

Multiphysics timeintegration for turbulent combustion at the exascale

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11 July 2024

SIAM Annual Meeting (AN24)

Spokane, WA

### **Pele Combustion Suite**

**Pele** is the Exascale Computing Project's (ECP's) application suite for high-fidelity detailed simulations of **turbulent combustion.** Pele supports detailed physics and geometrical flexibility to evaluate design and operational characteristics of clean, efficient combustors for automotive, industrial, and aviation applications, including:

- Advanced internal combustion engines (e.g., RCCI)
- Novel supercritical CO<sub>2</sub> power cycles
- Rotating detonation engines
- Supersonic cavity flame holders
- Aviation combustors for sustainable drop-in JetA replacements

#### Pele combustion simulation and analysis suite:

- o *PeleC* (compressible)
- PeleLMeX (low Mach) replaces PeleLM
- *PelePhysics* (thermodynamics, transport, chemistry models)
- *PeleAnalysis* (in-situ, post-processing/analysis)
- *PeleMP* [multi-physics] (soot, radiation, Lagrangian sprays)
- o *PeleProduction* (collaboration hub)

#### https://github.com/AMReX-Combustion

#### Pele's KPP2 Combustion Challenge Problem: RCCI





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# PeleC

#### **PeleC: Compressible reacting flow solver**

- Conservation of species mass, momentum, total energy
- Embedded boundary capability for internal/external flows with complex geometries
- Time-explicit Runge-Kutta (RK)-based advance
  - Time-explicit RK variants for diffusion and hyperbolic: (PPM, PLM, WENO, MOL)
  - SUNDIALS-driven ODE integration for finite-rate chemical kinetics (CVODE, ARKODE)
- PelePhysics provides finite-rate chemistry models, equations of state and transport properties, non-ideal thermodynamics and chemistry modifications, and tabulated lookup table and neural-net models available
- PeleMP provides access to optional multiphase (spray) fuel models via AMReX particle capability, in addition to soot and radiation models



Rotating detonation engine Sreejith NA et al., early results, 2022



Quad fuel jet into shaped piston bowl Henry de Frahan et al., early results, 2022



# PeleLMeX

#### **PeleLMeX: Low Mach reacting flow solver**

- Conservation of species mass, momentum, enthalpy
- Embedded boundary capability for internal/external flows with complex geometries
- Iterative/implicit SDC-variants for time-stepping tightly coupled ADR systems resulting from large dt (enabled by the low Mach algorithm)
  - Semi-implicit (Crank-Nicolson) diffusion, monotonic Godunov hyperbolics
  - SUNDIALS-driven implicit ODE integration for finite-rate chemical kinetics
- PelePhysics provides finite-rate chemistry models, equations of state and transport properties. Tabulated lookup table and neural-net-based models available
- PeleMP provides access to optional multiphase (spray) fuel models (Lagrangian, AMReX particles based), moment-based soot models, and radiation transport
- Available LES models include Smagorinsky, WALE, and Sigma
- PeleLMeX's non-subcycled integrator supports AMR with closed-chamber pressurization due to fueling and heat release





Quad n-dodecane jets into KPP PB Wimer, Esclapez, et al., in prep, 2022

Lagrangian fuel sprays in PeleLMeX (droplets colored by T) Ariente et al., in prep, 2022

Lx = 11\*D

Gas turbine premixer, PeleLM

CICS Spring Technical Meeting, 2022

M. Vabre, B. Savard, et al.,

https://github.com/AMReX-Combustion/PeleLMeX

## PelePhysics

#### An open-source combustion physics library

- https://github.com/AMReX-Combustion/PelePhysics
- EOS: ideal gas mixtures (CHEMKIN), Soave-Redlich-Kwong (SRK), and EOS lookup tables and/or NN's
- Models and parameters for thermodynamics
- Mixture-averaged and unity Le transport properties, including extensions for non-ideal gases
- Finite-rate chemistry integration via SUNDIALS
- Python-based C++ generator, CEPTR, converts combustion chemistry models into production rate and reaction Jacobian code for CPU/GPU evaluation

Turbulent sCO<sub>2</sub> jet with SRK EOS [2]



## Pele Suite Exascale Applications

#### **Reacting Flows CFD Challenge:**

- CFD of real-world IC engines requires high parallelism and significant computational resources
- High-resolution reactivity-controlled combustion ignition (RCCI) engine
- Our exascale demo required over 20 billion grid cells and was scaled over 56,000 AMD MI250X GPUs using ORNL's Frontier supercomputer

#### **Chemistry Integration is EXPENSIVE!**

- The majority of the computational cost was used for the evaluation of the 53 species chemical reaction mechanism
- Runs of this scale become intractable for design studies, parameter variations, or other *outer-loop* investigations



#### High-Resolution Pele simulation of an RCCI engine

Nicholas T. Wimer, Lucas Esclapez, Nicholas Brunhart-Lupo, Marc Henry de Frahan, Mohammad Rahimi, Malik Hassanaly, Jon Rood, Shashank Yellapantula, Hariswaran Sitaraman, Bruce Perry, Michael Martin, Olga Doronina, Sreejith N. Appukuttan, Martin Rieth, and Marc Day

Phys. Rev. Fluids 8, 110511 - Published 17 November 2023

#### How can we speed up the chemical evaluator?

#### **Pele: Discretization Notes**

- Flames in low-speed flows (w/no detonations):  $(D + R)_{stiff}$  in a Lagrangian frame moving at speeds comparable to the flame speed,  $S_L$ 
  - ${\cal R}$  creates radicals,  ${\cal D}$  spreads them:  $au_{stiff} \sim 10^{-12} 10^{-7}$  s
  - Advective (e.g. turbulence),  ${\cal A}$  processes:  $au_{Adv} \geq 10^{-5}$  s
  - Pele codes explicitly timestep advection at  $\Delta t_{CFD}$  and advance  ${\cal S}_{chem}({\cal A},{\cal D})$  w/SUNDIALS
- Thermodynamic pressure stays spatially constant,  $p_{th} = p_{th}(t)$ 
  - Evolving flow satisfies  $\nabla \cdot U = S$ . This is a constraint on the velocity that is NOT automatically satisfied by conservation.
  - Our finite-volume scheme advects mass, energy with predicted face- and time-centered centered velocities that satisfy the divergence constraint

### PeleC: Compressible Conservation Equations (PeleLMeX similar, but low Mach)

$$\begin{aligned} \frac{\partial U}{\partial t} + \nabla \cdot \Gamma &= S \end{aligned} \qquad U = \left[\rho, \ \rho \vec{v}, \ \rho E, \ \rho \mathcal{Y}_k, \ \rho \mathcal{A}_\ell, \ \rho \mathcal{B}_m\right] \qquad \text{Chemistry sources} \\ \Gamma &= \left[\rho \vec{v}, \ \rho \vec{v} \vec{v} + P, \ \rho E \vec{v} + P \cdot \vec{v} + \mathcal{Q} + \mathcal{Q}_{rad}, \ \rho \mathcal{Y}_k \vec{v} + \mathcal{F}_k, \ \rho \mathcal{A}_\ell \vec{v}, \ \rho \mathcal{B}_m \vec{v}\right] \\ S &= \left[S_\rho, \ \rho \vec{g} + S_{\rho \vec{v}}, \ \rho \vec{v} \cdot \vec{g} + S_{\rho E}, \ \rho \dot{\omega}_k + S_{\rho \mathcal{Y}_k}, \ S_{\rho \mathcal{A}_\ell}, \ S_{\rho \mathcal{B}_m}\right] \\ P &= p\mathcal{I} + \Pi, \ (\mathcal{F}_k, \Pi, \mathcal{Q}): \ diffusive \ transport, \ S_*: \ external \ sources \ (e.g. \ evaporated \ fuels) \end{aligned}$$



Finite Volume Discretization with AMR

$$\frac{\partial}{\partial t} \iiint_{\Omega} U dV + \iint_{\partial \Omega} \Gamma \cdot dA = \iiint_{\Omega} S dV$$

$$\rightarrow \quad \frac{\partial \mathcal{U}}{\partial t} = F_{AD} + \mathcal{S}_{chem}$$

### PeleC Time Advance (PeleLMeX similar w/projections, implicit diffusion)

Second-order predictor-corrector split time-advance

•  $\mathcal{U}' = \mathcal{U}^n + \Delta t \left( F_{AD} \left( \mathcal{U}^n \right) + \mathcal{S}^n_{chem} \right)$ 

• 
$$\mathcal{U}'' = \mathcal{U}^n + \frac{\Delta t}{2} \left( F_{AD} \left( \mathcal{U}^n \right) + F_{AD} \left( \mathcal{U}' \right) \right) + \Delta t \cdot \mathcal{S}_{chem}^n$$

• 
$$F = \frac{1}{2} \left( F_{AD} \left( \mathcal{U}^n \right) + F_{AD} \left( \mathcal{U}'' \right) \right)$$

- $\mathcal{S}_{chem}^{n+1} = \mathcal{S}(\mathcal{U}^n, F)$  Local, provided by SUNDIALS
- $\mathcal{U}^{n+1} = \mathcal{U}^n + \Delta t \left( F + \mathcal{S}_{chem}^{n+1} \right)$

#### Pele -> SUNDIALS

Pele provides the following to SUNDIALS:

• 
$$\mathcal{S}_{chem,k} = \rho \dot{\omega}_k = \sum_{n=1}^{N_R} \left( K_{Fn} \prod_{j=1}^{N_s} [X_j]^{\nu'_{kj}} - K_{Rn} \prod_{j=1}^{N_s} [X_j]^{\nu''_{kj}} \right)$$

State component scales via "typical values" (component-wise absolute and relative tolerance)

$$a_{\text{tol},i} = a_{\text{tol}} \cdot \max\left(\frac{1}{2}(\min(y_i) + \max(y_i)), \tilde{y}_i\right)$$
  
of for implicit solves (the usual case):

- ...an
- Jacobian,  $\mathcal{J}_{ki} = \frac{\partial \mathcal{S}_{chem,k}}{\partial (\rho \mathcal{Y}_i)}$
- Preconditioner (approximate  $\mathcal{J}_{ki}^{-1}$ )





NOTE: Data ordering impacts sparsity pattern for Jacobian

## Done. Right?

Issues:

- 1. In combustion, we are changing fuels all the time. New models require new code for the production rates, Jacobians, preconditioners, etc. Producing these routines by hand is tedious and error-prone
- 2. The robustness of Newton-Krylov schemes depends heavily on the quality of the preconditioner used. Forming a good preconditioner requires much of the same infrastructure as forming the Jacobian itself.
- 3. The QSSA (discussed next) provides a rigorous approach to reducing the size of models, at the expense of dramatically complicating the coding tasks
- 4. We have found numerical Jacobians are a viable path in general, but they tend to be extremely sensitive to variable scaling issues, and difficult to manage generically

### Quasi-Steady-State Approximation (QSSA)

- Chemical evolution timescales can vary many orders of magnitude across species
  - Timescales of fuel consumption, product generation typically match those of the flow
  - Production/destruction of intermediates can be ten (or more) orders of magnitude faster
  - Robust/accurate integration of such stiff systems is not practical
- The Quasi-Steady State Approximation (QSSA) assumes species, K, with short characteristic timescales and low concentrations at always at steady-state:  $S_{chem,K} \approx 0$ 
  - QSSA species are NOT transported in Pele
  - Cellwise, QSSA species concentrations become algebraic expressions/constraints as a function of the rest of the state as it evolves
- With QSSA the size of the chemical reaction mechanism (number of species and reactions) and stiffness can be **drastically reduced**
- "Science" Research Questions:
  - What is the benefit of explicitly forming the Jacobian (vs using a Newton-Krylov type strategy)?
  - What is the impact of the QSSA on the accuracy of predictions in our demo problem?

### **CEPTR Jacobian Implementation**

- Chemical mechanisms are large and stiff systems
- Integration is typically implicit by means of a reaction Jacobian

$$J_{i,j} = \left(\frac{\partial \dot{\omega_i}}{\partial C_j}\right)_{CQSS,k} + \left(\frac{\partial \dot{\omega}}{\partial C_{QSS,k}}\right)_{C_j} \frac{dC_{QSS,k}}{dC_j}$$

expressions

The second term is difficult due to

chain rule terms and long

- Non-QSSA mechanisms can simply accumulate contributions to the Jacobian one reaction at a time
- With QSSA, there is a set of **non-QSSA species** (transported) and a set of **QSSA species** (algebraically related to the transported species)
- Use SymPy to symbolically track the full derivatives of each of the QSS and non-QSS species interactions
- The final Jacobian contains many repeated calculations from the chain rule nature of interactions



### **Computational Setup**

- We set up a **multi-pulse n-dodecane jet** that is reminiscent of the RCCI conditions
- Skeletal mechanism is 53 species and QSSA mechanism is 35 species
- Domain is filled with **methane/air** mixture at **900 K** and 60 atm
- Fuel jet is n-dodecane at 470 K
- Two pulses of fuel each for 0.5 ms with a 0.5 ms dwell time in between

#### Mechanism Sweep:

- Skeletal mechanism using traditional Newton-Krylov method
- Reduced QSS mechanism using traditional NK
- Reduced QSS mechanism using analytic Jacobian formulation
  - 96 different iterations on the pre-computed terms for memory optimization
- Approximately 150 million grid cells



# **Qualitative Comparisons**

### time: 0.75 ms



## **Qualitative Comparisons**

#### time: 0.75 ms



## Volume Average Comparisons



- Temperature and Density are very consistent
- Heat Release Rate shows the largest differences with QSS NK as the main outlier
- The QSS with analytical Jacobian formulation is in close agreement to the Skeletal mechanism

# **Chemical Mechanism Optimizations**



### **Relative Performance**



- **Consistent results** over a wide range of precomputations
- Therefore, we can make liberal **GPU optimizations without sacrificing performance** of the chemical mechanism evaluation

### Strong Scaling of Chemistry Evaluation

- The simulation was run on a range of nodes from 8 – 128 (64 – 1024 GPUs)
- The evaluation of the reaction mechanism timing is consistent with the expected ideal
- The reaction mechanism formulations do not impact the scalability of the Pele suite



Scaling of Mechanism Evaluation

## Conclusions

- Implementation of automated QSSA mechanism generator
  - Using CEPTR in PelePhysics
  - We formulate an entirely symbolic analytic Jacobian using SymPy/SymEngine
  - We implement a series of optimizations to reduce memory pressure for GPU optimizations with little/no penalty in computational cost
- Evaluated on multi-pulse, high-pressure jet
  - Approximately 4x speed improvement with respect to skeletal and 2x with respect to Newton-Krylov
  - Best results between full skeletal mechanism and QSS analytic Jacobian
  - Improved stability and runtime with analytic Jacobian (noted across a broad range of models/configurations now)



# Thank you!

#### www.nrel.gov

NREL/PR-2C00-90542

#### EXASCALE COMPUTING PROJECT

This work was authored by the National Renewable Energy Laboratory, operated by Alliance for Sustainable Energy, LLC, for the U.S. Department of Energy (DOE) under Contract No. DE-AC36-08GO28308. Funding provided by U.S. Department of Energy Office of Science and National Nuclear Security Administration. The views expressed in the article do not necessarily represent the views of the DOE or the U.S. Government. The U.S. Government retains and the publisher, by accepting the article for publication, acknowledges that the U.S. Government retains a nonexclusive, paid-up, irrevocable, worldwide license to publish or reproduce the published form of this work, or allow others to do so, for U.S. Government purposes. This research was supported by the Exascale Computing Project (17-SC-20-SC), a collaborative effort of the U.S. Department of Energy Office of Science and the National Nuclear Security Administration. A portion of the research was performed using computational resources sponsored by the Department of Energy's Office of Energy Efficiency and Renewable Energy and located at the National Renewable Energy Laboratory. This research used resources of the Oak Ridge Leadership Computing Facility, which is a DOE Office of Science User Facility supported under Contract DE-AC05-000R22725.

