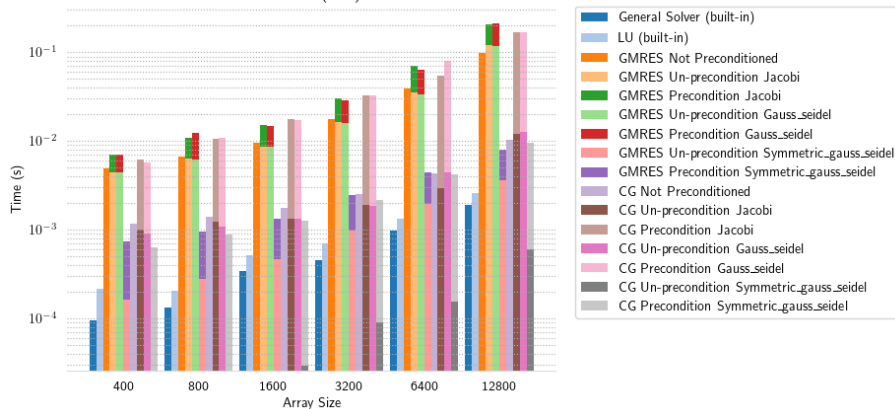
Runtime Plots I

Runtime ($c=0.5$)



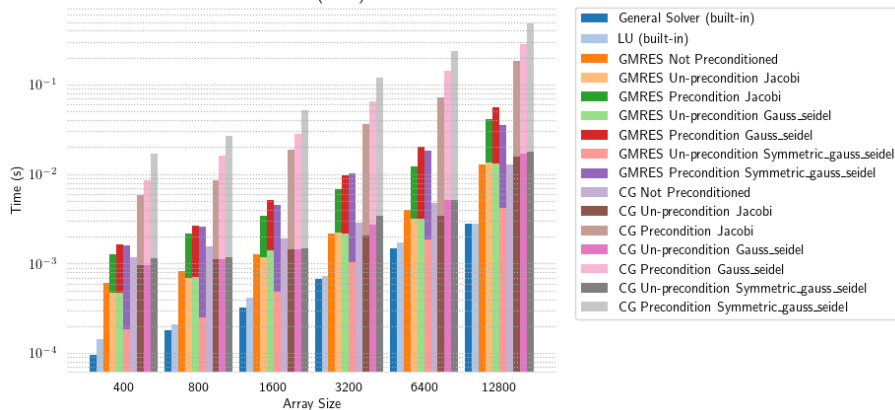
Runtime Plots II

Runtime (c=1)



Runtime Plots III

Runtime (c=10)



Code Output

```
n: 480
c: 1

LU error: 0.0

precondition: None

GMRES
num iterations: 47
norm error: 4.58653364590355e-15
errors list: [1.0, 0.4472135954999579, 0.21821789023
total runtime (true): 7.700920104980469e-05
total runtime: 0.004621028900146484
precondition time: 0.0001976498028751953
non-precondition time: 0.004423379898071289

Conjugate Gradient
num iterations: 100
norm error: 286.5959730253588
errors list: [1.0, 0.5, 1.222222222222222, 3.3650519
total runtime (true): 7.700920104980469e-05
total runtime: 0.001104116439819336
precondition time: 0.0003116138828897422
non-precondition time: 8.0007925033569335938
```

```
precondition: PreconditionEnum.JACOBI
```

```
GMRES
num iterations: 46
norm error: 9.173067288213951e-15
errors list: [0.5, 0.22360679774997896, 0.1091089451
total runtime (true): 7.700920104980469e-05
total runtime: 0.006892681121826172
precondition time: 0.002471923828125
non-precondition time: 0.004420757293701172
```

```
Conjugate Gradient
num iterations: 100
norm error: 1.2487480063507874
errors list: [0.5, 0.25, 0.140625, 0.114067925347222
total runtime (true): 7.700920104980469e-05
total runtime: 0.0067291259765625
precondition time: 0.005760908126831055
non-precondition time: 0.0009682178497314453
```

```
precondition: PreconditionEnum.GAUSS_SEIDEL
```

```
GMRES
num iterations: 46
norm error: 9.173067288213951e-15
errors list: [0.5, 0.22360679774997896, 0.1091089451
total runtime (true): 7.700920104980469e-05
total runtime: 0.006891965866088867
precondition time: 0.0025691986083984375
non-precondition time: 0.00432276725769843
```

```
Conjugate Gradient
num iterations: 100
norm error: 1.2487480063507874
errors list: [0.5, 0.25, 0.140625, 0.114067925347222
total runtime (true): 7.700920104980469e-05
total runtime: 0.0060389041900634766
precondition time: 0.00509983428955078125
non-precondition time: 0.0009405612945556641
```

```
precondition: PreconditionEnum.SYMMETRIC_GAUSS_SEIDEL
```

```
GMRES
ENCOUNTER EXACT SOLUTION
num iterations: 1
norm error: 0.0
errors list: [0.5773502691896257, 0]
total runtime (true): 7.700920104980469e-05
total runtime: 0.00078606060552978516
precondition time: 0.0006048679351806641
non-precondition time: 0.0001811981201171875
```

```
Conjugate Gradient
num iterations: 1
norm error: 0.0
errors list: [0.5, 0.0]
total runtime (true): 7.700920104980469e-05
total runtime: 0.0006761550903320312
precondition time: 0.0006520748138427734
non-precondition time: 2.4000276489257812e-05
```

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Questions?