asQ - ParaDiag methods with Firedrake and PETSc

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Outline

Parallel-in-time motivation

ParaDiag methods

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Time-dependent O/PDEs

$$\mathbf{M}\partial_t u + \mathbf{K} u = b(t)$$

$$M\left(\frac{u^{n+1}-u^n}{\Delta t}\right) + \theta K u^{n+1} + (1-\theta) K u^n = b^{n+1}$$

$$\left(\frac{\mathbf{M}}{\Delta t} + \theta \mathbf{K}\right) u^{n+1} = b^{n+1} - \left(\frac{-\mathbf{M}}{\Delta t} + (1-\theta) \mathbf{K}\right) u^n$$

Performance of serial method

$$\left(\frac{\boldsymbol{M}}{\Delta t} + \theta \boldsymbol{K}\right) u^{n+1} = \tilde{b}$$

Work: $W_s = K_s M_s N_x^q N_t \sim N_x^q N_t$

Processors: $P_s \sim N_x$

Time:
$$T_s = \frac{W_s}{P_s} = K_s M_s N_t N_x^{q-1} \sim N_t N_x^{q-1}$$

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$$\left(\frac{-\boldsymbol{M}}{\Delta t} + (1 - \theta) \boldsymbol{K}\right) u^{n} + \left(\frac{\boldsymbol{M}}{\Delta t} + \theta \boldsymbol{K}\right) u^{n+1} = b^{n+1}$$

$$\mathbf{A}_0 u^n + \mathbf{A}_1 u^{n+1} = b^{n+1}$$

$${m A}_0 u^n + {m A}_1 u^{n+1} = b^{n+1}$$
 ${m A}_0 u^0 + {m A}_1 u^1 = b^1$

$$\mathbf{A}_{0}u^{n} + \mathbf{A}_{1}u^{n+1} = b^{n+1}$$
 $\mathbf{A}_{0}u^{0} + \mathbf{A}_{1}u^{1} = b^{1}$
 $\mathbf{A}_{0}u^{1} + \mathbf{A}_{1}u^{2} = b^{2}$

$$\mathbf{A}_0 u^n + \mathbf{A}_1 u^{n+1} = b^{n+1}$$

$$A_0 u^0 + A_1 u^1 = b^1$$
 $A_0 u^1 + A_1 u^2 = b^2$
 $A_0 u^2 + A_1 u^3 = b^3$

$$\mathbf{A}_0 u^n + \mathbf{A}_1 u^{n+1} = b^{n+1}$$

$$u^1$$

$$\mathbf{A}_0 u^n + \mathbf{A}_1 u^{n+1} = b^{n+1}$$

$$\begin{pmatrix} \mathbf{A}_1 & & & & \\ \mathbf{A}_0 & \mathbf{A}_1 & & & \\ & \mathbf{A}_0 & \mathbf{A}_1 & & \\ & & \mathbf{A}_0 & \mathbf{A}_1 \end{pmatrix} \begin{pmatrix} u^1 \\ u^2 \\ u^3 \\ u^4 \end{pmatrix} = \begin{pmatrix} b^1 - \mathbf{A}_0 u^0 \\ b^2 \\ b^3 \\ b^4 \end{pmatrix}$$

$$Au = b$$

Kronecker products

$$\mathbf{A} \in \mathbb{R}^{n \times n}$$
, $\mathbf{B} \in \mathbb{R}^{m \times m}$, $\mathbf{A} \otimes \mathbf{B} \in \mathbb{R}^{nm \times nm}$

$$\mathbf{A} \otimes \mathbf{B} = \begin{pmatrix} a_{00}\mathbf{B} & a_{01}\mathbf{B} & a_{02}\mathbf{B} \\ a_{10}\mathbf{B} & a_{11}\mathbf{B} & a_{12}\mathbf{B} \\ a_{20}\mathbf{B} & a_{21}\mathbf{B} & a_{22}\mathbf{B} \end{pmatrix}$$

$$(AC) \otimes (BD) = (A \otimes B) (C \otimes D)$$

 $A \otimes B = (A \otimes I) (I \otimes B) = (I \otimes B) (A \otimes I)$

θ -method all-at-once sytem

$$\boldsymbol{M} \left(\frac{u^{n+1} - u^n}{\Delta t} \right) + (1 - \theta) \boldsymbol{K} u^n + \theta \boldsymbol{K} u^{n+1} = b^{n+1}$$

$$\boldsymbol{A} = \begin{pmatrix} \frac{1}{\Delta t} & & & \\ \frac{-1}{\Delta t} & \frac{1}{\Delta t} & & \\ & \frac{-1}{\Delta t} & \frac{1}{\Delta t} & \\ & & \frac{-1}{\Delta t} & \frac{1}{\Delta t} \end{pmatrix} \otimes \boldsymbol{M} + \begin{pmatrix} \theta & & & \\ (1 - \theta) & \theta & & \\ & & (1 - \theta) & \theta & \\ & & & (1 - \theta) & \theta \end{pmatrix} \otimes \boldsymbol{K}$$

$$\mathbf{A} = \mathbf{B}_1 \otimes \mathbf{M} + \mathbf{B}_2 \otimes \mathbf{K}$$

ParaDiag-II: Circulant preconditioners

$$\begin{pmatrix} \frac{1}{\Delta t} & & \frac{-1}{\Delta t} \\ \frac{-1}{\Delta t} & \frac{1}{\Delta t} & & \\ & \frac{-1}{\Delta t} & \frac{1}{\Delta t} & \\ & & \frac{-1}{\Delta t} & \frac{1}{\Delta t} \end{pmatrix} \otimes \boldsymbol{M} + \begin{pmatrix} \theta & & (1-\theta) \\ (1-\theta) & \theta & & \\ & & (1-\theta) & \theta & \\ & & & (1-\theta) & \theta \end{pmatrix} \otimes \boldsymbol{K}$$

$$\begin{aligned} & \boldsymbol{P} = \boldsymbol{C}_1 \otimes \boldsymbol{M} + \boldsymbol{C}_2 \otimes \boldsymbol{K} \approx \boldsymbol{A} \\ & \boldsymbol{P} = \left(\boldsymbol{V} \otimes \boldsymbol{I}_{\times} \right) \left(\boldsymbol{D}_1 \otimes \boldsymbol{M} + \boldsymbol{D}_2 \otimes \boldsymbol{K} \right) \left(\boldsymbol{V}^{-1} \otimes \boldsymbol{I}_{\times} \right) \end{aligned}$$

• $C_{1,2} = VD_{1,2}V^{-1}$ are simultaneously diagonalisable

ParaDiag-II: Circulant preconditioners

$$\begin{pmatrix} \frac{1}{\Delta t} & & \frac{-\alpha}{\Delta t} \\ \frac{-1}{\Delta t} & \frac{1}{\Delta t} & & \\ & \frac{-1}{\Delta t} & \frac{1}{\Delta t} & \\ & & \frac{-1}{\Delta t} & \frac{1}{\Delta t} \end{pmatrix} \otimes \boldsymbol{M} + \begin{pmatrix} \theta & & \alpha \left(1 - \theta\right) \\ \left(1 - \theta\right) & \theta & & \\ & \left(1 - \theta\right) & \theta & \\ & & \left(1 - \theta\right) & \theta \end{pmatrix} \otimes \boldsymbol{K}$$

$$P^{(\alpha)} = C_1^{(\alpha)} \otimes M + C_2^{(\alpha)} \otimes K \approx A$$

$$P^{(\alpha)} = (V \otimes I_X) (D_1 \otimes M + D_2 \otimes K) (V^{-1} \otimes I_X)$$

- $C_{1,2}^{(\alpha)} = VD_{1,2}V^{-1}$ are simultaneously diagonalisable
- $\sim \alpha \in (0,1]$, and in practice can be very small ($\approx 10^{-4}$)

Circulant diagonalisation

$$P = C_1 \otimes M + C_2 \otimes K,$$
 $C_j = V D_j V^{-1},$ $P = (V \otimes I_x) (D_1 \otimes M + D_2 \otimes K) (V^{-1} \otimes I_x)$

$$\mathbf{\Lambda}_k = (\lambda_{1,k} \mathbf{M} + \lambda_{2,k} \mathbf{K}) \quad \forall k \in [1, N_t]$$

$$V = \Gamma_{\alpha}^{-1} \mathcal{F}^{-1}, \quad D_j = \operatorname{diag}\left(\mathcal{F}\Gamma_{\alpha} c_j\right), \quad \Gamma_{\alpha} = \operatorname{diag}\left(\alpha^{\frac{k-1}{N_t}}\right) \forall k \in [1, N_t]$$

$$\mathbf{P} = (\mathbf{V} \otimes \mathbf{I}_{\times}) (\mathbf{D}_{1} \otimes \mathbf{M} + \mathbf{D}_{2} \otimes \mathbf{K}) (\mathbf{V}^{-1} \otimes \mathbf{I}_{\times})$$

$$Px = b$$

$$P = (V \otimes I_{\times}) (D_1 \otimes M + D_2 \otimes K) (V^{-1} \otimes I_{\times})$$

$$Px = b$$

• Step-(a)
$$\mathbf{y}_1 = (\mathbf{V}^{-1} \otimes \mathbf{I}_{\times}) \mathbf{b}$$

$$P = (V \otimes I_{\times}) (D_1 \otimes M + D_2 \otimes K) (V^{-1} \otimes I_{\times})$$

$$Px = b$$

- Step-(a) $\mathbf{y}_1 = (\mathbf{V}^{-1} \otimes \mathbf{I}_x) \mathbf{b}$
- ► Step-(b) $(\lambda_{1,j}\mathbf{M} + \lambda_{2,j}\mathbf{K})\mathbf{y}_{2,n} = \mathbf{y}_{1,n} \quad \forall j \in [1, N_t]$

$$P = (V \otimes I_{\times}) (D_1 \otimes M + D_2 \otimes K) (V^{-1} \otimes I_{\times})$$

$$Px = b$$

- ► Step-(b) $(\lambda_{1,j}\mathbf{M} + \lambda_{2,j}\mathbf{K})\mathbf{y}_{2,n} = \mathbf{y}_{1,n} \quad \forall j \in [1, N_t]$
- Step-(c) $\mathbf{x} = (\mathbf{V} \otimes \mathbf{I}_{x}) \mathbf{y}_{2}$

$$P = (V \otimes I_{\times}) (D_1 \otimes M + D_2 \otimes K) (V^{-1} \otimes I_{\times})$$

$$Px = b$$

- ► Step-(b) $(\lambda_{1,j}\mathbf{M} + \lambda_{2,j}\mathbf{K})\mathbf{y}_{2,n} = \mathbf{y}_{1,n} \quad \forall j \in [1, N_t]$
- Step-(c) $\mathbf{x} = (\mathbf{V} \otimes \mathbf{I}_{x}) \mathbf{y}_{2}$

All-at-once solution strategies

- Preconditioned Krylov method: $P^{-1}Ax = P^{-1}b$ $P^{-1}A$ has N_x non-unit eigenvalues
- ► Richardson iteration: $Px^{k+1} = (P A)x^k + b$ Convergence rate bounded by $\alpha/(1 - \alpha)$
- ▶ Roundoff error: $\mathcal{O}(\epsilon \alpha^{-1})$ if ϵ is machine precision

Performance model vs time-serial method

$$(\beta_1 \mathbf{M} + \beta_2 \mathbf{K}) x = \tilde{b}$$

Serial:
$$(\beta_1 = 1/\Delta t, \beta_2 = \theta)$$

$$W_s = K_s M_s (N_x^q N_t)$$

$$P_s \sim N_x$$

$$T_s \sim \frac{W_s}{P_s} = K_s M_s N_t N_x^{q-1}$$

Parallel:
$$(\beta_1 = \lambda_1, \beta_2 = \lambda_2)$$

$$W_p = 2 \, K_p \, M_p \, \big(\, N_x^q \, \, N_t \big)$$

$$P_p \sim 2N_x N_t$$

$$T_p \sim \frac{W_p}{P_p} = K_p M_p N_x^{q-1} + T_c$$

Ideal speedup bound

Speedup:
$$S = \frac{T_s}{T_p} = \left(\frac{N_t}{\gamma \omega}\right) \frac{1}{1 + T_c/T_b}$$

"Difficulty" measures:
$$\gamma = \frac{K_p}{K_s}$$
, $\omega = \frac{M_p}{M_s}$

Block solve time:
$$T_b = K_p N_x^{q-1}$$

Communication time: T_c

Efficiency:
$$E = \frac{S}{P_p/P_s} = \frac{1}{2\gamma\omega} \frac{1}{1 + T_c/T_b}$$

Nonlinear all-at-once system

$$M\partial_t \boldsymbol{u} + \boldsymbol{f}(\boldsymbol{u}) = 0$$

$$\begin{pmatrix} \frac{1}{\Delta t} & & & \\ \frac{-1}{\Delta t} & \frac{1}{\Delta t} & & & \\ & \frac{-1}{\Delta t} & \frac{1}{\Delta t} & & \\ & & \frac{-1}{\Delta t} & \frac{1}{\Delta t} \end{pmatrix} \otimes \boldsymbol{M} + \begin{pmatrix} \theta & & & & \\ (1-\theta) & \theta & & & \\ & & (1-\theta) & \theta & \\ & & & (1-\theta) & \theta \end{pmatrix} \otimes \boldsymbol{I}_{x} \begin{pmatrix} \boldsymbol{f}(\boldsymbol{u}^{1}) \\ \boldsymbol{f}(\boldsymbol{u}^{2}) \\ \boldsymbol{f}(\boldsymbol{u}^{3}) \\ \boldsymbol{f}(\boldsymbol{u}^{4}) \end{pmatrix}$$

Nonlinear all-at-once Jacobian

$$\boldsymbol{M}\partial_{t}\boldsymbol{u} + \boldsymbol{f}(\boldsymbol{u}) = 0$$

$$\begin{pmatrix} \theta \nabla \boldsymbol{f}(\boldsymbol{u}^{1}) & & \\ (1-\theta) \nabla \boldsymbol{f}(\boldsymbol{u}^{1}) & \theta \nabla \boldsymbol{f}(\boldsymbol{u}^{2}) & \\ & (1-\theta) \nabla \boldsymbol{f}(\boldsymbol{u}^{2}) & \theta \nabla \boldsymbol{f}(\boldsymbol{u}^{3}) & \\ & & (1-\theta) \nabla \boldsymbol{f}(\boldsymbol{u}^{3}) & \theta \nabla \boldsymbol{f}(\boldsymbol{u}^{4}) \end{pmatrix}$$

Nonlinear all-at-once preconditioner

$$\mathbf{M}\partial_{t}\mathbf{u} + \mathbf{f}(\mathbf{u}) = 0$$

$$\begin{pmatrix} \frac{1}{\Delta t} & \frac{-\alpha}{\Delta t} \\ \frac{-1}{\Delta t} & \frac{1}{\Delta t} & \\ \frac{-1}{\Delta t} & \frac{1}{\Delta t} & \\ & \frac{-1}{\Delta t} & \frac{1}{\Delta t} \end{pmatrix} \otimes \mathbf{M} + \begin{pmatrix} \theta & \alpha(1-\theta) \\ (1-\theta) & \theta & \\ & (1-\theta) & \theta \\ & & (1-\theta) & \theta \end{pmatrix} \otimes \overline{\nabla \mathbf{f}(\mathbf{u})}$$

$$\overline{\nabla \mathbf{f}(\mathbf{u})} = \sum_{t=0}^{N_{t}} \frac{\nabla \mathbf{f}(\mathbf{u}^{n})}{N_{t}} \quad \text{or} \quad \nabla \mathbf{f}(\overline{\mathbf{u}}) = \nabla \mathbf{f}\left(\sum_{t=0}^{N_{t}} \frac{\mathbf{u}^{n}}{N_{t}}\right)$$

Nonlinear problems

Nonlinear system: $\mathbf{M}\partial_t u + \mathbf{f}(u)$

All-at-once system: $(\boldsymbol{B}_1 \otimes \boldsymbol{M}) \boldsymbol{u} + (\boldsymbol{B}_2 \otimes \boldsymbol{I}_x) \boldsymbol{F}(\boldsymbol{u})$

All-at-once Jacobian: $(\boldsymbol{B}_1 \otimes \boldsymbol{M}) + (\boldsymbol{B}_2 \otimes \boldsymbol{I}_{\times}) \nabla \boldsymbol{F}(\boldsymbol{u})$

ParaDiag Jacobian: $(C_1 \otimes M) + (C_2 \otimes \nabla f(\overline{u}))$

Time average: $\overline{\boldsymbol{u}} = \sum_{n=1}^{N_t} \boldsymbol{u}^n / N_t$

Outline

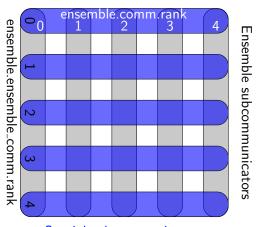
Parallel-in-time motivation

ParaDiag methods

asQ

- A library for implementing ParaDiag methods using Firedrake and PETSc
- General to any time-dependent PDE expressable in UFL
- Provides linear and nonlinear ParaDiag methods
- Parallelised in both time and space
- Aims to be flexible enough to be a sandbox, but performant enough to run "real" cases

Space-time parallelism: firedrake.Ensemble



Spatial subcommunicators

```
### === Solve the heat equation using ParaDiag === ###
from firedrake import *
import asQ
# Space-time partition and all-at-once function
partition = [2, 2, 2, 2]
ensemble = asQ.create_ensemble(partition)
mesh = UnitSquareMesh(32, 32, comm=ensemble.comm)
x, y = SpatialCoordinate(mesh)
V = FunctionSpace(mesh, "CG", degree=1)
ics = Function(V).project(sin(2*pi*x)*cos(pi*y)
aaofunc = asQ.AllAtOnceFunction(ensemble, partition, V)
aaofunc.assign(ics)
```

```
# Finite element forms
def form_mass(u, v): # M
    return inner(u, v)*dx
def form_function(u, v, t): # K
    return inner(grad(u), grad(v))*dx
bcs = [DirichletBC(V, 0, subdomain=1)]
dt = 0.1
theta = 1
aaoform = asQ.AllAtOnceForm(aaofunc, dt, theta,
                             form_mass, form_function,
                             bcs=bcs)
```

```
# ParaDiag solver and windowing
solver = asQ.AllAtOnceSolver(
    aaoform, aaofunc, jacobian_form=aaoform,
    solver_parameters=solver_parameters)
u = Function(V)
ofile = VTKFile(f"output{rank}.vtk", mesh.comm)
for i in range (nwindows):
    solver.solve(rhs=None)
    ofile.write(aaofunc[0])
    aaofunc.bcast_field(-1, u)
    aaofunc.assign(u)
```

```
solver_parameters = {
    "snes_type": "ksponly",
    "ksp_monitor": None,
    "ksp_type": "gmres",
    "ksp_rtol": 1e-8,
    "pc_type": "python",
    "pc_python_type": "asQ.CirculantPC",
    "circulant": {
        "alpha": 1e-4,
        "state": "initial".
        "block": {
            "mat_type": "aij",
            "ksp_type": "chebyshev",
            "ksp_max_it": 3,
            "pc_type": "ilu"}},
    "aaos_jacobian_state": "current"}
```