

Partnership for Advanced Combustion Engines (PACE) – A Light-Duty National Laboratory Combustion Consortium

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Overview

Timeline

5-year program

- Start Date: FY19 Q3
- End date: FY23 Q4

• Percent complete: ~46%

US fiscal years run from October 1 through September 30

Budget

- Total PACE FY21 budget \$8,920K
- Approximate allocations
 - Knock/LSPI mitigation ~ 30%
 - Dilute combustion ~ 30%
 - Emissions reduction ~ 40%

Budget breakdown by task provided in reviewer only slides

Barriers and Technical Targets*

- Development of enhanced understanding and predictive models to address:
 - Limits to SI engine efficiency (knock/LSPI)
 - Barriers to highly dilute combustion
 - Emissions reduction (cold start)
 - * Aligned with USDRIVE ACEC Tech Team Priority 1: Dilute gasoline comb.

Partners

- ANL, LLNL, NREL, ORNL, SNL
- USDRIVE ACEC Tech Team

Project leads

- I. Ekoto (SNL)
- S. Som (ANL)
- M. McNenly (LLNL) J. Szybist (ORNL)



Approximately 400 million light-duty vehicles with ICEs will be sold in the US between now and 2050 (2.4 billion worldwide)

- Improving ICE efficiency is a critical element of a path toward lower petroleum consumption and greenhouse gas emissions
- Tailpipe pollutant emissions can be reduced to near zero – alleviating urban environmental concerns
- US competitiveness secures over 900K manufacturing jobs (Mar. 2021 -U.S. Bureau of Labor Statistics)





Relevance | Improved ICEs are key to energy and environmental security

Our end-goal will be reached through progress in three key areas:

- Knock and LSPI mitigation for stoichiometric SI engines
- Improving stability & efficiency of highly dilute combustion
- Reducing emissions to a zero-impact level (focus on cold-start)



Relevance | Improved ICEs are key to energy and environmental security

Maximize conventiona SI efficiency & power density Unlock highefficiency dilute combustion

- Improved knock and LSPI control
 - Near-term benefit with potential for > 5% efficiency improvement

- Highest potential efficiency gains but significant barriers
 - Mid-term, high EGR stoichiometric for > 12% efficiency gain



- Deeper understanding of cold start physics and chemistry enables numerically-aided design and calibration
 - Tier 3 Bin 20 emissions levels and beyond

Approach | Research is aligned with industry priorities and workflows

- Strong alignment with USDRIVE ACEC Tech Team
- Ab initio engagement on plans and <u>success measures</u>







Approach | Leverages DOE investment in HPC and ML/AI

• HPC will be used as a microscope – illuminating processes that are inaccessible to experimentation

Source: ANL

 HPC simulations provide a benchmark for accuracy of engineering simulations

- Machine learning and pattern recognition will be applied to
 - Resolve decades-old problems (*e.g.* root causes of cyclic variability)
 - Detect and mitigate abnormal combustion (instability, knock, LSPI, misfire)
 - Develop expert systems enabling optimal CFD-based design
 - Develop data-driven efficient sub-models





Approach | All tasks support 7 Major Outcomes



Improved knock and LSPI control

Major Outcome 1: Models for combustion system analysis accurately predict knock response to design changes

<u>Success measure</u>: Simulation of changes in engine geometry or operating conditions predicts the change in KLSA within 1 degree over the knock limited operating range of the engine, with a 5X reduction in simulation time.

Major Outcome 2: Data analytics enable operation and real-time control to mitigate knock/LSPI

Success measure: before knock occur Similar machine learning tasks for Dilute and Cold Start pillars leading to GO/NO-GO for FY23 engine validation

least 10 CAD

Major Outcome 3: Develop new multi-step phenomenological mechanism for LSPI that captures wallwetting, lubricant, and geometry effects

<u>Success measure</u>: Phenomenological model captures relevant physical causes of preignition and demonstrates pathways to reduce the occurrence of LSPI by 50%.

Major Outcome 4: Improved high-load igniter performance and igniter durability enabled by predictive modeling

<u>Success measure</u>: An ignition model including spark-plug geometry and electrical discharge circuit details predicts spark stretch, blowouts and restrikes, and flame initiation, propagation and quenching. The model can accurately predict 0-5% mass burned fraction within 10% (mean and standard deviation).



Approach | All tasks support 7 Major Outcomes



High-efficiency, low-variability dilute combustion systems

Major Outcome 5: Major Outcome 5: Homogeneous and stratified lean/dilute engine efficiency and emissions are accurately predicted

<u>Success measure</u>: Validated simulations predict the change in burn duration and COV (to within 10%) relative to a baseline configuration for a change in engine design at 30% dilution and ACEC 3 bar/1300 rpm test point.

Major Outcome 6: Develop viable advanced igniters and control methods that expand existing dilution limits and enable stable catalyst heating operation

<u>Success measure</u>: Prototype igniters and control strategies ignition control methods enable stable ignition for EGR dilution rates of up to 40% or air dilution rates of up to 50% with no adverse impact on pollutant emissions relative to the stock OEM configuration. Demonstrate ignition system can maintain stable combustion at high exhaust heat flux conditions seen during cold start. ACEC 3 bar/1300 rpm test point.

Major Outcome 7: ignition combustion Success measure: Co and a pathway identifi

PACE sharpens research focus on tasks with greatest relevance to the light-duty OEMs

pression

100 N-m-s⁻¹ erating points



Approach | All tasks support 7 Major Outcomes



Cold-start design and calibration capability

Major Outcome 8: Validated cold start modeling capability that accurately predicts injection and spark timing trends on:

- Cold-start engine-out emissions
- Combustion phasing and stability
- Exhaust heat losses and oxidation (accurate catalyst feed-gas enthalpy and composition)

Success measures:

- > 80% accuracy in predicting engine-out emissions and stability for nominal conditions (relative error in emission level < 20%)
- ACEC cold start protocol COV must be less than 20% (> 80% accuracy)
- > 80% accuracy predicting feed-gas emissions and stability for operating conditions matched to PACE cold-start experimental data set
- > 90% accuracy in predicting heat losses in hot end exhaust for varying heat flux conditions



Severe Acute Respiratory Syndrome coronavirus 2 (SARS-CoV-2)



Photo Credit: A. Eckert, MSMI, D. Higgins, MAMS

- All national labs shutdown March 2020, except ORNL
- Supercomputing resources diverted to emergency COVID-19 projects.
- Experimental facilities started returning to service Summer 2020 (except ANL APS and ORNL SNS).
- Simulation and theory teams at ANL, LLNL, NREL, ORNL, and SNL are still remotely working.
- Most milestones were delivered on time, while nearly all those that were late were delayed only 3 months (or less).

True testament of the ingenuity and team spirit of the PIs to adapt to the new reality, reprioritize tasks, and continue steady progress on the PACE objectives



Accomplishment | PACE 2.3L common platform engine purchase complete

- PACE engine contract awarded to AVL Sept. 2020 for two engines
- Common platform designed to match a production 2.3L turbocharged engine as closely as possible
 - ANL metal single-cylinder engine
 - SNL optical single-cylinder engine
- AVL delivery on-track for Nov. 2021
- ORNL to install two similar engines
 - single cylinder with multi-cylinder block (on-track for Q3FY21)
 - multi-cylinder for cold-start tests

273 C



AVL thermal design captures multi-cylinder wall temperatures





| metric | RD5-87(#2A) | PACE-1 | PACE-20 | EPA Tier III |
|----------------------------------|-------------|--------|---------|-----------------|
| RON ¹ | 92.3 | 91.8 | 92.1 | 92.3 |
| MON ¹ | 84.6 | 82.3 | 84.5 | 84.5 |
| H:C ¹ | 1.98 | 1.97 | 1.97 | 1.96 |
| density [g/mL] ¹ | 0.75 | 0.75 | 0.74 | 0.75 |
| PMI ² | 1.68 | 1.56 | 1.50 | |
| | D86 | D86 | D86 | D86 |
| T ₁₀ [C] ¹ | 57.8 | 60.4 | 57.9 | 52.7 |
| T ₅₀ [C] ¹ | 101.3 | 100.6 | 89.9 | 91.2 |
| T ₉₀ [C] ¹ | 157.9 | 165.8 | 166.0 | 159.5 |

¹ Measurements from SWRI and Gage (ASTM D2699, D2700, D5291, D4052, D86)

² Calculated from DHA (ASTM D6729)

- Extensive validation completed since AMR20:
 - low and intermediate heat release in engine and RCM (ace139 Wagnon)
 - free spray plumes (ace143 Powell)
 - spray wall interaction (ace144 Pickett)
 - spray collapse in engines (ace167 Torelli)
 - soot volume fraction (ace168 Manin)
 - soot pyrolysis (ace168 Manin)
 - flame speed (ace139 Wagnon)
 - improved chemistry (ace139 Wagnon)
- 15 barrels (54 gal.) ordered from Gage
- 7 barrels of PACE-20 BOB also ordered



Accomplishment | Discontinued FY20 project to bolster investment in 4 areas

Focused PACE on two DOE Exascale Computing Program codes leveraging \$18M ASCR investment in NEK5000 (\$8M) and Pele (\$10M)

Dedicated CFD integration tasks for Major Outcomes 1 and 5 to quantify new models:

- spray-wall

- spray nozzle

- ignition
- flame quenching
- turbulent-chemistry emissions
- conj. heat transfer algorithmic speedup

Securing common set high-pressure injectors

- 25 Bosch Motor Sports injectors (500 bar)
- complete laser light sheet patternation data
- best match will be the primary set with the rest serving as spares for the program

Increased Spray Simulation Task Support

- full accuracy test 17 free spray conds.
- spray w/cross flow vaporization in VOF
- near-field plume corrections
- non-flashing vaporization
- discrete particle to Eulerian for wall films

Built new optimization tools for accurate and smaller reduced mechanisms

- Eliminates main human-in-the-loop iteration
- Easily optimize over 80 reactions (simult.)
- 10⁸ ignition calculations used in a month to reduce PACE-20 model to 315 species



Accomplishment | 1 month with new tools is set to save an expected 25% in CFD costs





Accomplishment | Baseline CFD performance established on success measures



PA

Response to last year's reviewer comments (wt. avg. = 3.35/4)

• **Q1 Reviewer 5:** "Delivering a usable, open-source product to OEMs is a key target... consider implementing technology transfer metrics into the program to track adoption."

The PACE leadership team has developed a "modeling menu" for the OEMs as part of our ongoing collaboration with the USDRIVE ACEC Tech Team. This document describes the features of more than 35 current sub-models, fuel mechanisms, tools, and best practices created or under development in the program that are available to the OEM and academic partners through PACE. It is a good idea to augment this document to include usage statistics by the OEMs for internal analysis and future reviews.

• Q1 Reviewer 5: "[T]here will need to be close connection to, support from, and endorsement by engine manufacturers that the technologies being developed are of interest and have the potential to be commercially viable. Especially for novel combustion programs, such as Major Outcomes 5 and 7"

PACE removed Major Outcome 7 (low temperature gasoline combustion) and the lean ignition research in Major Outcome 5 to further increase the program's focus and speed development on the highest impact barriers for light-duty engines. The PACE leadership team is also actively helping the OEMs in the USDRIVE ACEC Tech Team to create the next roadmap extension for light-duty vehicle research.

• Q1 Reviewer 5: "While recognizing that the move to a single engine platform may increase the speed to deliver modeling advancements, this reviewer was concerned that limiting to one engine design will impact the ability to deliver model validation. Improved, validated, modeling tools should be engine-agnostic and validated across a range of different engine platforms"

The PACE program will continue with two engines in FY22 at ANL, SNL and ORNL: the existing SG2, and the new PACE common platform. We agree that the modeling advancements should be independent of the engine. However, the benefit of learning simultaneously about sub-model performance on different experimental campaigns is too great given our mandate to move our research faster on a limited budget. The new platform will eliminate uncertainties attributed to different geometries and will increase confidence in the relevance of the models by predicting performance in a modern spark-ignition engine. Further, the success measures for Major Outcomes 1, 5, and 8 are for a blind prediction of a substantial design change, which forces the models to be as engine-independent as possible.



Program-Level Collaborations

- USDRIVE ACEC Tech Team invited PACE leadership to help create the next extension to the light-duty vehicle research roadmap
- Advanced Engine Combustion MOU partners
 - Light-duty OEMs Ford, General Motors, Stellantis (formerly FCA)
 - Heavy-duty OEMs Caterpillar, Cummins, Daimler, GE, John Deere, Navistar, PACCAR, Progress Rail, Volvo, Wabtec
 - Energy companies BP America, Chevron, ExxonMobil, Marathon, Phillips 66, Shell
 Commercial CFD ANSYS, Convergent Science, Gamma Tech., Siemens
- Overlap and synergies with Co-Optima on a project level
- More than 50 additional project-level collaborations with U.S. and international universities and private/public research institutions



Added FY20

Project-Level Collaborations

Manufacturers

Ford (2) General Motors (5) Hyundai Stellantis Toyota

Suppliers

Bosch Bosch Motor Sports Delphi Hyundai KEFICO Mahle Tenneco Transient Plasma Sys.

Software Vendors

ANSYS AVL FIRE Convergent Science (6) Gamma Tech (2) Pointwise Siemens/StarCCM+

Fuels & Lubricants

Aramco Chevron Lubrizol

Research Councils

Comp. Chem. Consort. Coordinating Research Council IFP Energies Nouvelles Inst. Motori (STEMS)

Universities

AIST, Japan Auburn City U. London CMT U. Polit. Valencia Colorado St. Imperial College lowa St. LTT Erlangen KAUST (3) Marquette U. Michigan Tech. Natl. U. Ireland – Galway Polit. Milano

Princeton **RWTH Aachen** TU Darmstadt UC Davis U. Central Florida U. Connecticut U. Hawaii U. Illinois U-C U. Mass. Amherst U. Mass Dartmouth U. Sci. Tech. of China U. Southern Calif. U. Texas

Government

Army Research Laboratory



Remaining Challenges and Barriers

- Coordinate PACE advances for +20 sub-models while evaluating their performance against four Major Outcomes (1, 4, 5 & 8) and prioritizing their future development with limited time and money.
- Leverage insight across different experimental campaigns simultaneously.
- Ensure that the proposed research can deliver significant efficiency gains (CO₂ savings), and emissions reductions.

• Accelerate knowledge and tool transfer to industry including the engine design workflows for the production engineers.

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Future Work | Any proposed future work is subject to change based on funding levels

- CFD Integration tasks for high-load knock (MO #1), dilute cyclic variability (MO #5), and cold start (MO #8) will report first integrated comparison of new PACE models at August AEC meeting
- Common PACE engines will be installed at ANL, SNL and ORNL
- PACE leadership team will help the OEMs develop the next ACEC Tech Team roadmap over the summer
- PACE simulation experts continue to update the public "modeling menu," which contains +35 sub-models, mechanisms, tools, and best practices
 - and create a public website to connect stakeholders with the resources
- Zero-RK 3.0 release (planned for Aug. 2021) will include: simplified interface to link with CFD codes, mechanism optimization tools, counterflow diffusion flames, PSR, and IC engine models

- and create common flame tables for the recommended PACE-20 mechanism

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Summary

Relevance

- ICE powered vehicles will be a significant component of the US fleet for many decades
- Significant improvements in both emissions and efficiency are possible and needed to meet environmental goals

Approach

- The work plan for PACE is focused on three key areas and is developed in coordination with the USDRIVE ACEC Tech Team
- Coordinated collaborations across 7 Major Outcomes using kinetics, fundamental measurements and engine experiments feeding into improved models

Programmatic Accomplishments

- Purchased common PACE engine platform single cylinder at ANL (metal) and SNL (optical), and single and multi at ORNL
- Recommended PACE-20 surrogate with more than a dozen new validation experiments
- Discontinued FY20 project to bolster investment in 4 areas: CFD integration tasks, spray models, new 500 bar injectors, and mechanism reduction tools
- Baseline CFD performance established on success measures for Major Outcomes 1, 5, and 8

Collaboration and Coordination

- PACE is a collaboration among five national laboratories working towards common objectives
- US DRIVE ACEC Tech Team, AEC MOU industry stakeholders, Commercial CFD, Co-Optima
- More than 50 project-level collaborations with U.S. and international universities and private/public research institutions

Proposed Future Research (any proposed future work is subject to change based on funding levels)

- Promoting closer coordination and increased relevance with a common, modern engine platform and gasoline surrogate
- Enhancing industry adoption of software and modeling tools by continually refining the hand-off with the ACEC tech team
- Building partnerships with other DOE offices on HPC, ML/AI, and Basic Energy Science

Technical Back-up Slides



24

Complete PACE Budget FY21

| Code and Work Flow Development | | | | |
|---|------|------------|------|------|
| | Lab | PI | FY20 | FY21 |
| A.M.05.01 Spray and Combustion model implementation | ANL | Ameen | 350k | 340k |
| A.M.05.02 Gridding, validation, and workflow development | ANL | Ameen | 350k | 300k |
| A.M.05.04 MO1 Integration | ANL | Som | | 75k |
| A.M.05.05 MO5 Integration | ANL | Scarcelli | | 75k |
| L.M.05.01 Accelerated multi-species transport in engine simulations | LLNL | Whitesides | 275k | 250k |
| L.M.05.02 Improved chemistry solver performance with machine learning | LLNL | Whitesides | | 250k |
| L.M.05.04 Scalable performance and CFD integration of ZERO-RK | LLNL | Whitesides | 275k | 75k |
| L.M.05.06 Mechanism Reduction | LLNL | Whitesides | | 75k |
| L.M.05.07 Accelerate Mechanism Reduction Tools | LLNL | Whitesides | | 75k |
| | | | | |
| Cold Start | Lab | PI | FY20 | FY21 |
| O.E.07 Multi-cyl Cold Start & surrogate testing | ORNL | Curran | 350k | 350k |
| S.E.07 Engine experiments characterizing wall films & PM formation | SNL | Sjoberg | 270k | 270k |
| | | | | |
| Compustion and Emissions | | | | |
| | Lab | PI | FY20 | FY21 |
| O.E.02 Effectiveness of EGR to mitigate knock throughout PT domain | ORNL | Szvbist | 220k | 175k |
| S.E.02 Experiments supporting particulate modeling wall film & pyrolysis | SNL | Manin | 500k | 400k |
| L.M.01 Improved Kinetics for Ignition Applications | LLNL | Pitz | 150k | |
| S.M.02.01 DNS and modeling of turbulent flame propagation & end gas ignition | SNL | Chen | 50k | 50k |
| S.M.02.02 Flame wall interactions | SNL | Nguyen | 150k | 50k |
| S.M.02.03 Engineering PAH Model Development | SNL | Hansen | | 100k |
| | 1 | | | |
| Data Analytics | | | | |
| Data Analytics | Lab | PI | FY20 | FY21 |
| O.E.08 Machine Learning and Nonlinear Dynamics | ORNL | Kaul | 200k | 200k |
| | | | | |
| Flows and Heat Transfer | | | | |
| | Lab | PI | FY20 | FY21 |
| O.E.06.01 Neutron diffraction for in situ measurements in an operating engine | ORNL | Wissink | 100k | 100k |
| O.E.06.02 Neutron Imaging of Advanced Combustion Technologies | ORNL | Wissink | 200k | 200k |
| O.M.06 Conjugate heat transfer | ORNL | Edwards | 350k | 350k |
| LA.M.06.01 Heat Transfer through Engine Metal | LANL | Carrington | 200k | |
| LA.M.06.02 Heat Mass Transfer in Liquid Species | LANL | Carrington | 200k | |
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Response to last year's reviewer comments (wt. avg. = 3.35/4)

• Q3 Reviewer 4: "With regards to cold-start emissions, the reviewer would have liked to see more information on how PACE will collaborate with aftertreatment research teams"

The PACE leadership and the CLEERS (Crosscut Lean Exhaust Emissions Reduction Simulations) leadership have started this year coordinating with the OEM members of the USDRIVE ACEC Tech Team. The goals are to define the simulation outputs needed from PACE and the the aftertreatment model in order to predict tailpipe emission.

• Q4 Reviewer 3: "... the major outcomes are not overly specific as to what final success looks like... More clarity is necessary to define a quantifiable result state, which could be considered a success"

All the major outcomes have a success measure that can be computed and validated against engine experiments. The success measures for Major Outcomes 1, 4, 5, and 8 are based on a priori prediction accuracy of an engine design change (geometry or operating conditions). Major Outcomes 2, 3, and 6, have success measures based on the experimental demonstration of a specific aspect of engine performance. In all cases, the quantifiable level to declare success is based on detailed feedback from the ACEC Tech Team each year.



Response to last year's reviewer comments (wt. avg. = 3.35/4)

• **Q4 Reviewer 4:** "The use of artificial intelligence (AI) and machine learning (ML) should be handled carefully... Machine learning should be used to handle turnaround time."

The ML based models developed at NREL have been designed to complement the physics-based models. The starting point for all the ML based models are state-of-the-art (SOA) physics-based models and are designed to reproduce the asymptotic behavior of physics-based models. We also tightly couple various physics-based realizability conditions during the ML model development process. In terms of testing to establish confidence in these ML based models, the team at NREL has been performing blind validation tests against the DNS data and in FY21-FY22 we plan to perform a-posteriori tests against experimental data. Additionally, these ML based models are also designed to efficiently utilize higher computing power afforded by the latest supercomputing technologies.

• Q4 Reviewer 5: "the reviewer still would have liked to see some additional information on the codes that will be used for the direct numerical simulations (DNS). Why are these codes needed if the ultimate goal is to get the models in commercial codes used by the OEMs?"

There are two ASCR-funded codes in the Exascale Computing Project that are used in the PACE program - NEK5000 and Pele. NEK5000 is needed to provide the highest fidelity turbulence resolution of all the in-cylinder flow processes (both non-reacting and reacting) using high-order, large eddy simulation (LES). The availability of such resolution allows the CFD integration tasks for Major Outcome 1 (high load knock) and Major Outcome 5 (dilute cyclic variability) to establish the best performance available on the success measures, and to isolate numerical errors in the flow-field coupling versus the standalone sub-models for other physical processes. Pele is capable of performing direct numerical simulation (DNS) of reacting flows but cannot simulate the entire cylinder during the full reacting cycle even with a complete takeover of the DOE's largest supercomputer. As such, it will be used as a microscope to investigate fundamental turbulent-combustion processes not currently resolved in engineering models with the aim of developing better sub-models for early flame kernel development, end-gas auto-ignition, and flame-wall interactions.



Abbreviations

| ACEC | Advanced Combustion & Emissions Control | GE | General Electric |
|---------|--|------|--|
| AEC MOU | Advanced Engine Combustion Memorandum of | HPC | High Performance Computing |
| | Understanding | ICE | Internal Combustion Engine |
| AI | Artificial Intelligence | KLSA | Knock Limited Spark Advance |
| AMR | Annual Merit Review | LLNL | Lawrence Livermore National Laboratory |
| ANL | Argonne National Laboratory | LSPI | Low Speed Pre-Ignition |
| APS | Advanced Photon Source | ML | Machine Learning |
| ASCR | Advanced Scientific Computing Research | MO | Major Outcome |
| AVL | Anstalt für Verbrennungskraftmaschinen List GmbH | NREL | National Renewable Energy Laboratory |
| BOB | Blendstock for Oxygenate Blending | OEM | Original Equipment Manufacturer |
| CA/CAD | Crank Angle in Degrees | ORNL | Oak Ridge National Laboratory |
| CFD | Computational Fluid Dynamics | PACE | Partnership for Advanced Combustion Engine |
| COV | Coefficient Of Variation | PI | Principal Investigator |
| CPU | Computer Processing Unit | RCM | Rapid Compression Machine |
| DOE | Department of Energy | SNL | Sandia National Laboratories |
| EGR | Exhaust Gas Recirculation | SNS | Spallation Neutron Source |
| EIA | U.S. Energy Information Administration | UDF | User Defined Function |
| EPA | Environmental Protection Agency | VOF | Volume of Fluid |
| FY | Fiscal Year | VTO | Vehicle Technologies Office |

