

**14th ETH-Conference on Combustion Generated Nanoparticles
August 1st – 4th 2010**

Paper/Poster-Abstract Form

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Title: Modelling soot formation in a Direct Injection Spark Ignition engine

Abstract: (min. 300 – max. 500 words)

The abstracts for papers and posters must contain unpublished information on the research subject: background, investigation methods, results and conclusions. Graphs and references are very welcome. Acronyms should be avoided. Abstracts with < 300 words can not be considered.

In this work, the formation of soot in a Direct Injection Spark Ignition (DISI) engine is simulated using the Stochastic Reactor Model (SRM) engine code. Turbulent mixing, convective heat transfer, direct injection and flame propagation are accounted for. In order to simulate flame propagation, the cylinder is divided into an unburned, entrained and burned zone, with the rate of entrainment being governed by empirical equations but combustion modelled with chemical kinetics. The model contains a detailed chemical mechanism as well as a highly detailed soot formation model, however computation times are relatively short. The soot model provides information on the morphology and chemical composition of soot aggregates along with bulk quantities, including soot mass, number density, volume fraction and surface area. The model is first calibrated by simulating experimental data from a Gasoline Direct Injection (GDI) Spark Ignition (SI) engine. The model is then used to simulate experimental data from the literature, where the numbers, sizes and derived mass particulate emissions from a 1.83 L, 4-cylinder, 4 valve production DISI engine were measured in the exhaust gas. Experimental results from different injection and spark timings are compared with the model, which is capable of reproducing qualitative trends in aggregate size distribution and emissions.

Secondly, we use this example of DISI soot modelling in order to illustrate more generally what can be achieved with present modelling approaches and what the limitations are. We discuss the role experimental data plays in the process of building models and propose a standardised, systematic way of storing and processing data. We emphasise in particular the importance of accounting for uncertainties in measurements and model parameters. We then demonstrate how such an infrastructure can be applied to quantitatively assess an empirical soot model against a large experimental database, highlighting potential model shortcomings and outliers in the data.

Short CV of presenter:

Amit Bhave is presently a Fellow at Hughes Hall, Cambridge and an Affiliate Research Fellow at the CoMo Group, Department of Chemical Engineering & Biotechnology, Cambridge. Amit's research interests include numerical modelling, low-emission combustion engines, chemical reactor design, and technology commercialisation. Amit completed his PhD at Cambridge and has Bachelors and Masters Degrees in Chemical Engineering. As the CEO, he manages cmcl innovations, a technology-intensive SME serving the automotive, chemical/materials and energy industries.

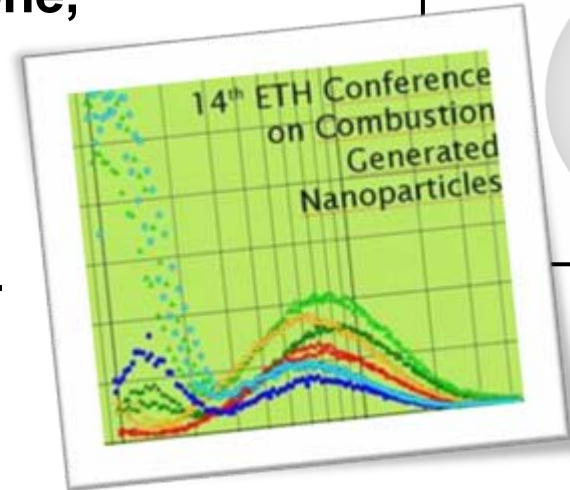
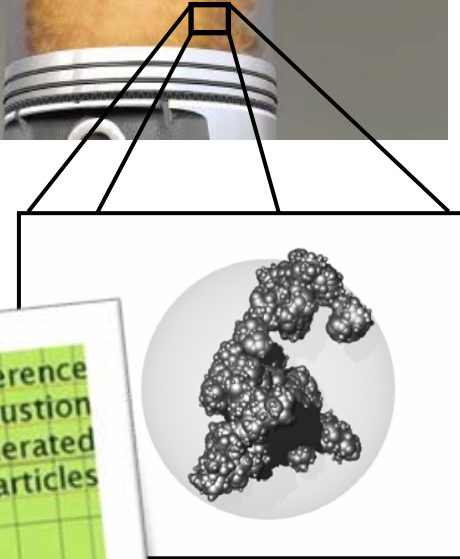
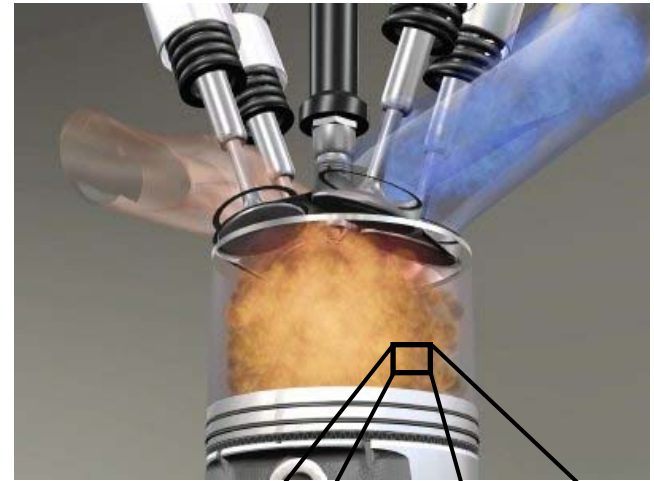
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Modelling soot formation in direct injection SI engines

Markus Kraft, Amit Bhave, Jonathan Etheridge, Sebastian Mosbach, Andreas Braumann, George Brownbridge, Andrew Smallbone, University of Cambridge

Antonis Dris, Robert McDavid
Applied Research Europe, Caterpillar Inc.



02 Aug. 2010



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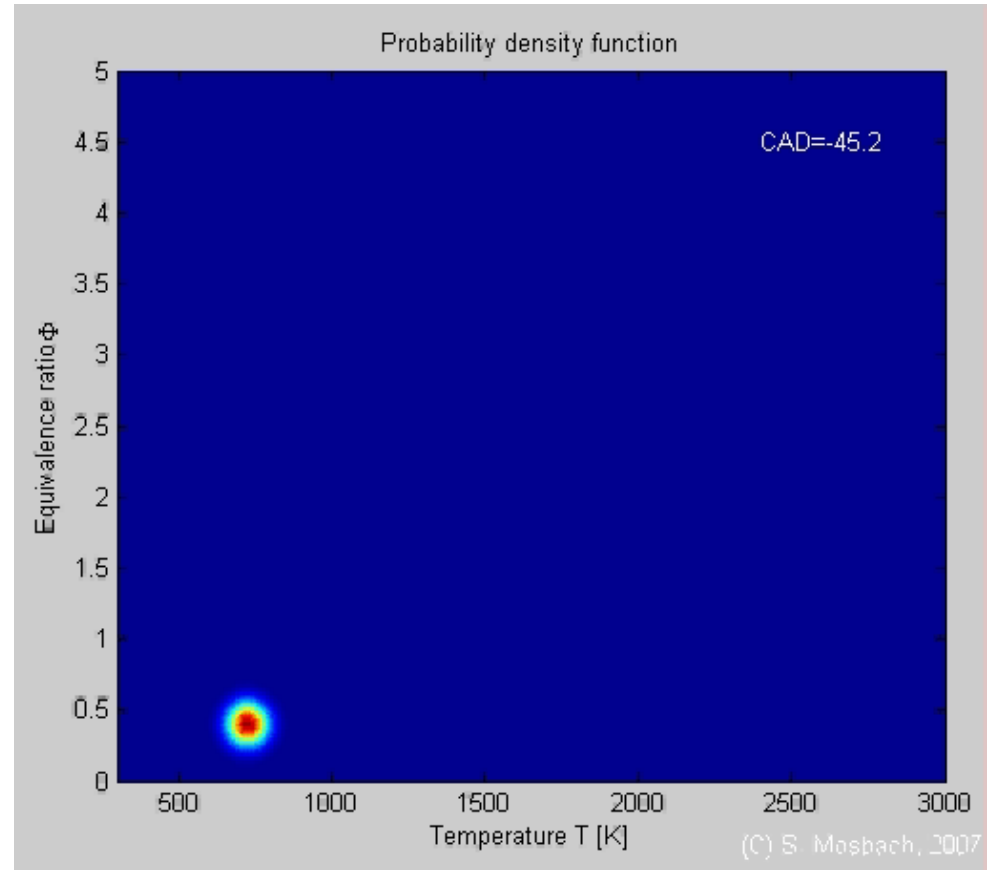
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Stochastic Reactor Model (SRM)

- Closed-volume in-cylinder processes.
- Turbulent mixing, heat transfer, direct injection, piston movement, spark ignition, soot formation.
- Detailed chemical model
- 208 species, 1002 reactions



Test case:
PFI and DI at 40 CAD BTDC



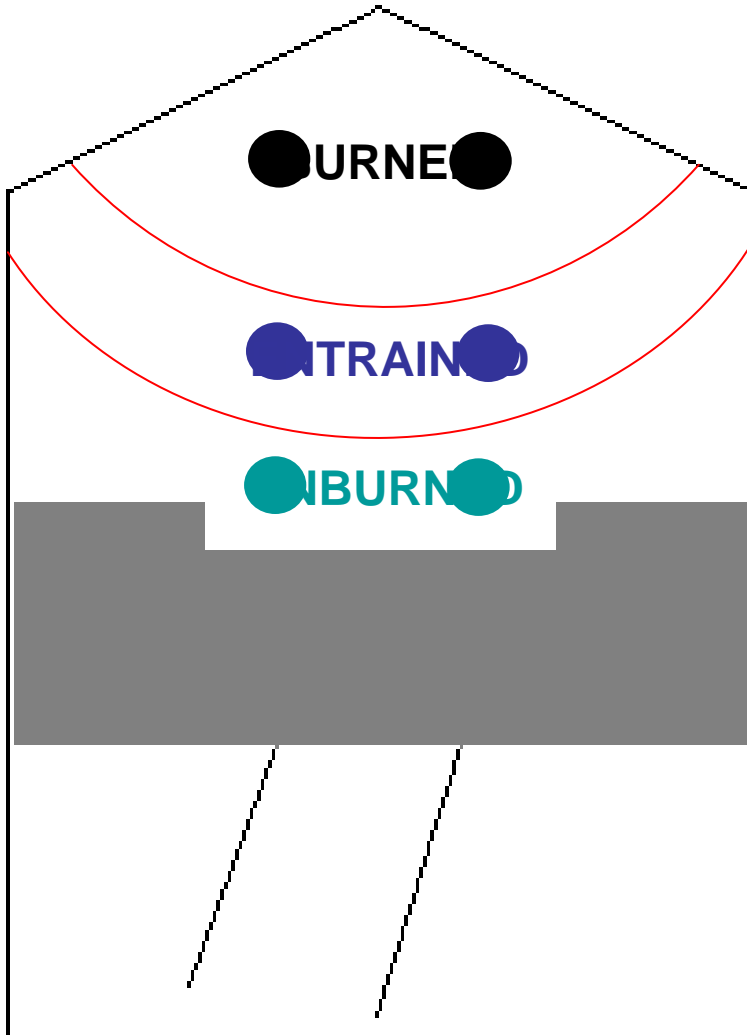
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SRM-SI



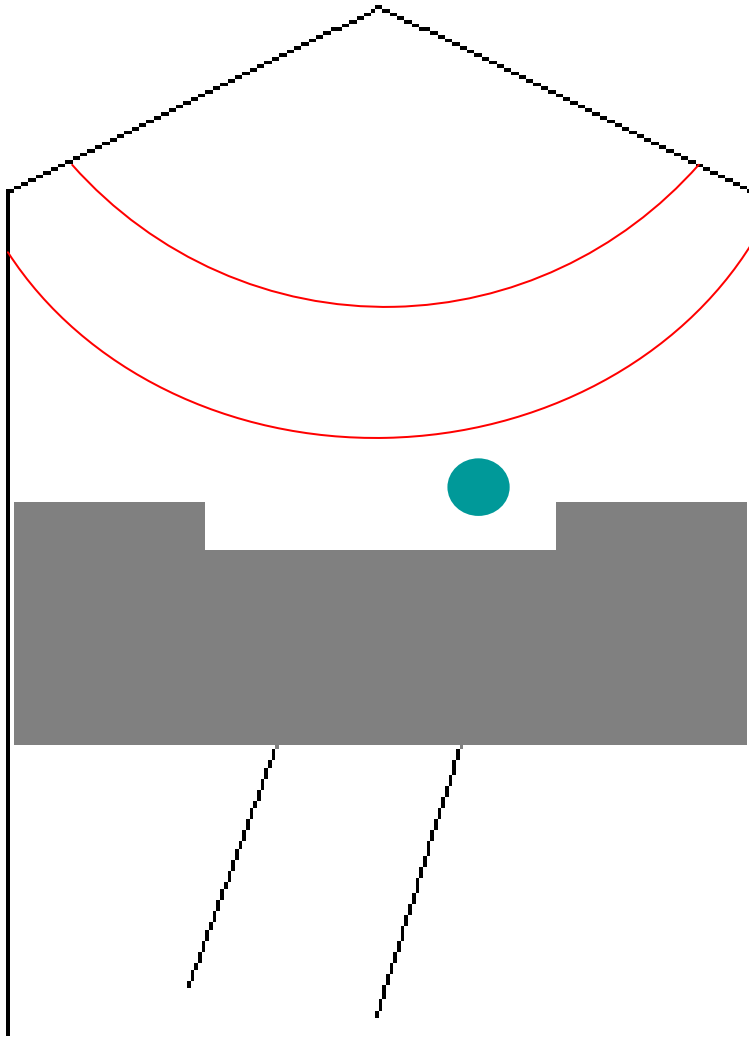
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SI model



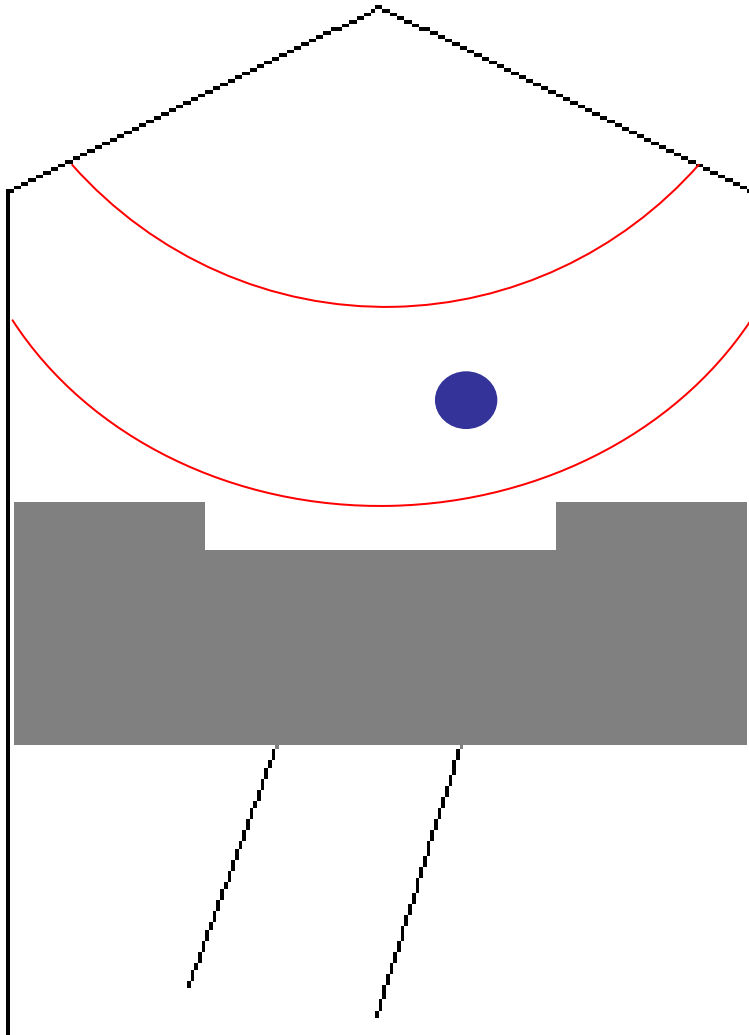
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SI model



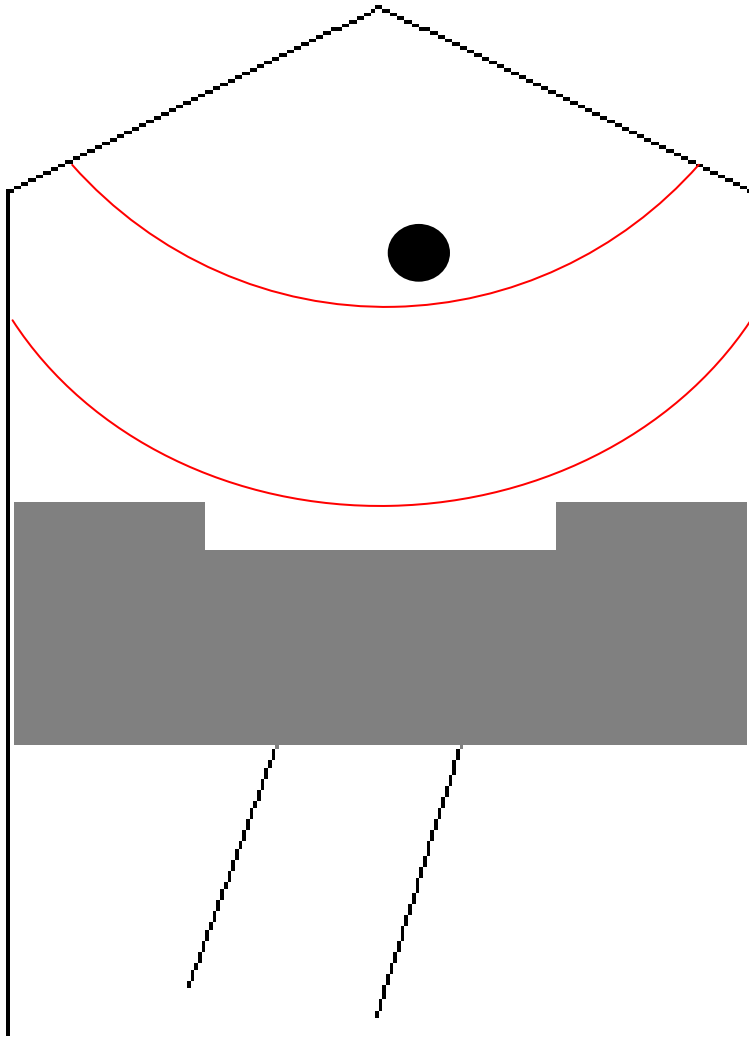
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SI model



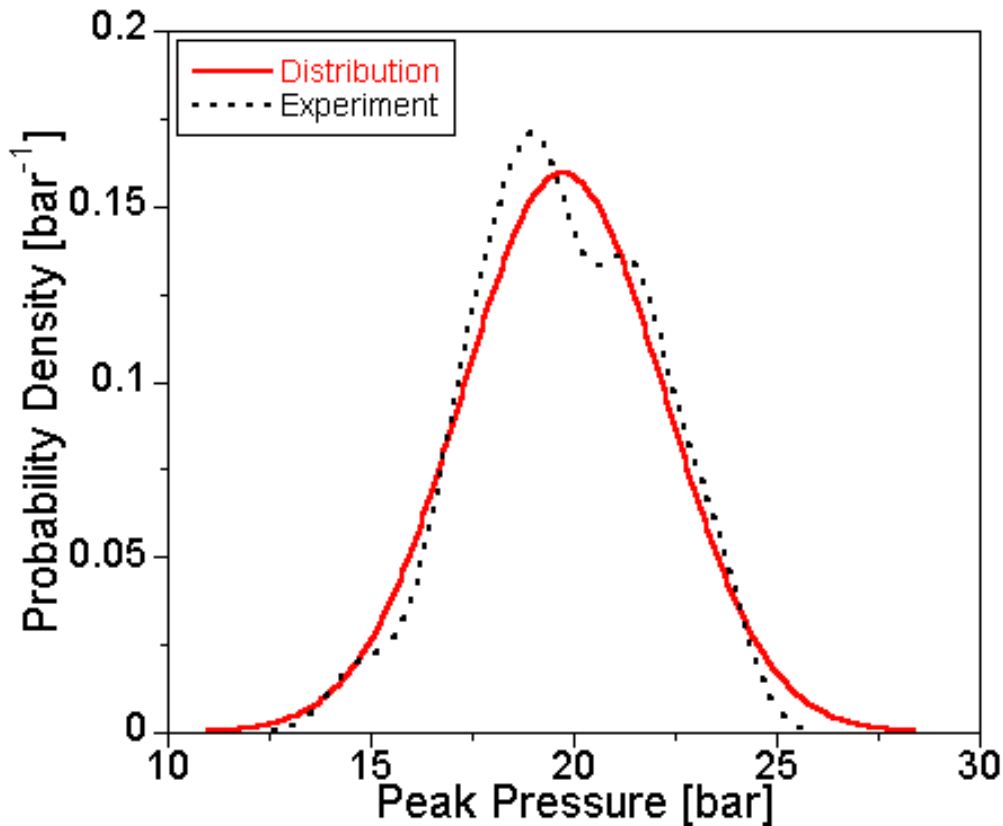
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SI engine CCV

Cycle-to-cycle variation (CCV)



Fuel	gasoline
Bore	87.5 mm
Stroke	83.0 mm
Con. rod length	146.3 mm
Disp. volume	499 cm ³
CR	12.0
Speed	1500 RPM
Air/fuel equiv. ratio	1.0
EGR	28.8%



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ENGINEERING

Hao Wu and Nick Collings
Hopkinson Laboratory

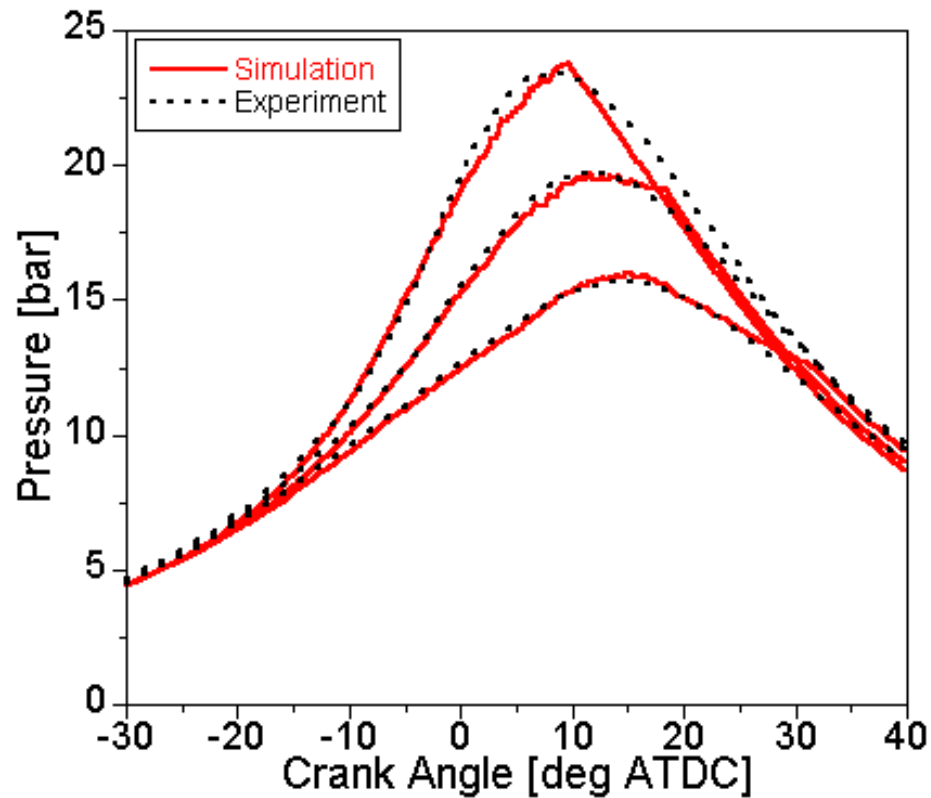


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SI model calibration



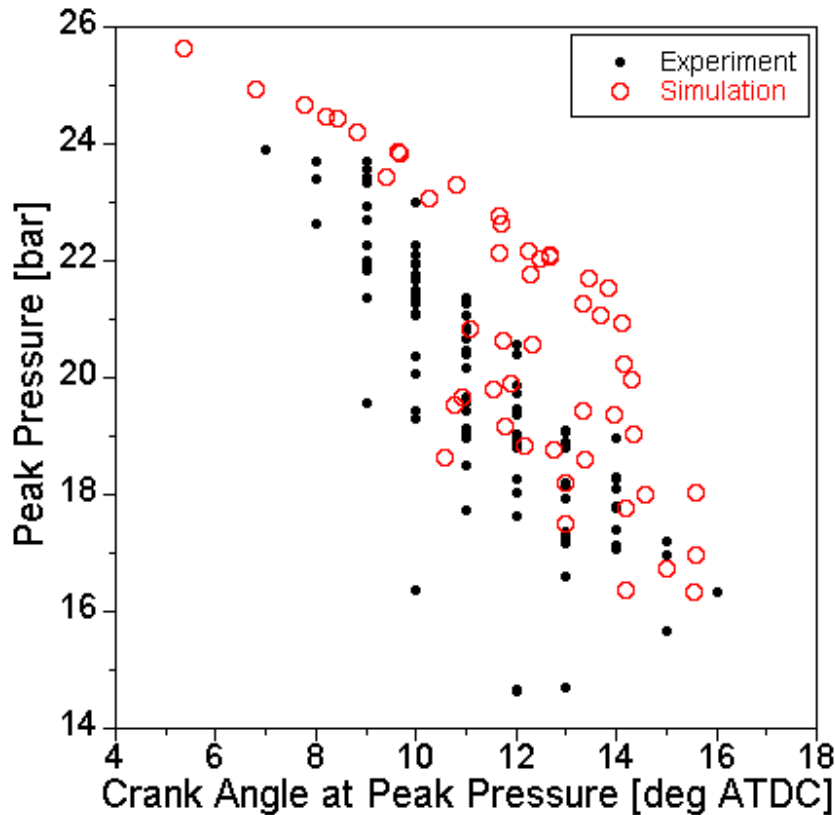
- Characteristic flame speed obtained from:

$$u_T = 0.08C\bar{u}_i \left(\frac{\rho_u}{\rho_i} \right)^{1/2}$$

- Constant, C, calibrated to match representative slow, medium and fast cycles.



Multi-cycle SI simulation



- Model coupled to GT-Power for multi-cycle simulation.
- 50 simulated and 96 experimental cycles.
- NO_x emissions:
 - 790 ppm simulation
 - 530 ppm experiment





DISI engine



- Late injection produces stratified mixture.
- Fuel rich regions close to spark gap.

Image from www.engineforall.com



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DISI engine experiments

- Data from Maricq *et al.*, SAE 1999-01-1530.
- Fuel comprised of 60% paraffin and 40% aromatic compounds.
- Fuel modelled as 60% iso-octane and 40% toluene.
- Exhaust measurements for various injection timings.

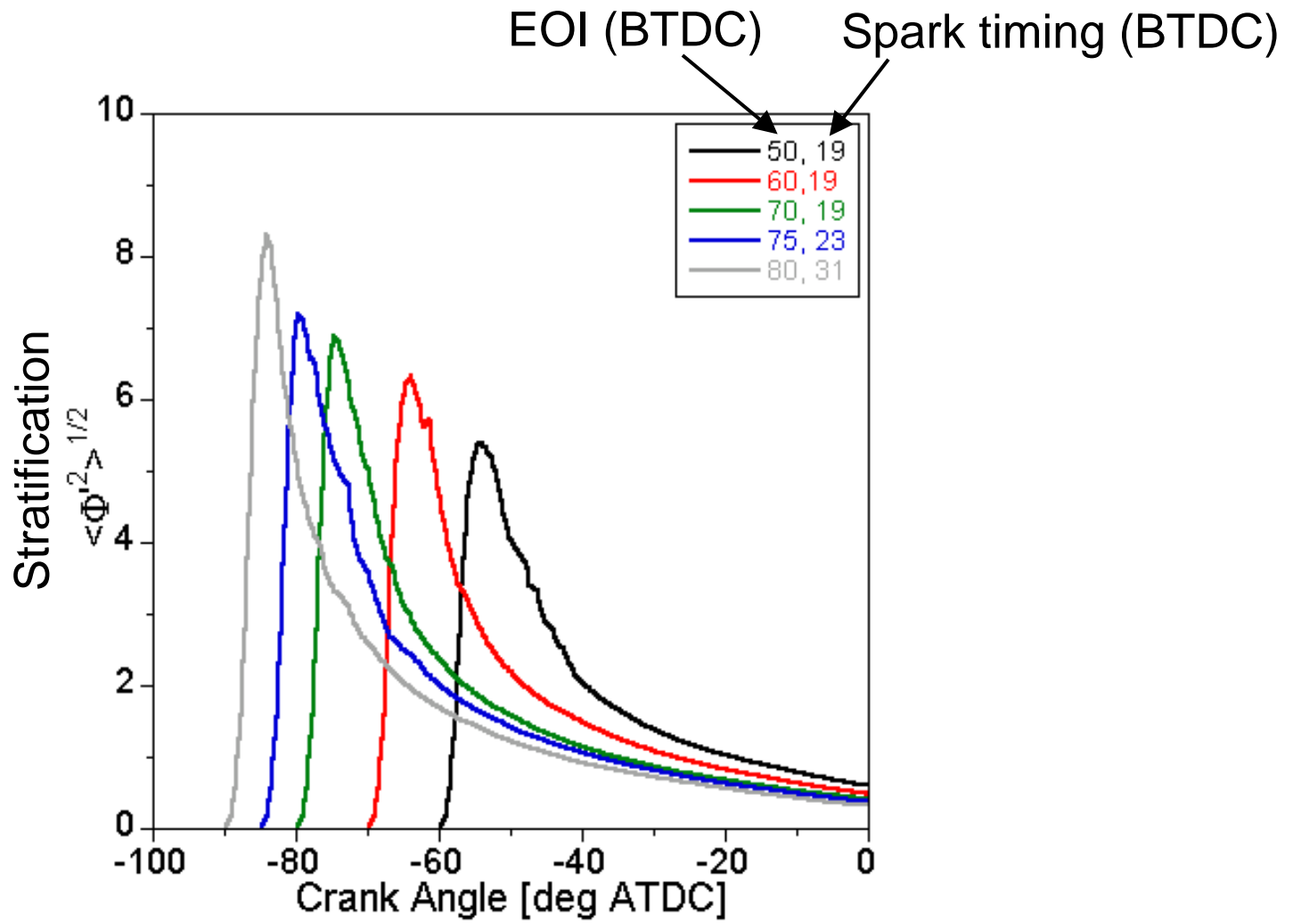
Cylinders	4
Bore [mm]	81.0
Stroke [mm]	89.0
Disp. volume [cm ³ /cyl]	457.5
Compression Ratio	12

1500 RPM, $\Phi=0.58$ (global)

Case	1	2	3	4	5
EOI [CAD ATDC]	-50	-60	-70	-75	-80
Spark Timing [CAD ATDC]	-19	-19	-19	-23	-31

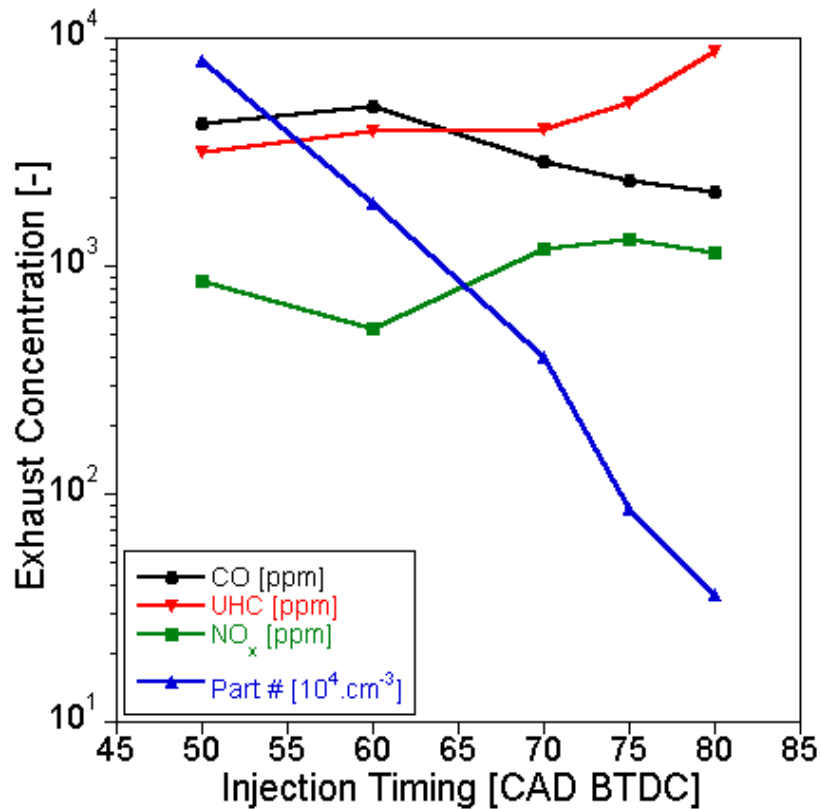


DISI engine simulation results

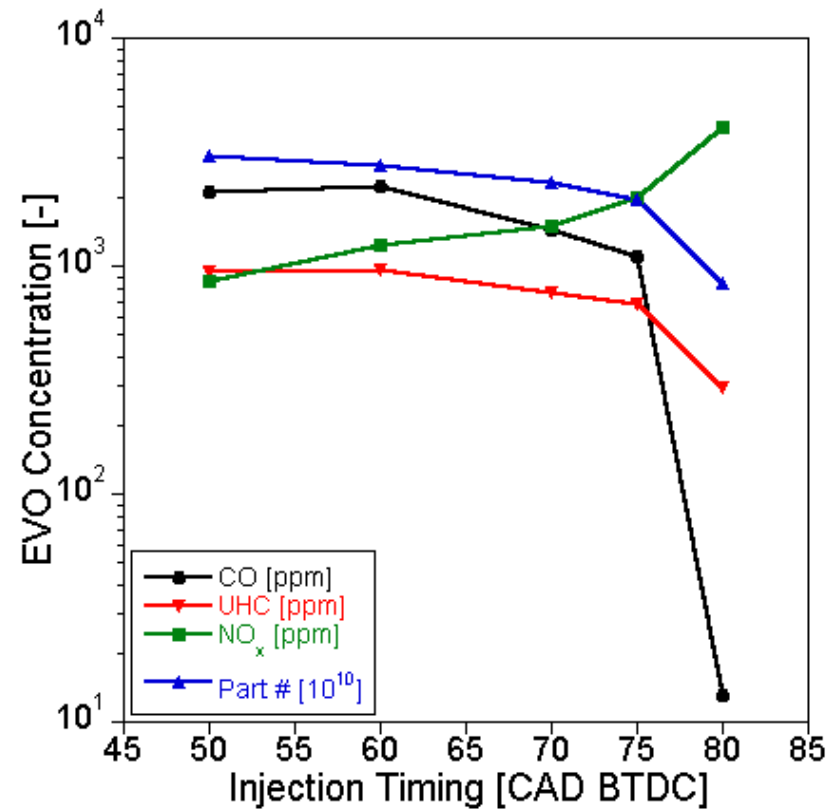


DISI engine emissions

Experiment

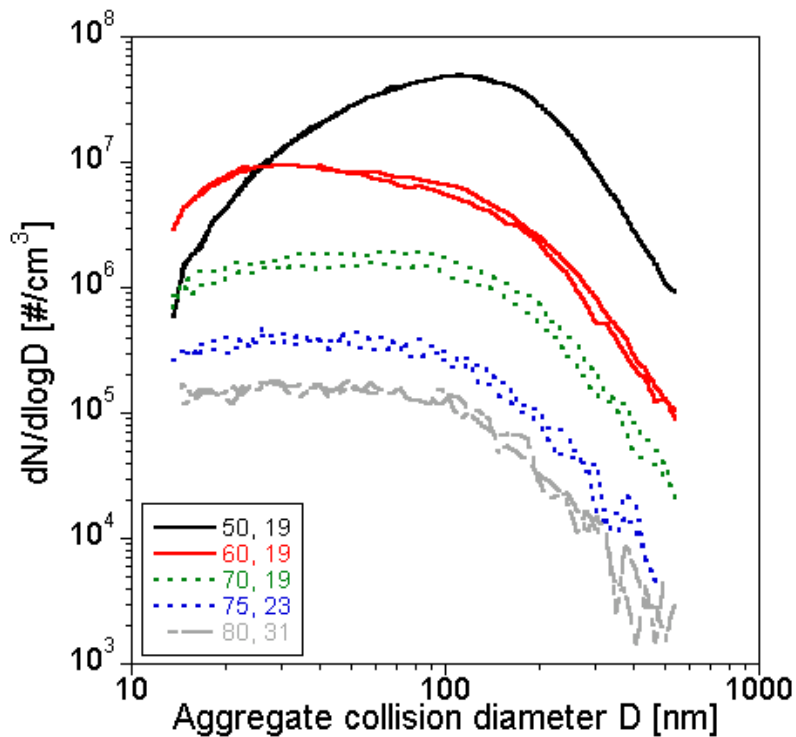


Simulation

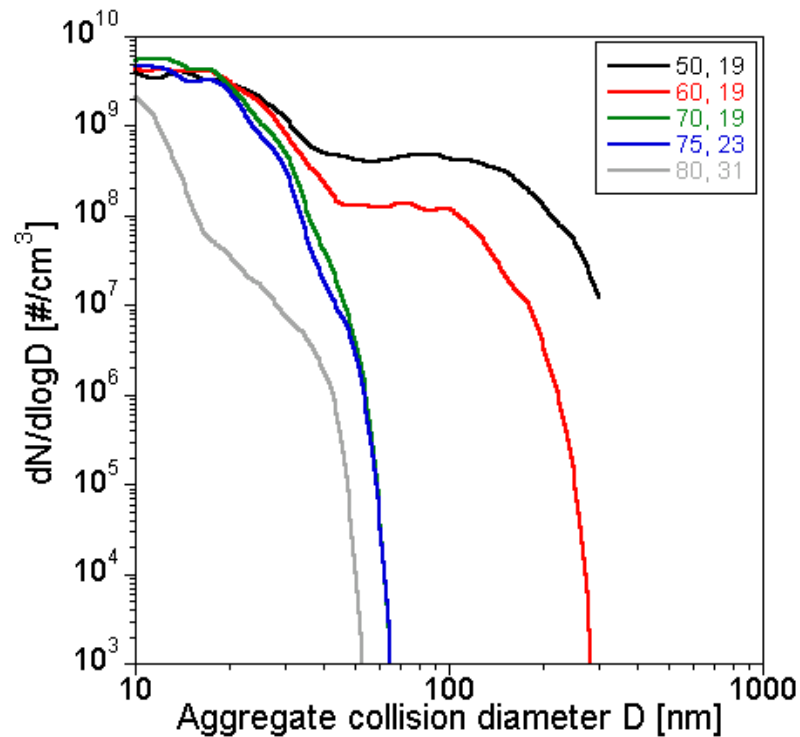


Particle size distributions

Experiment



Simulation



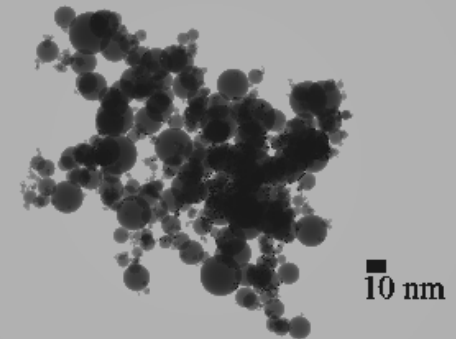
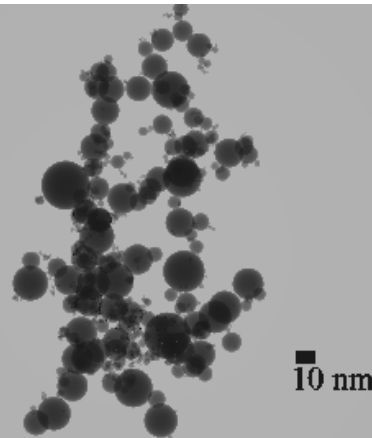
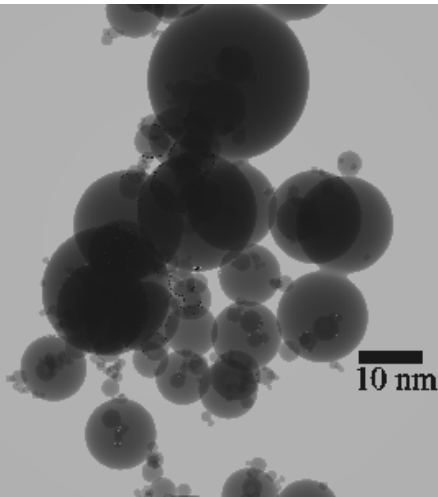


Soot in DISI engine

2.6 CAD ATDC

12.6 CAD ATDC

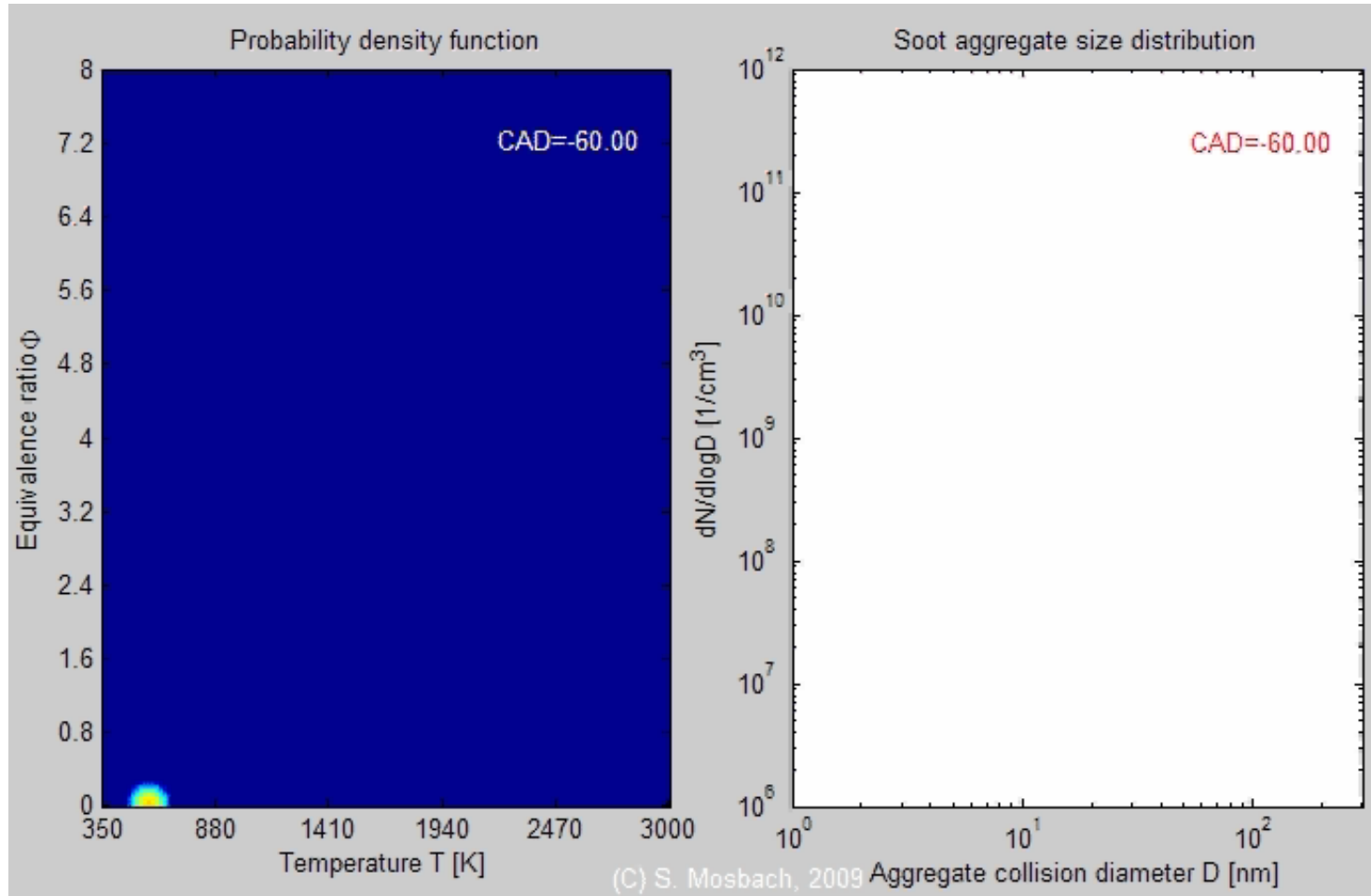
32.6 CAD ATDC



CAD [deg ATDC]	2.6	12.6	32.6
No. Primaries	492	1315	2083
Coll. Diam [nm]	70	108	137



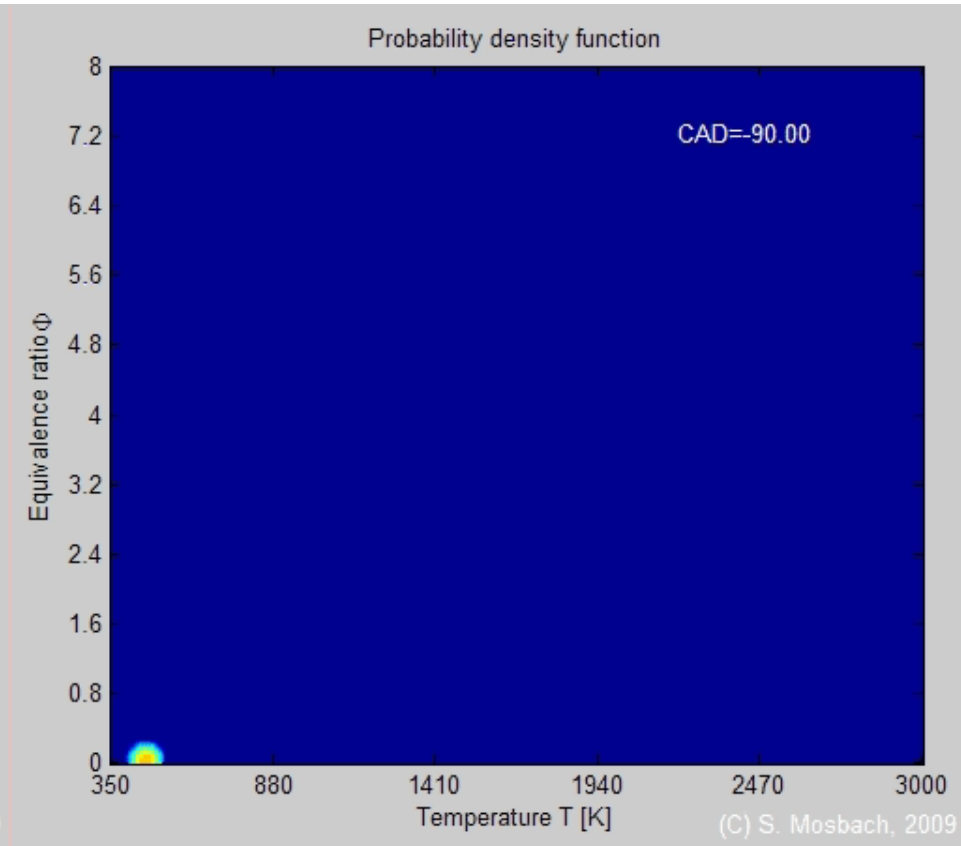
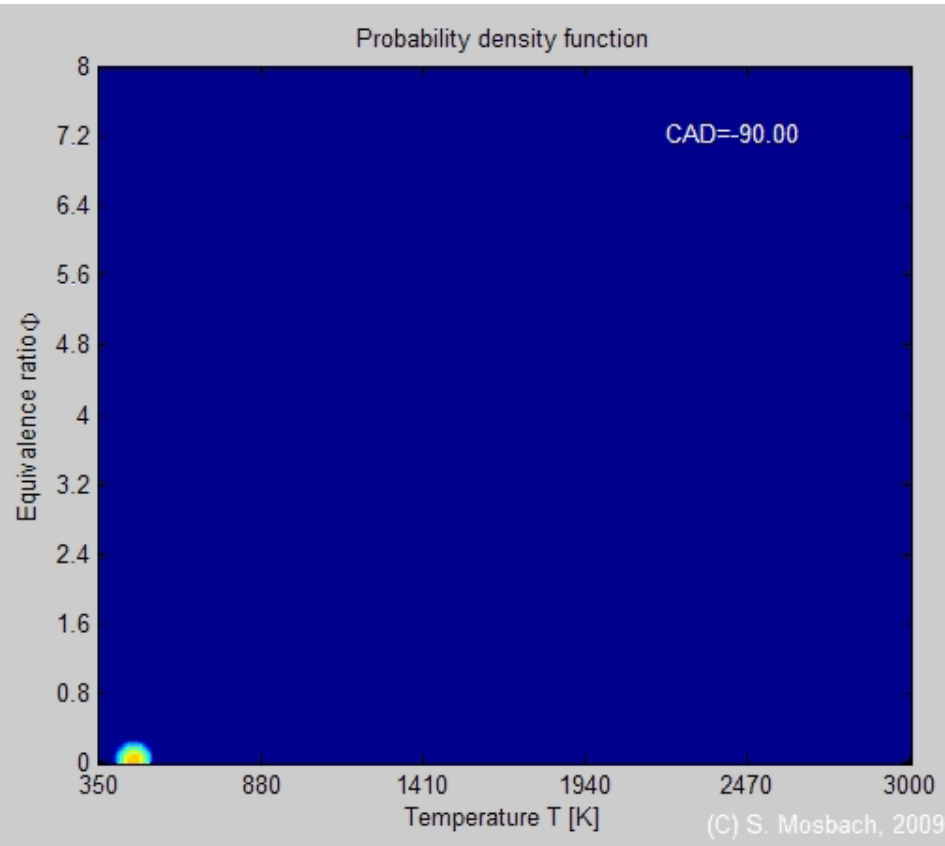
Temporal evolution (late injection)



Comparison early/late injection

EOI -80 CAD ATDC
Spark -31 CAD ATDC

EOI -50 CAD ATDC
Spark -19 CAD ATDC



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Problem

Current engine model development

- Experimental data in a variety of formats, sometimes largely unstructured, often incomplete
- Uncertainties/errors associated with experimental data typically unknown or unavailable
- Too many models and “tuneable”/unknown model parameters
- How “good” (or not) is a particular model?

=> Ad hoc, fragmented, short-term approach





Solution: Process Informatics

We need a robust **integrated methodology** to help us work systematically and efficiently:

- Effective use of cost-intensive experimental data through **data standardisation**
- Systematic and robust model development through **systematic optimisation**, taking into account **uncertainties**
- Suggesting “useful” future experiments



A data model: engineML

Consistent format

- point data (e.g. rpm, CO, u_i)
- time resolved data (p -CA)
- apparatus (production engine, research engine)
- errors
- data type (consistent units)
- raw or processed
- experimental or model

eXtensible Markup Language (XML)

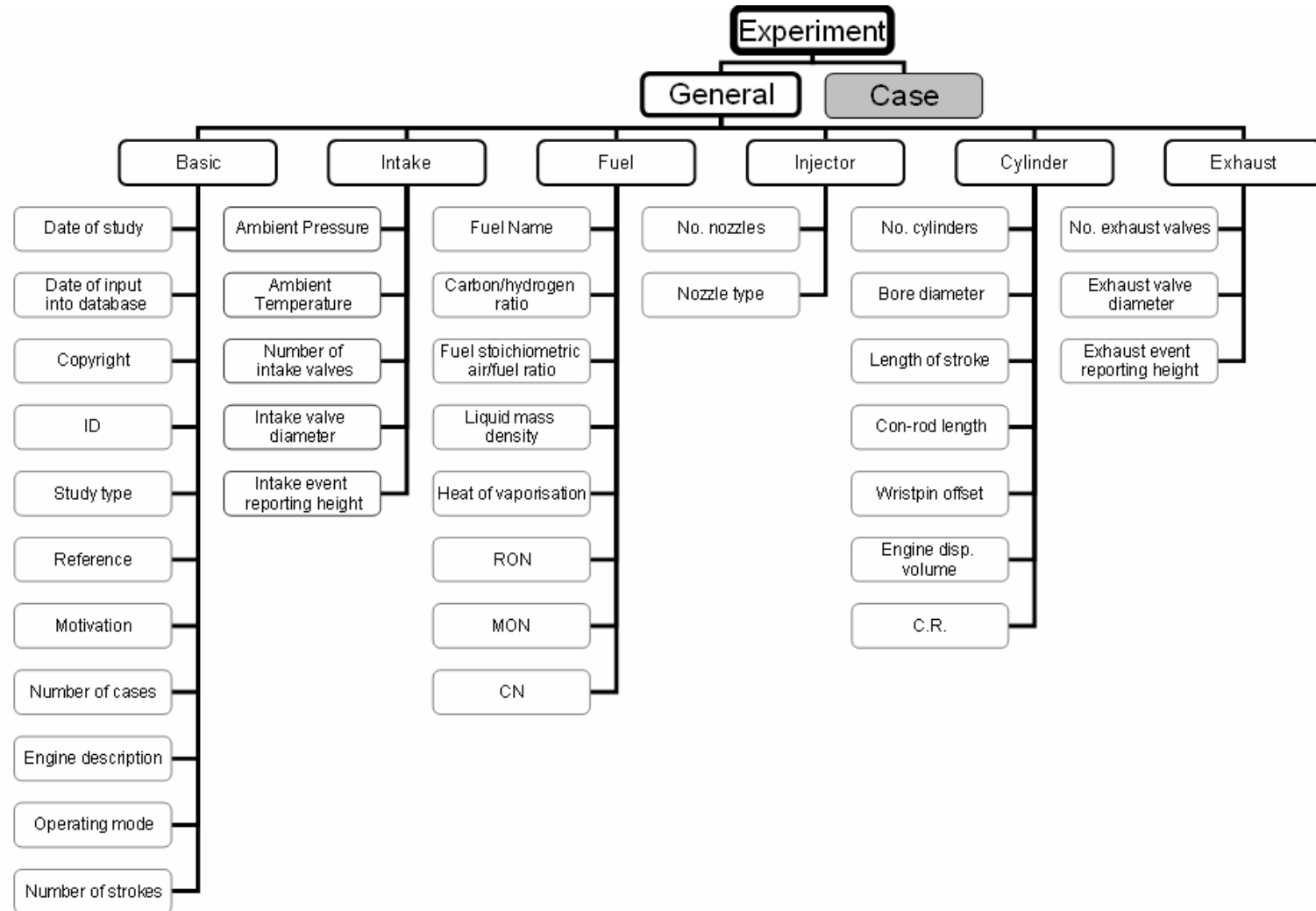
- machine and human readable, tagged with metadata
- highly structured (tree), easily queried
- can be validated against schema

Easily accessible database

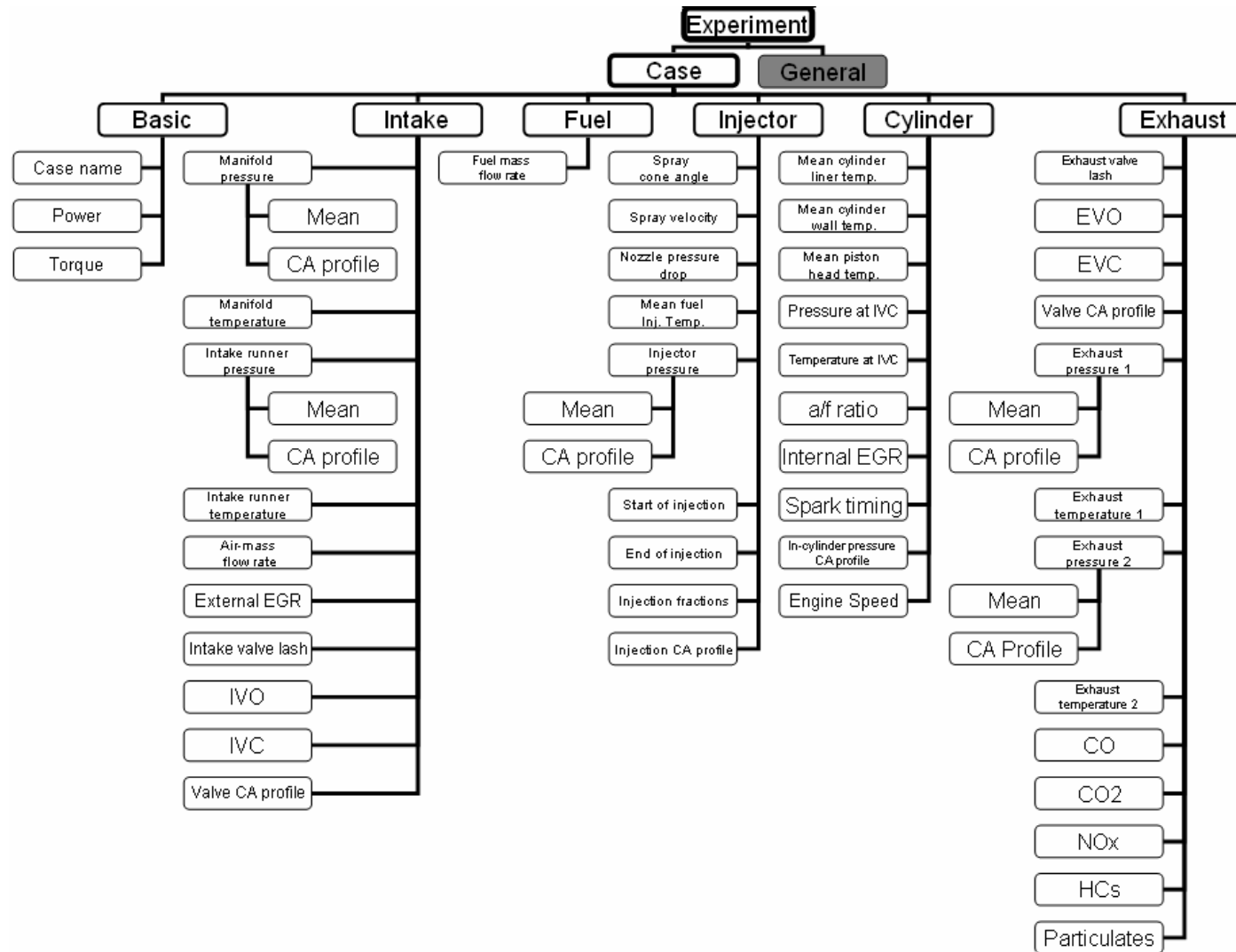
- read by model code
- data stored consistently
- old data never “lost”



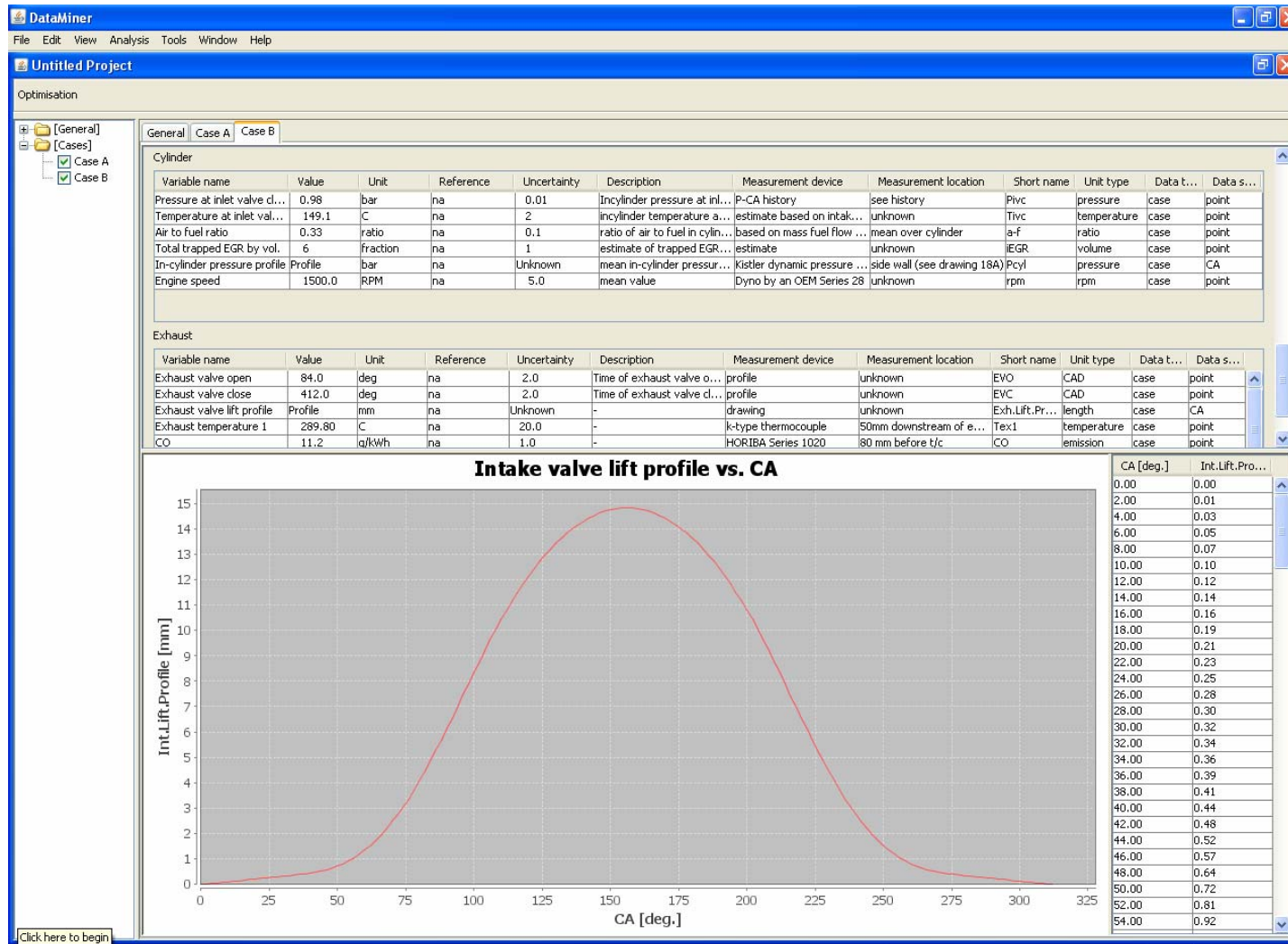
General data



Case data



Data visualisation





Model parameter optimisation

We must accept that model parameters exist and have to be tuned to experimental data!

What is the best way to fit these parameters?

- As many data points as possible
- Uncertainties in experimental data as well as model parameters must be accounted for
- Use experimental data to reduce model parameter uncertainties
- Use experimental and model uncertainties to identify possible outliers in the data, or model shortcomings

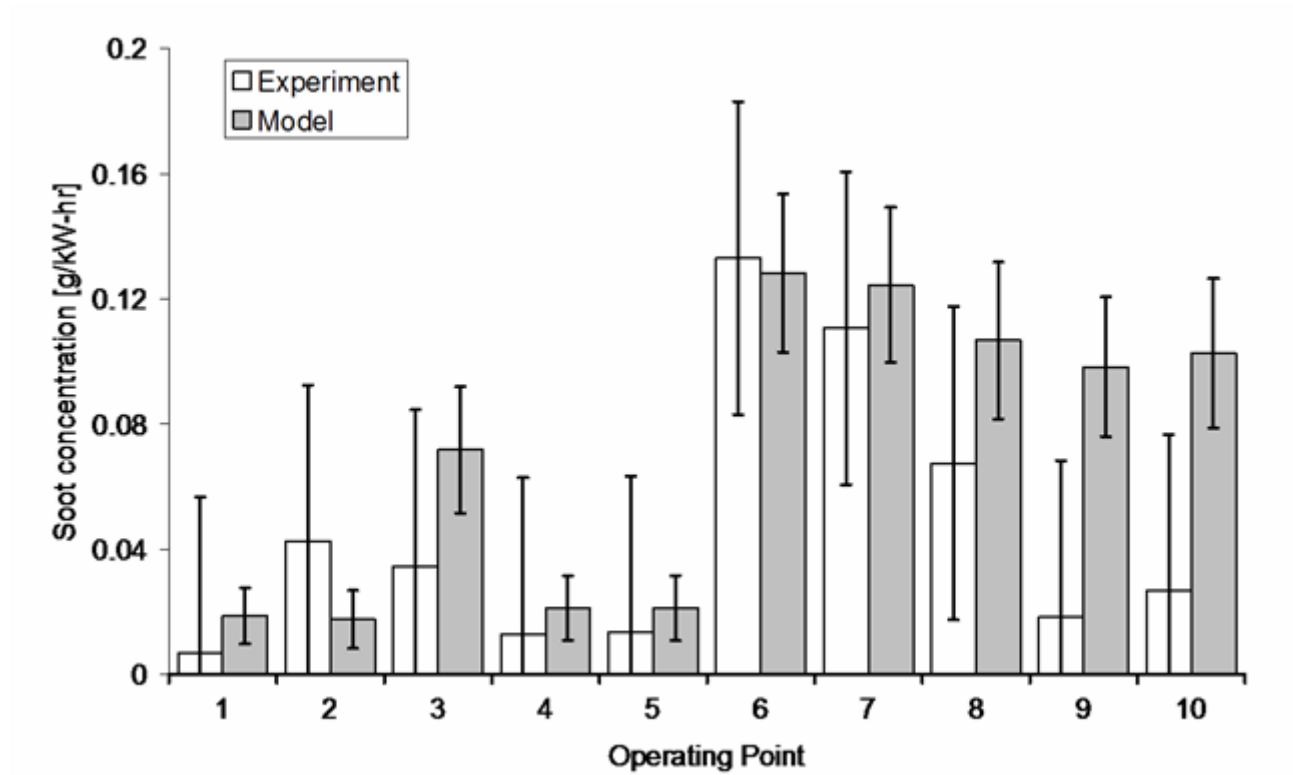


Application: Diesel soot

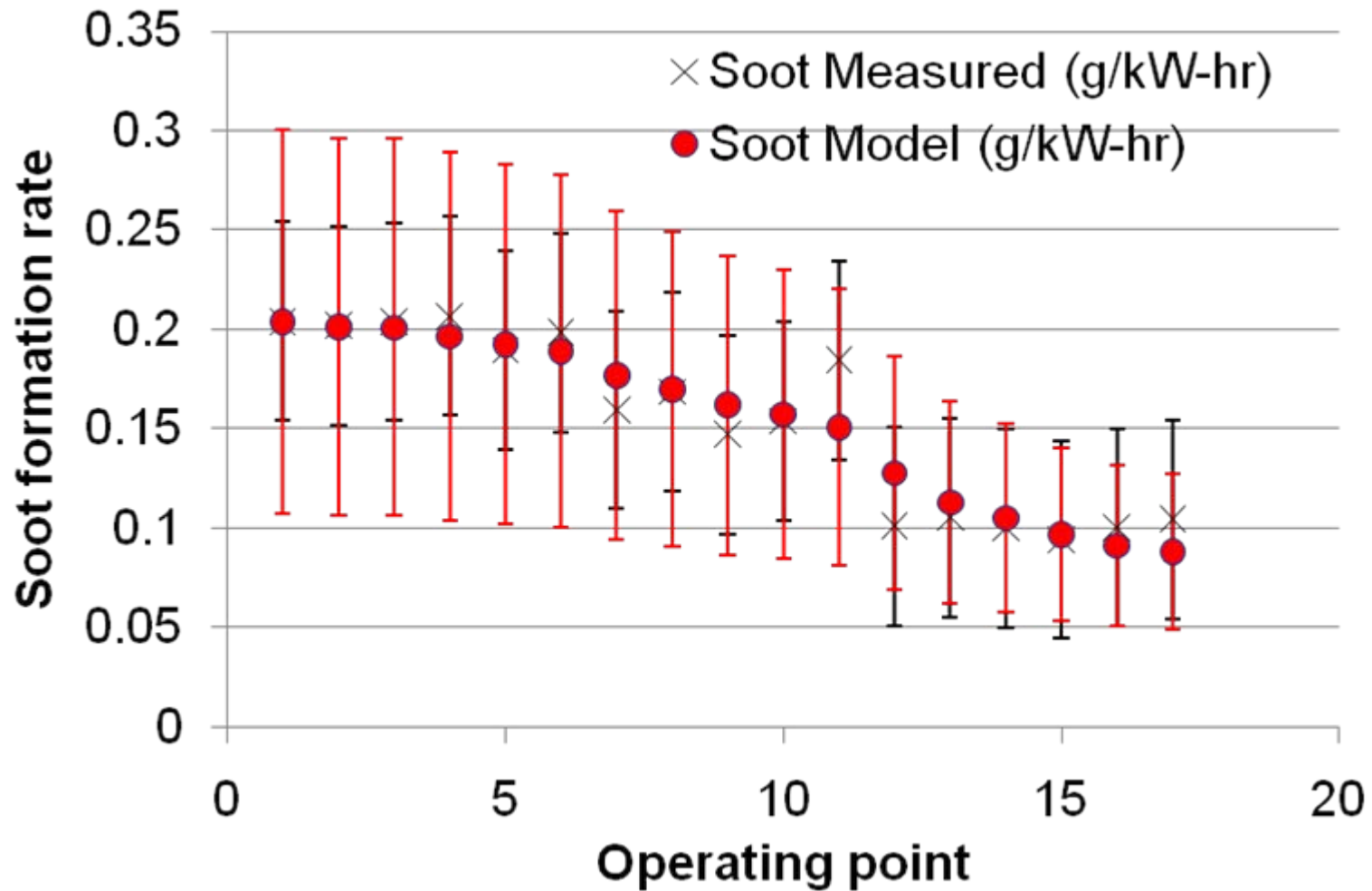
Empirical soot model (Plee):

$$\text{soot}[g] = A \cdot \text{mps}^B \cdot \text{phi}^C \cdot \exp\left(\frac{D}{T_f}\right)$$

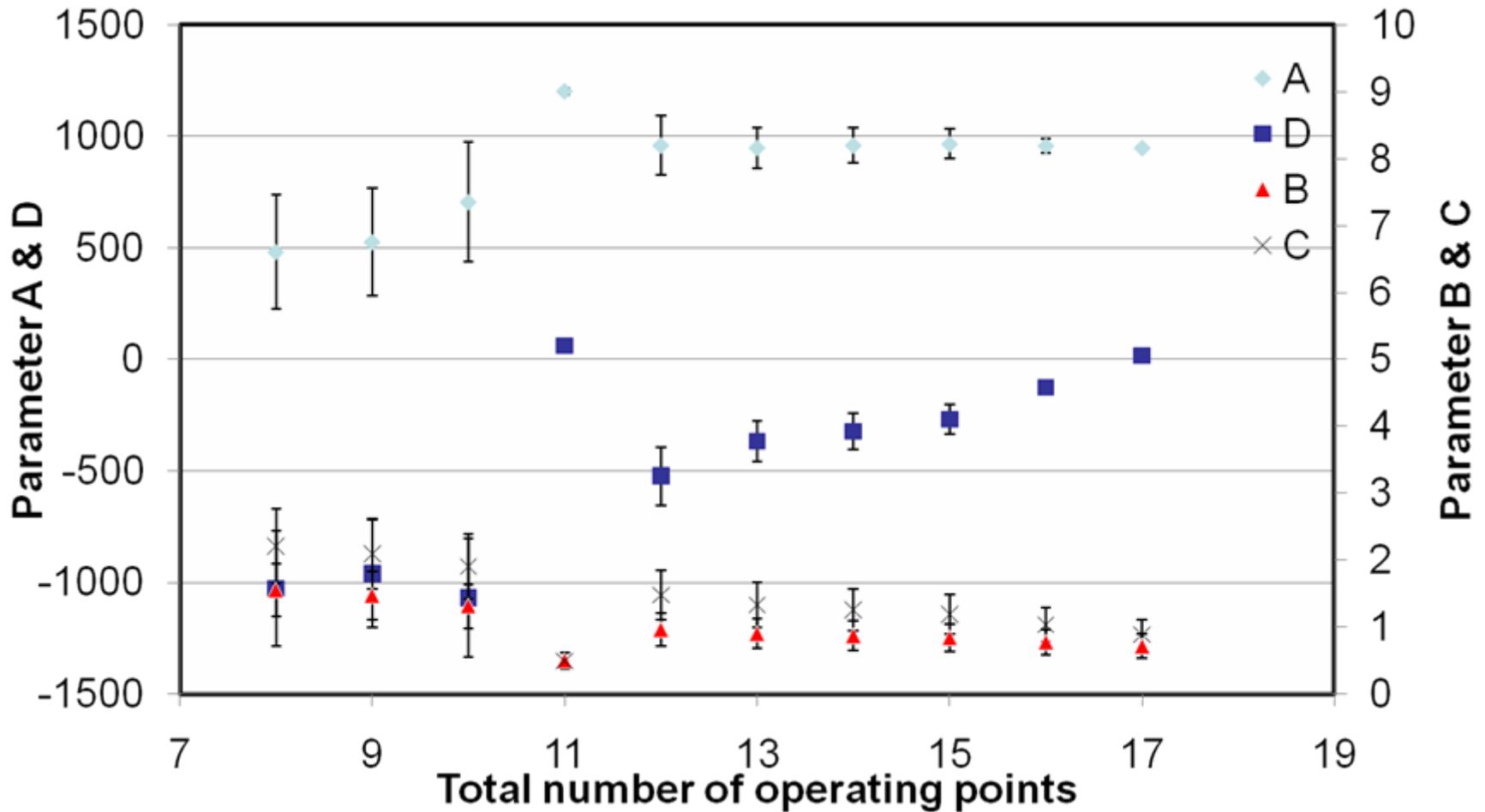
Optimised parameters A, B, C, D,
against database of
503 operating points from 7 engines



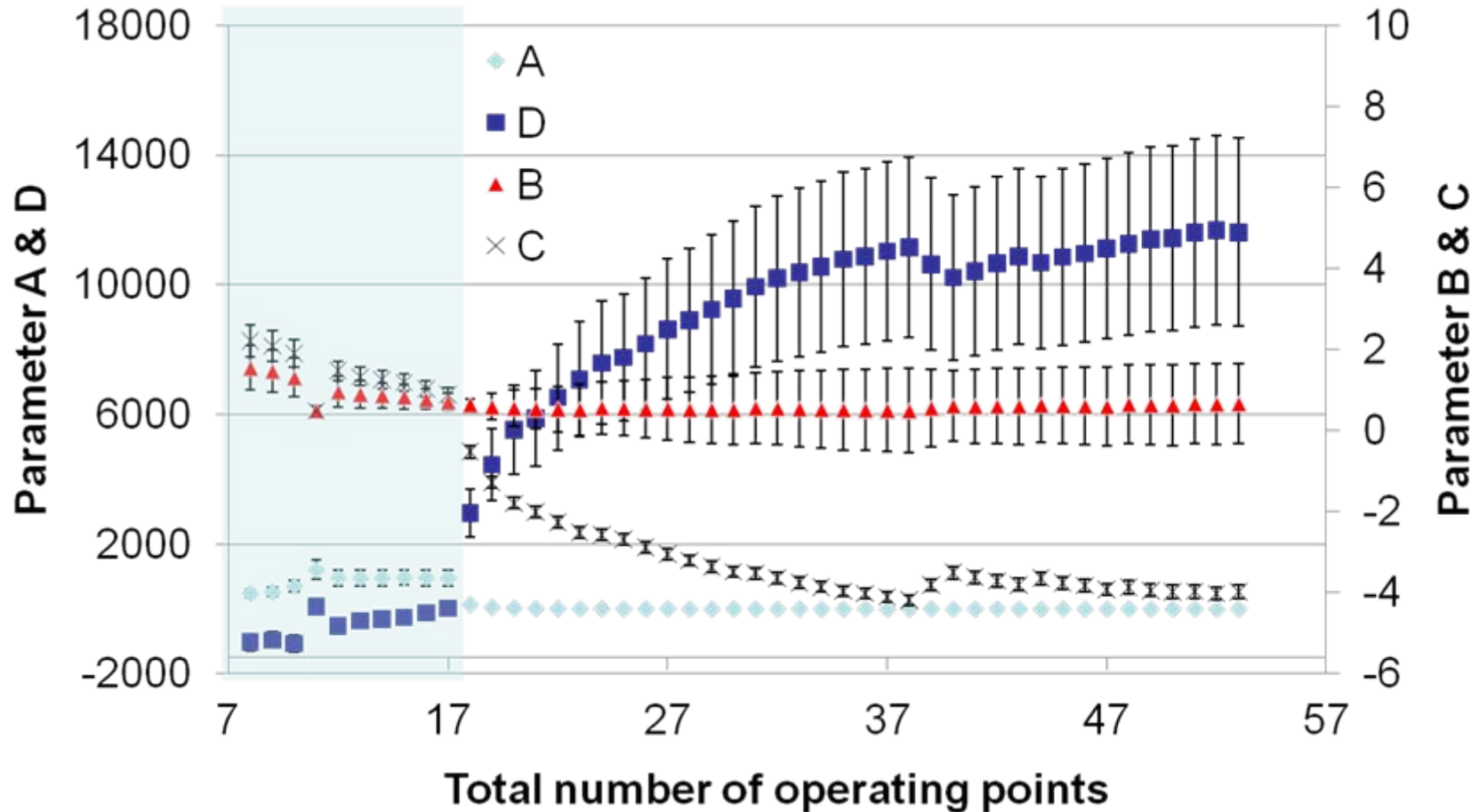
Example engine



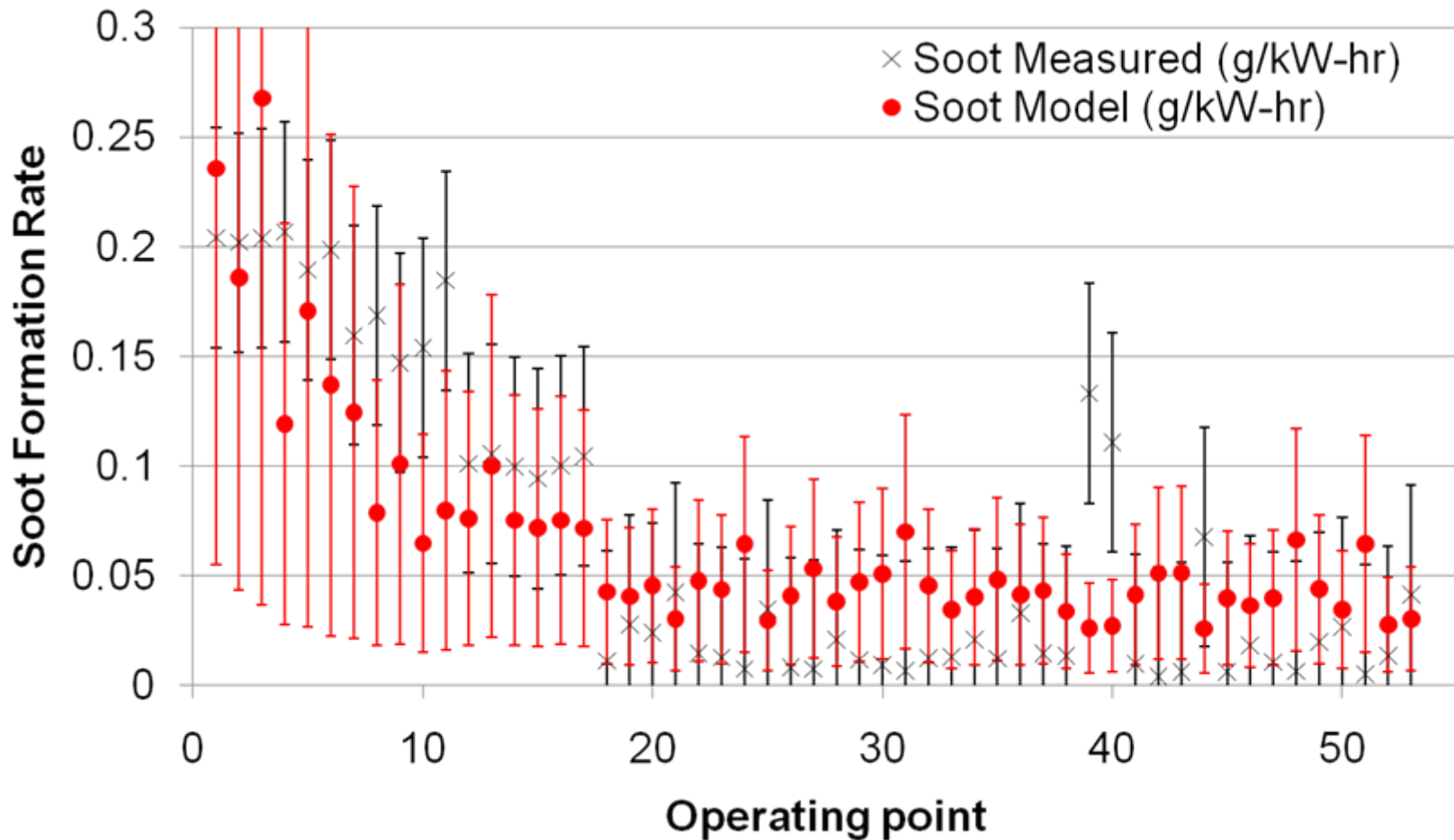
Parameter optimisation



Add data from second engine



Add data from second engine



Summary

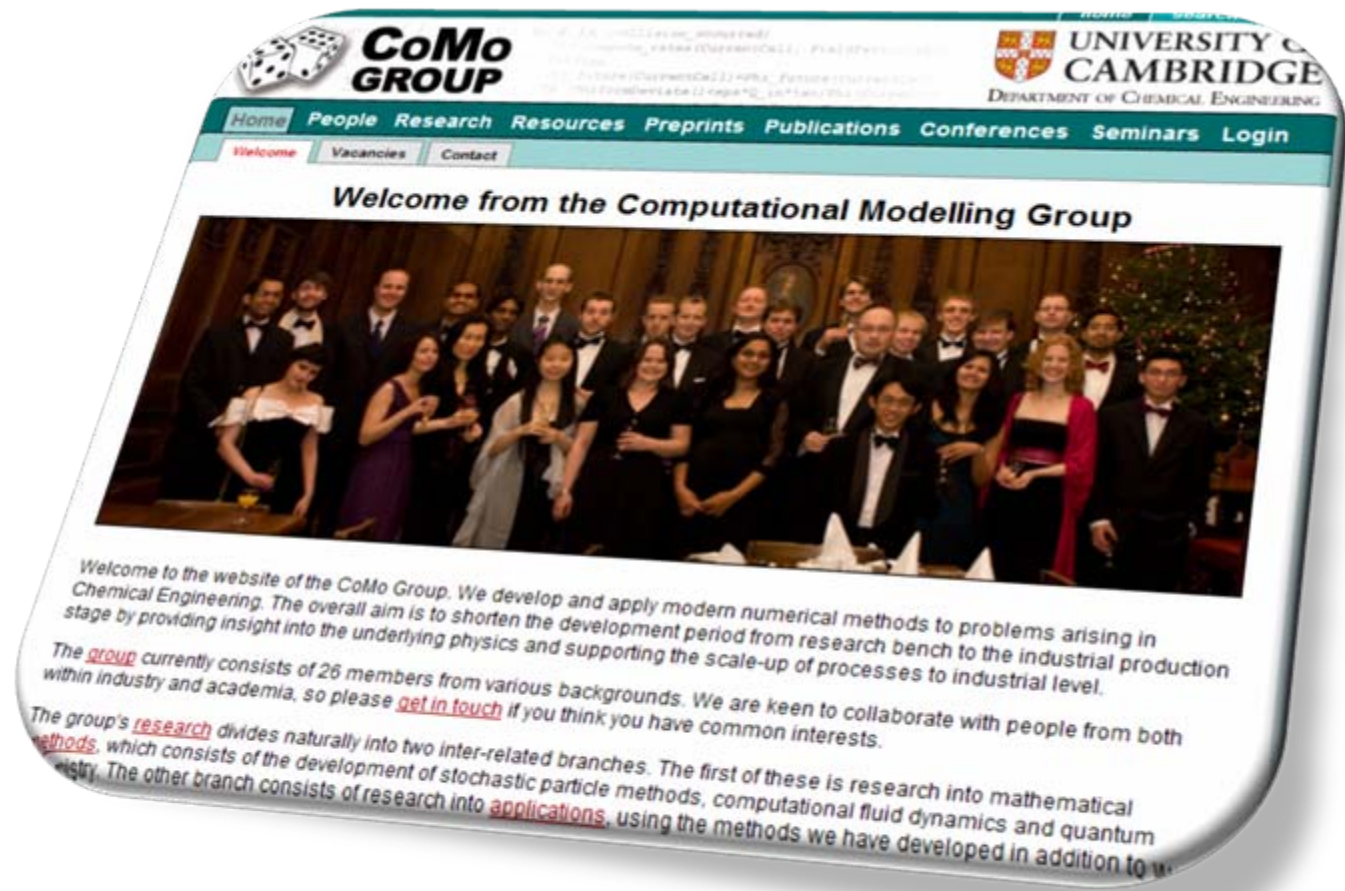
- Results of detailed soot modelling in a DISI engine have been presented.
- A Process Informatics based methodology has been proposed for robust engine model development.
- A standardised, machine-readable format, engineML, has been presented.
- Optimisation results including model parameter and experimental uncertainties have been presented for an empirical diesel soot model.





Thank you!

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Additional Slides



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SI model calibration

- Relation between C and the peak pressure obtained.
- Used with peak pressure distribution to provide C during each cycle of a multi-cycle simulation.

