

UConnRCMPy: Python-based Data Analysis for Rapid Compression Machines

Bryan Weber

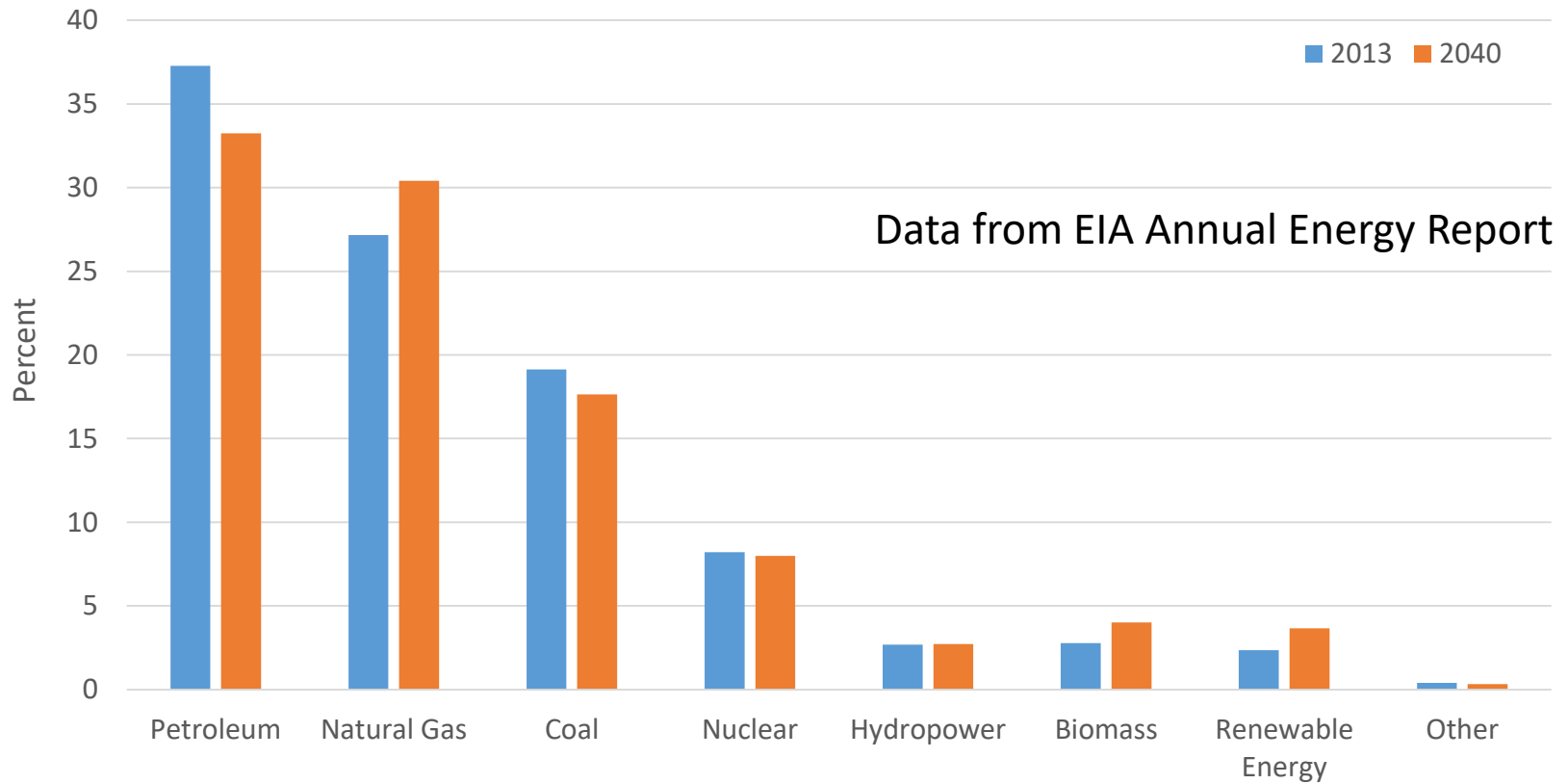
Chih-Jen Sung

University of Connecticut

Combustion drives the energy economy

- Combustion is expected to remain the dominant energy conversion mechanism for many years
- The combustion of fossil fuels has been implicated in a number of harmful effects on human health, the environment, and the economy
- Two solutions have been proposed:
 - Better engines
 - Better fuels

Combustion drives the energy economy

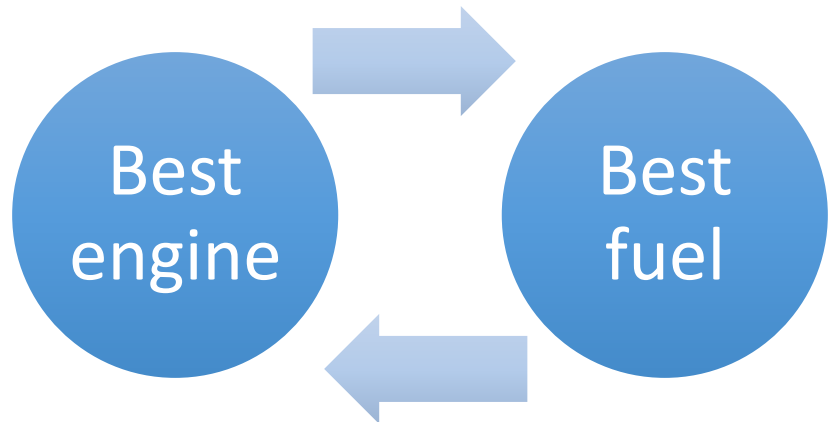


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We need both solutions to make substantial progress

- Selecting the best alternative fuel requires knowledge of the best engine design, which requires deciding which is the best fuel...
- Testing every fuel in every engine is too expensive and too time consuming



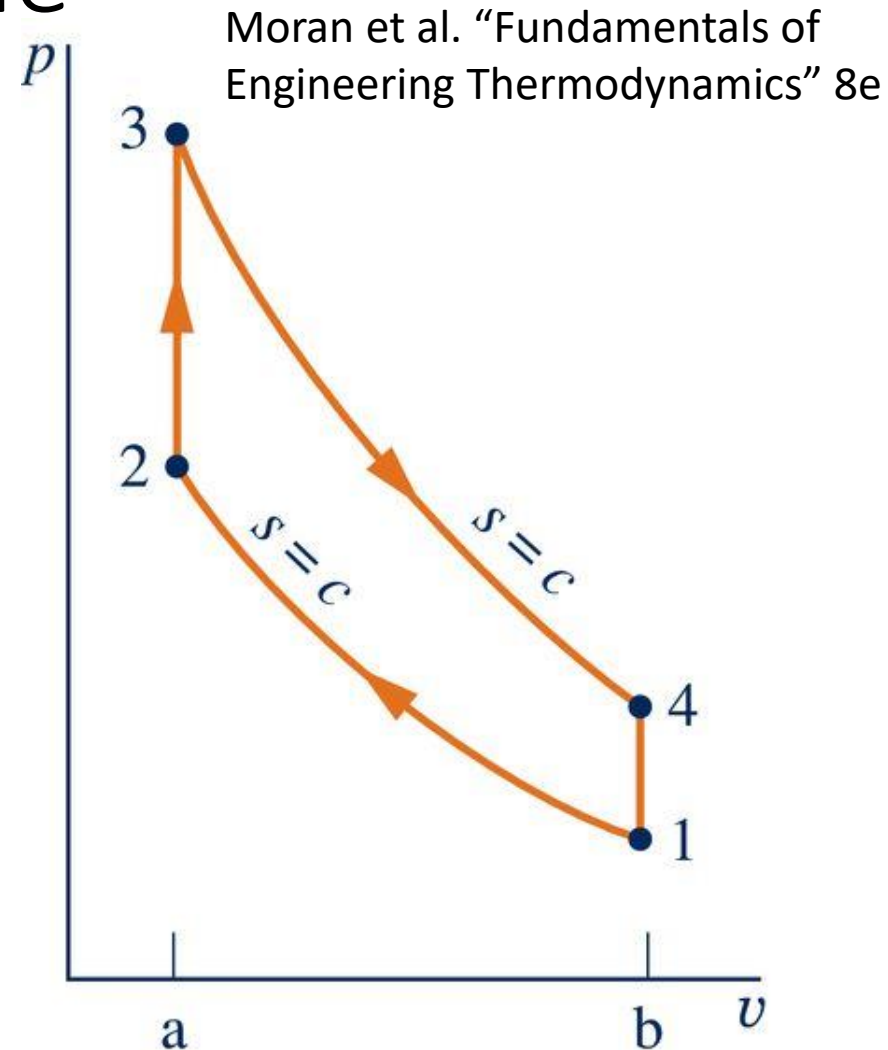
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- Computer-aided design can be employed to design fuel-flexible, high-efficiency, low-emissions engines **if the models we use are predictive**



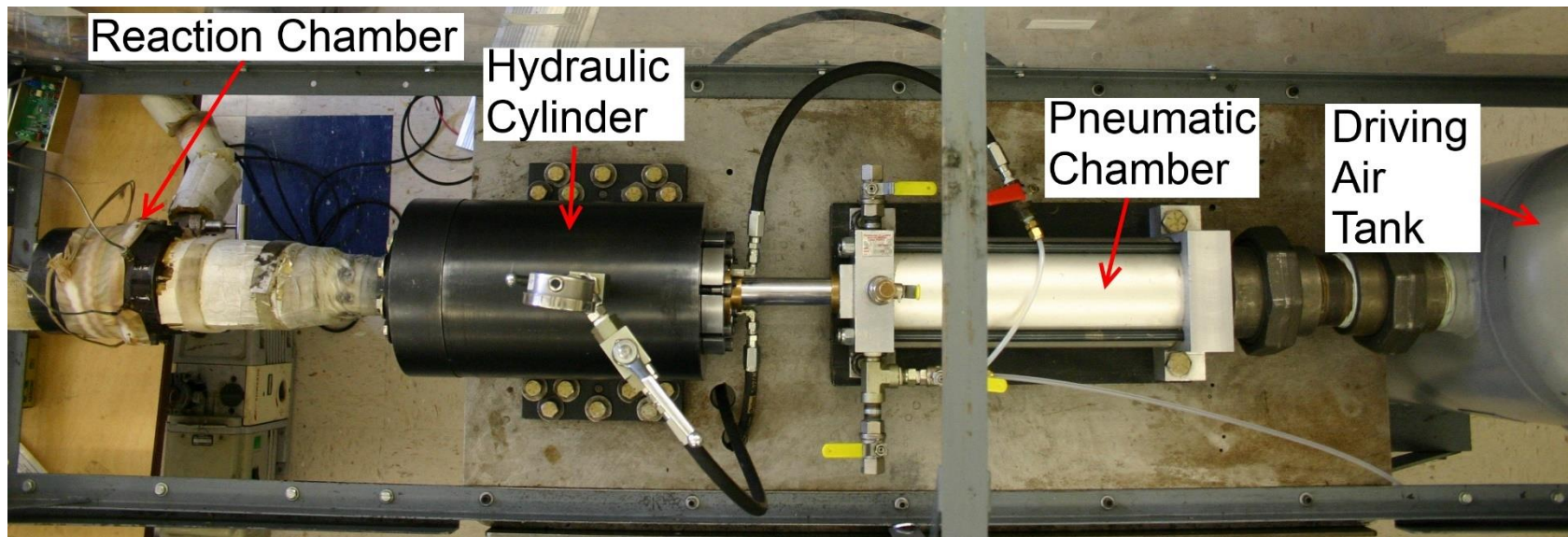
We want to do experiments that model a real engine

- Real engines operate at high pressure and low temperature
- Real engines are affected by:
 - Intake/Exhaust
 - Inhomogeneity
 - Fluid Mechanics
 - Power output

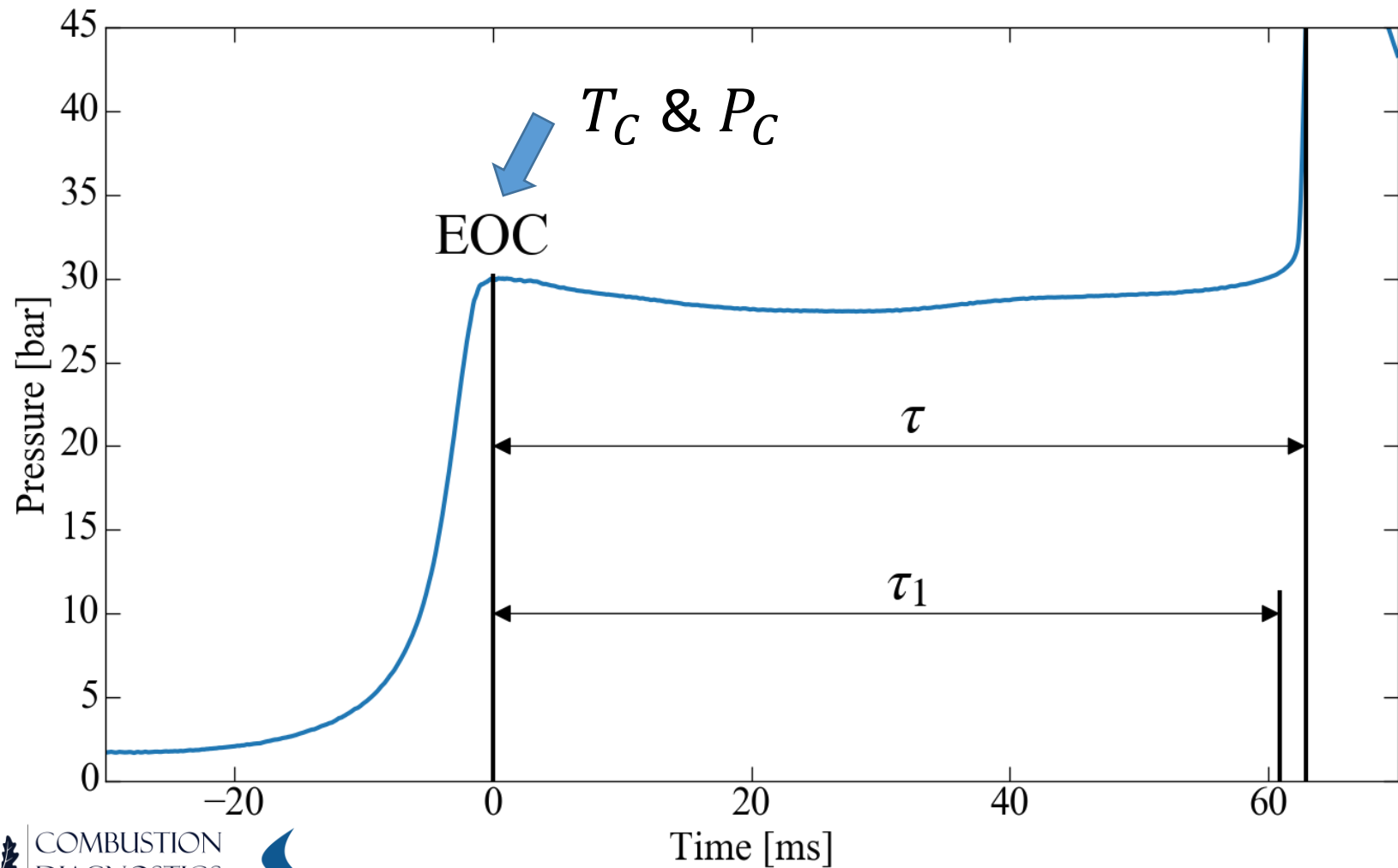


Rapid Compression Machine

- Experiments are conducted in a Rapid Compression Machine (RCM)
- High pressure and low temperature conditions
- Minimal effects of fluid mechanics and inhomogeneity



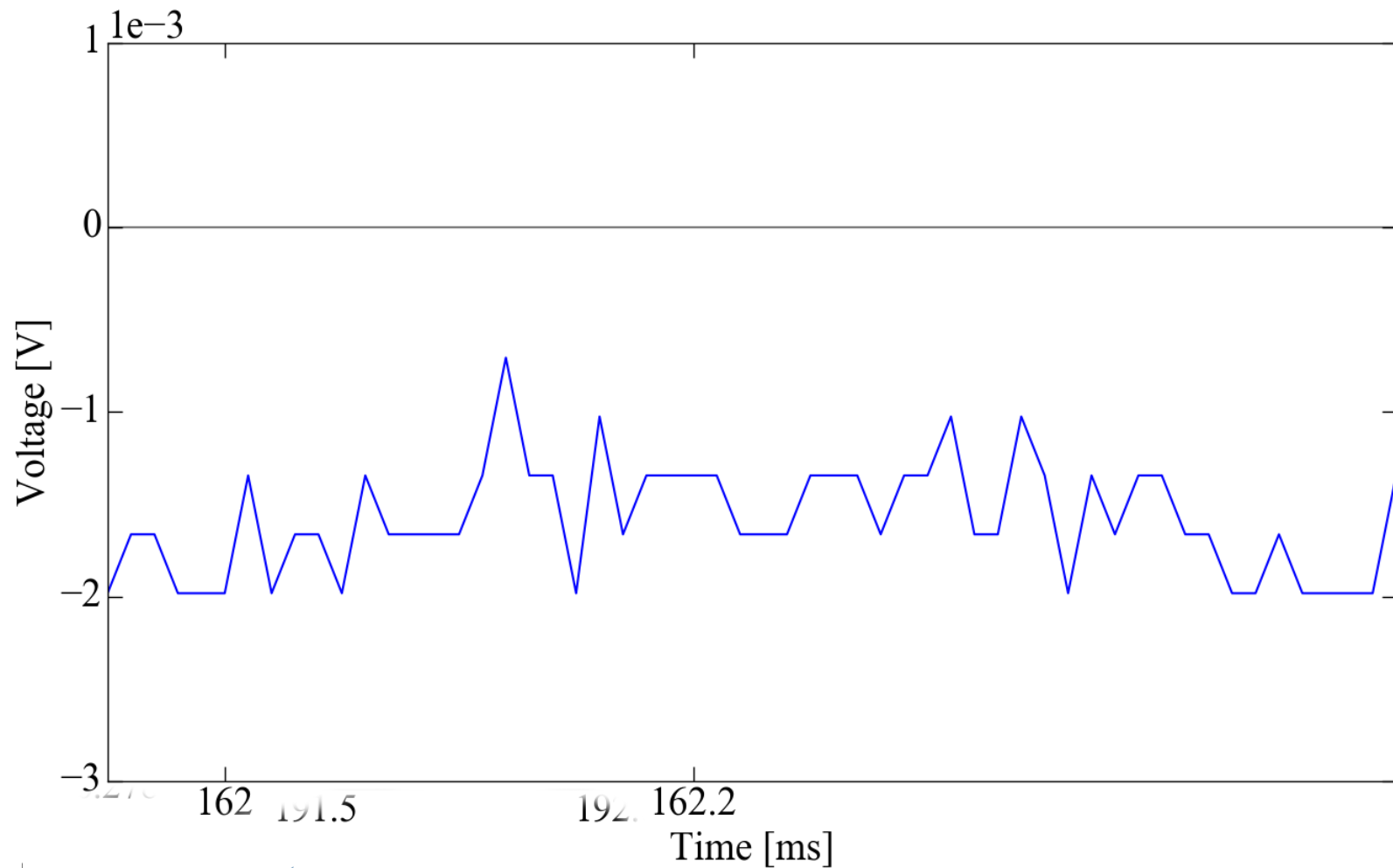
Rapid Compression Machine



What do we measure?

- The reaction chamber pressure is the primary measurement in the RCM
- A dynamic pressure transducer produces a charge output that is converted to a 0–10 V output
- Nominally, the initial voltage before compression is 0 V
- The voltage must be processed to compute the pressure, temperature, and ignition delay

What do we measure?



~~Problems~~ Engineering Opportunities

- The signal is noisy → Error in P_C
- There is an offset in the initial voltage → Error in P_0, T_C
- There are 25+ RCMs in the world, and everyone uses a different data analysis procedure
- Reproducibility is important!

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THIS WEEK

EDITORIALS

POSTDOCS More pay but fewer jobs on the way **p.438**

WORLD VIEW Treat antibiotic resistance as an ecological crisis **p.439**



DRONES Tiny flying robots with power to stick around **p.441**

Reality check on reproducibility

A survey of Nature readers revealed a high level of concern about the problem of irreproducible results. Researchers, funders and journals need to work together to make research more reliable.

Let's use Python to write a data analysis framework with the following goals:

1. Reproducible analysis across researchers
2. Documented design choices for filter criteria, etc.
3. Citable, open-source publication of code

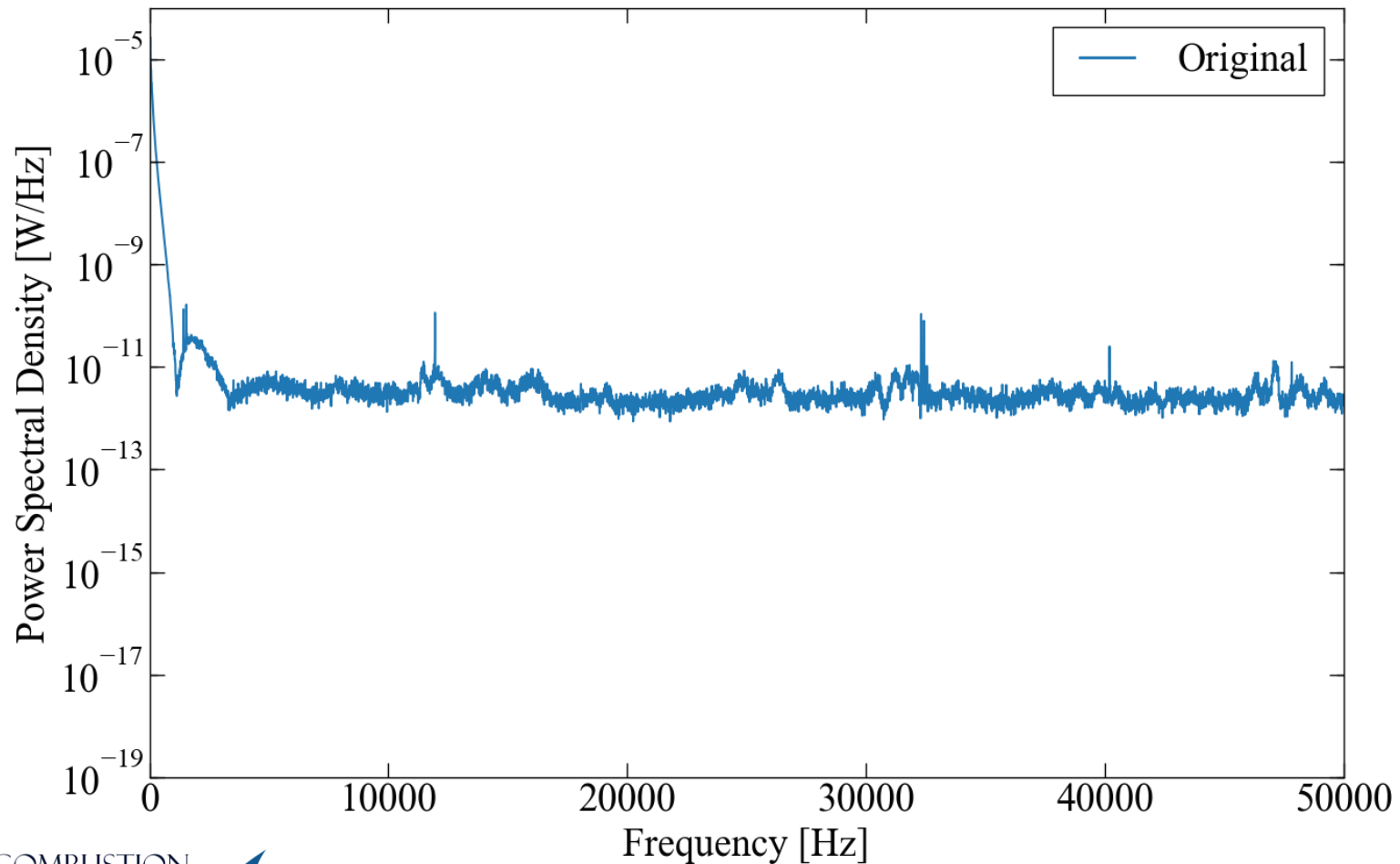
UConnRCMPy

<https://github.com/bryanwweber/UConnRCMPy>

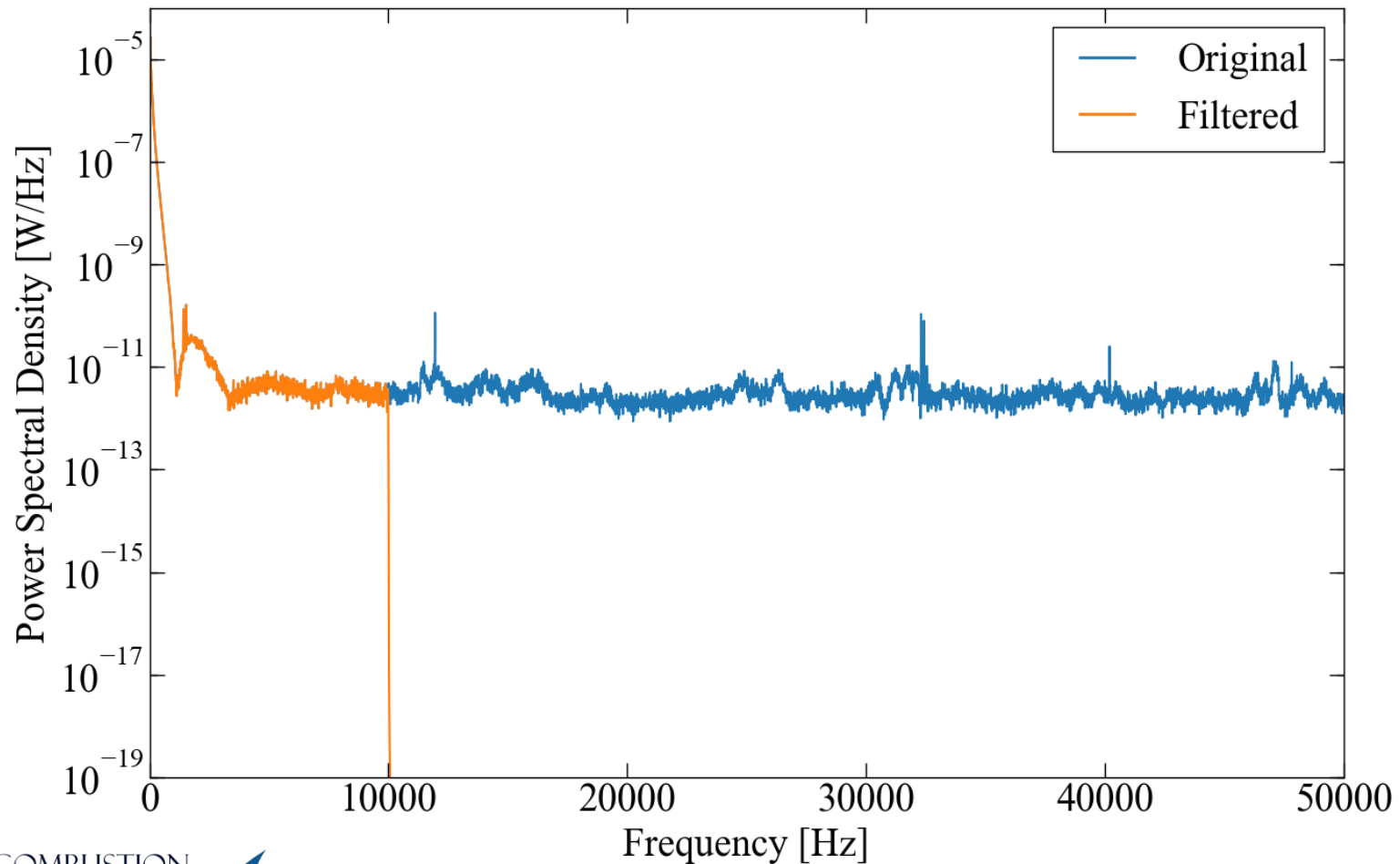
Features of UConnRCMPy

- Filtering and smoothing the raw voltage generated by the pressure transducer
- Converting the voltage trace into a pressure trace using settings recorded from the equipment
- Processing the pressure trace to determine parameters of interest in reporting the experiments, including the ignition delay and machine-specific effects on the experiment
- Conducting simulations utilizing the experimental information to calculate the temperature during the experiment

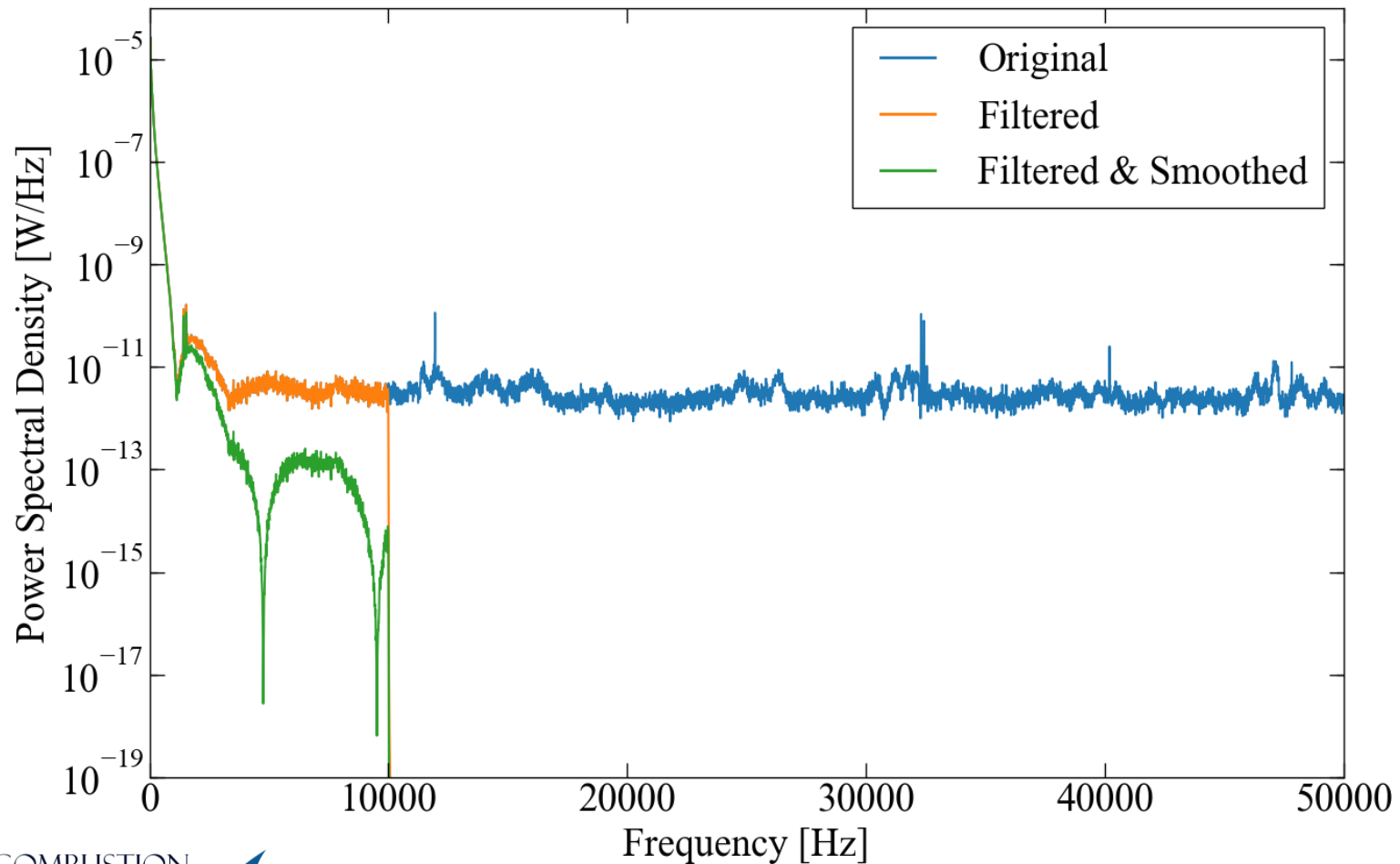
Filtering and Smoothing



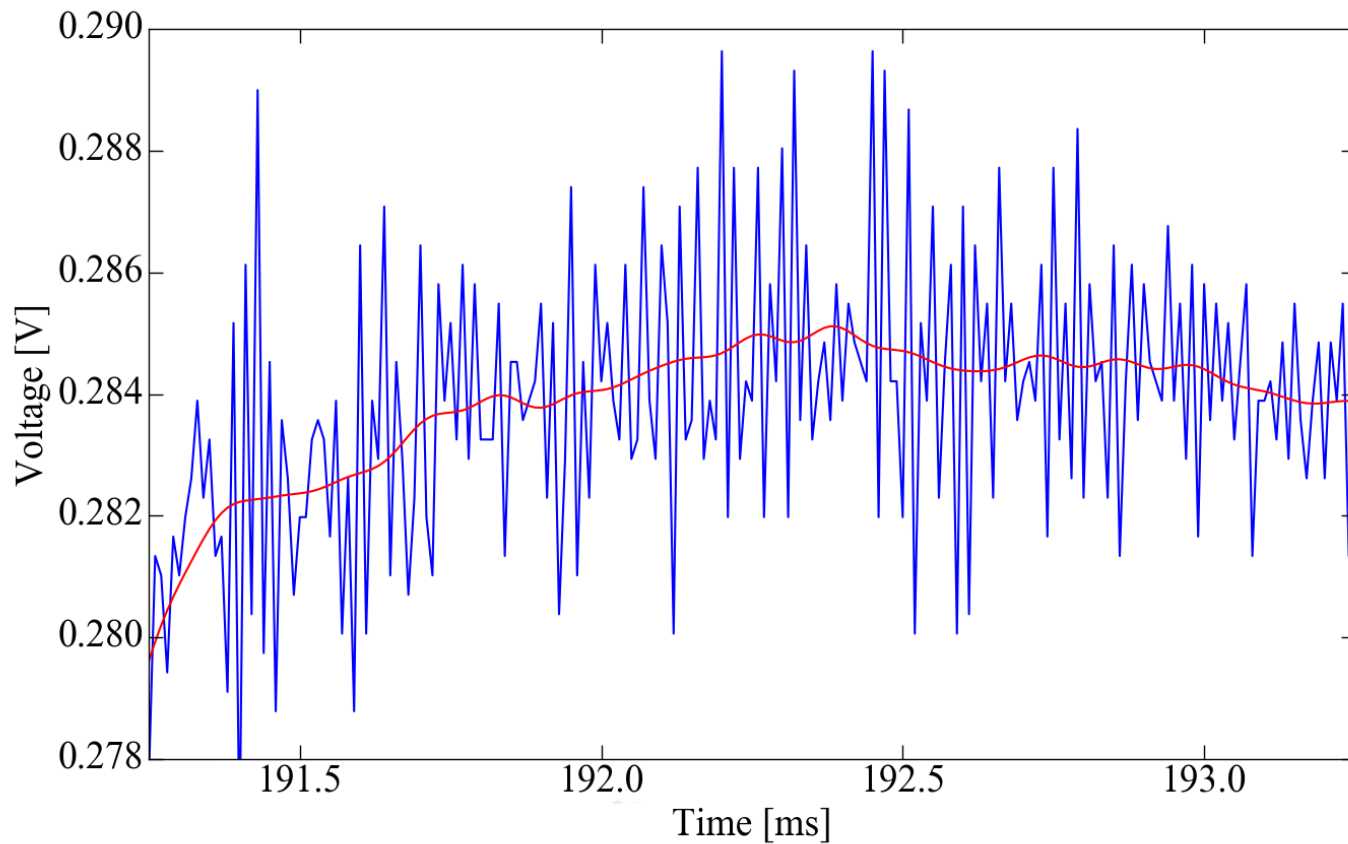
Filtering and Smoothing



Filtering and Smoothing

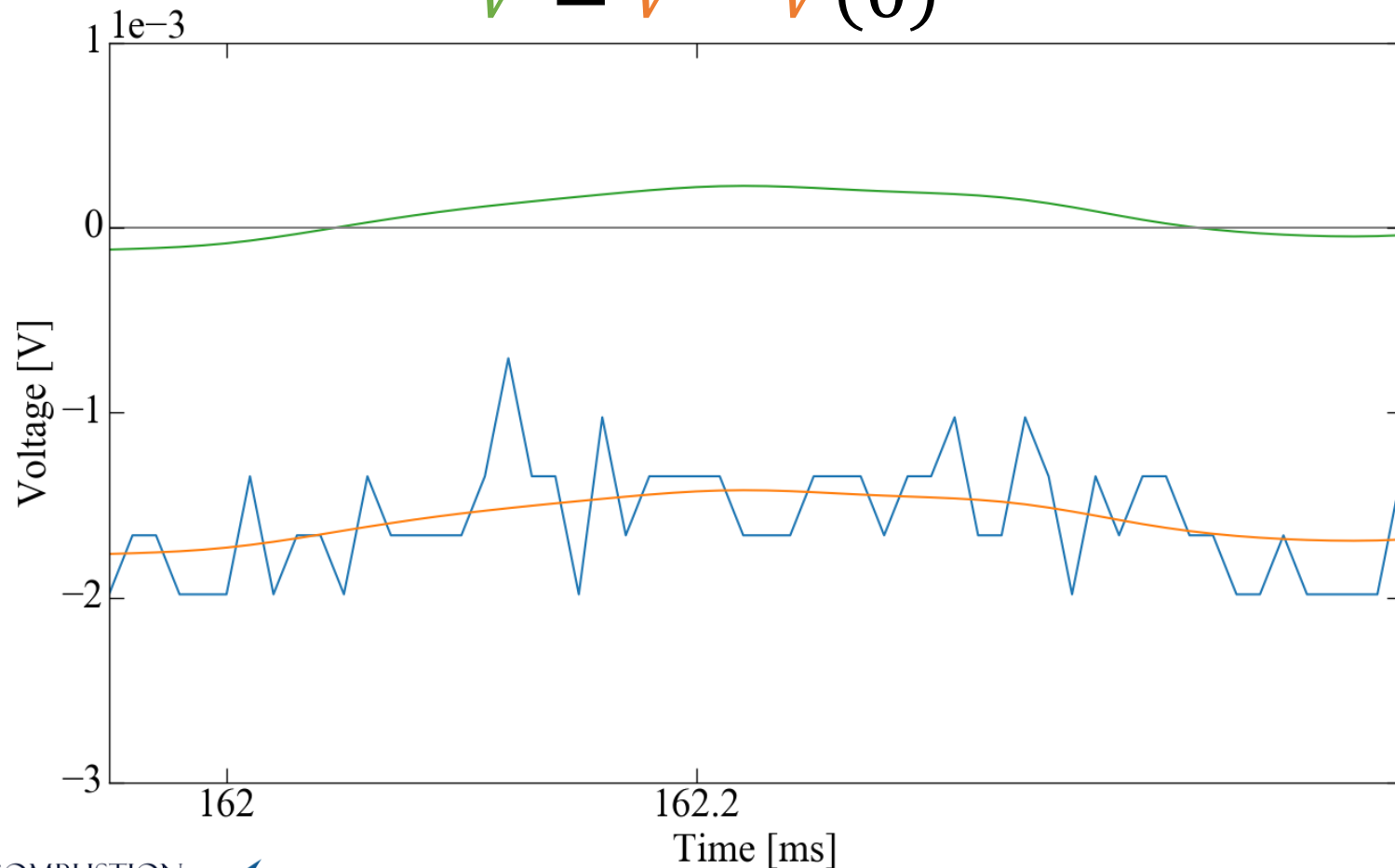


Filtering and Smoothing

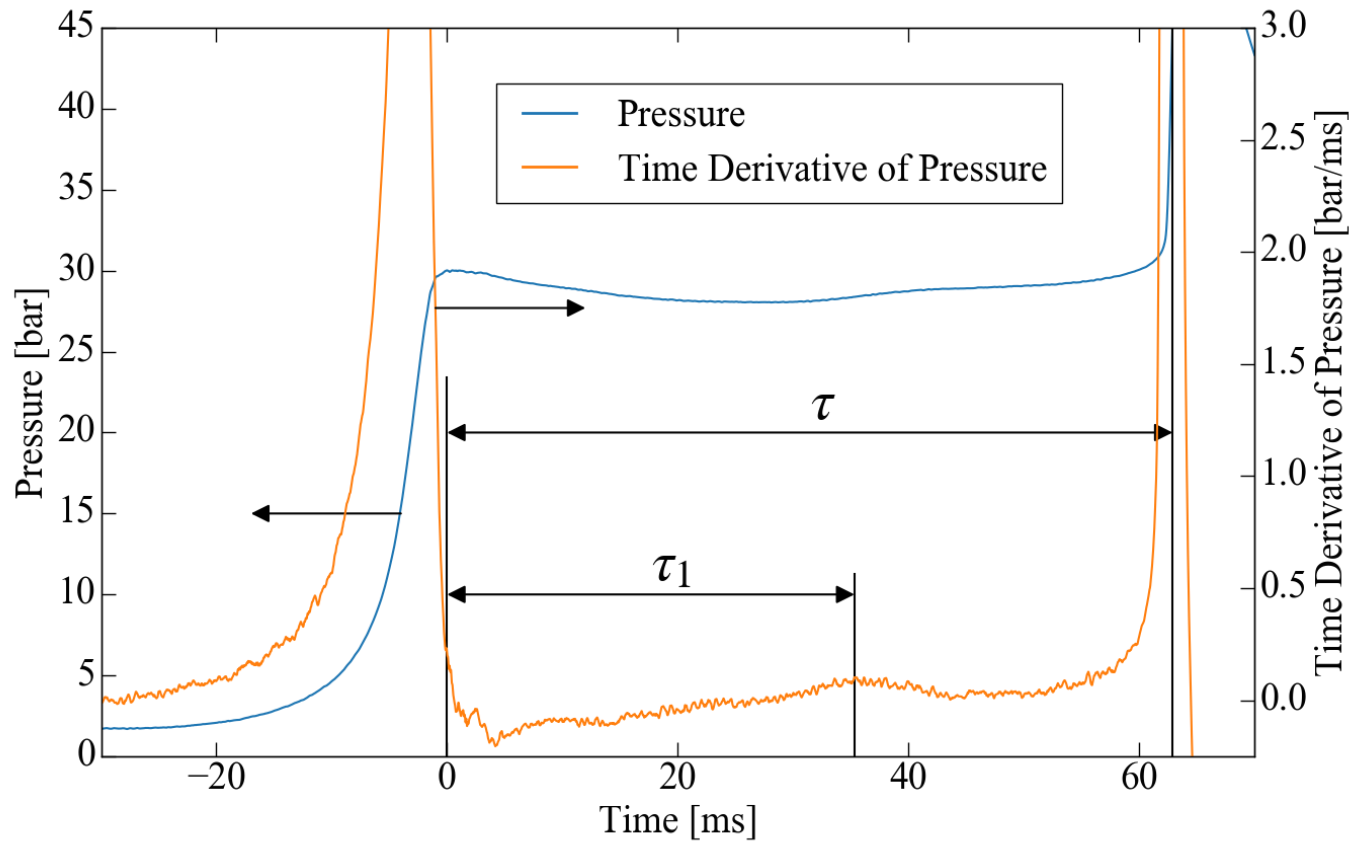


Correcting the Offset

$$V = \overline{V} - \overline{V}(0)$$



Computing Ignition Delay

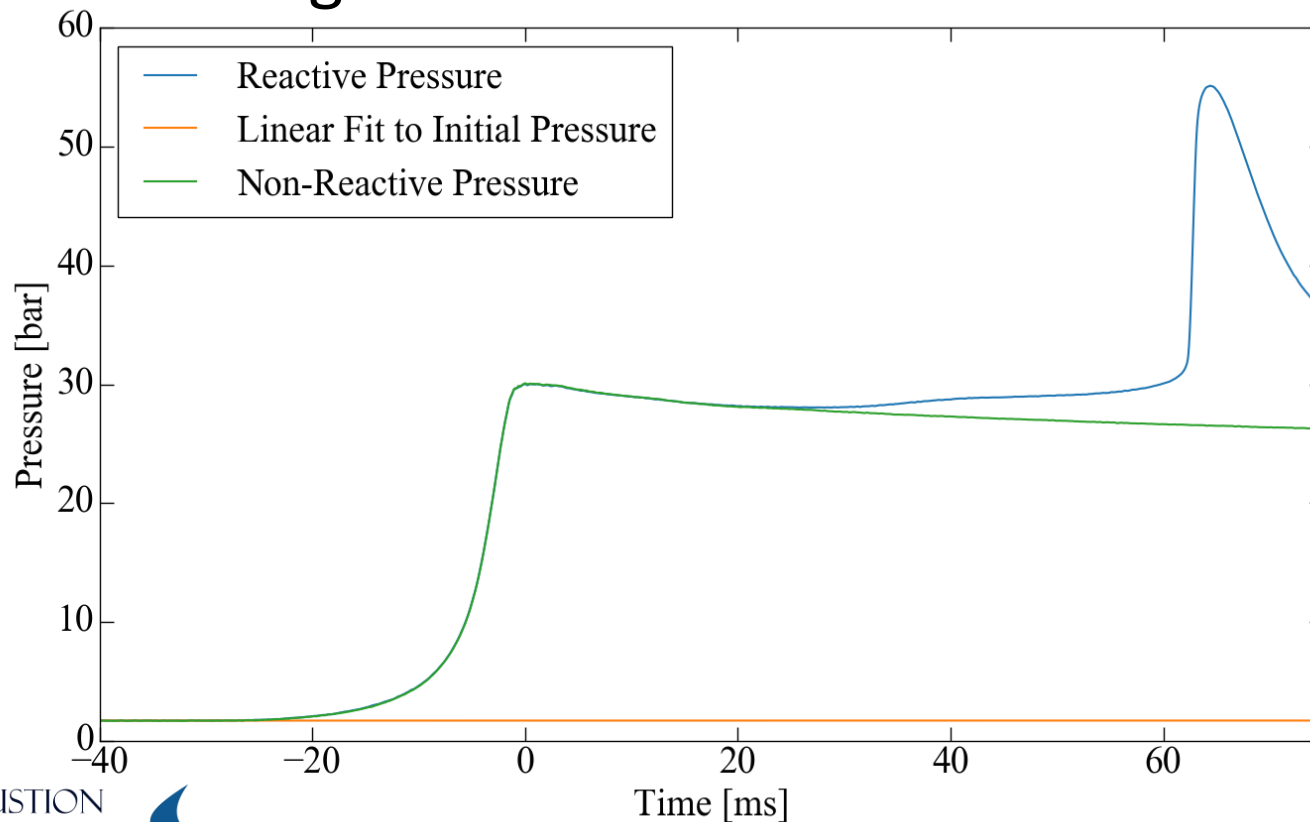


$$P = V * F + P_0$$

$$\tau = \max \left(\frac{dP}{dt} \right)$$

Modeling facility effects

- Replace oxygen with nitrogen and run the experiment again



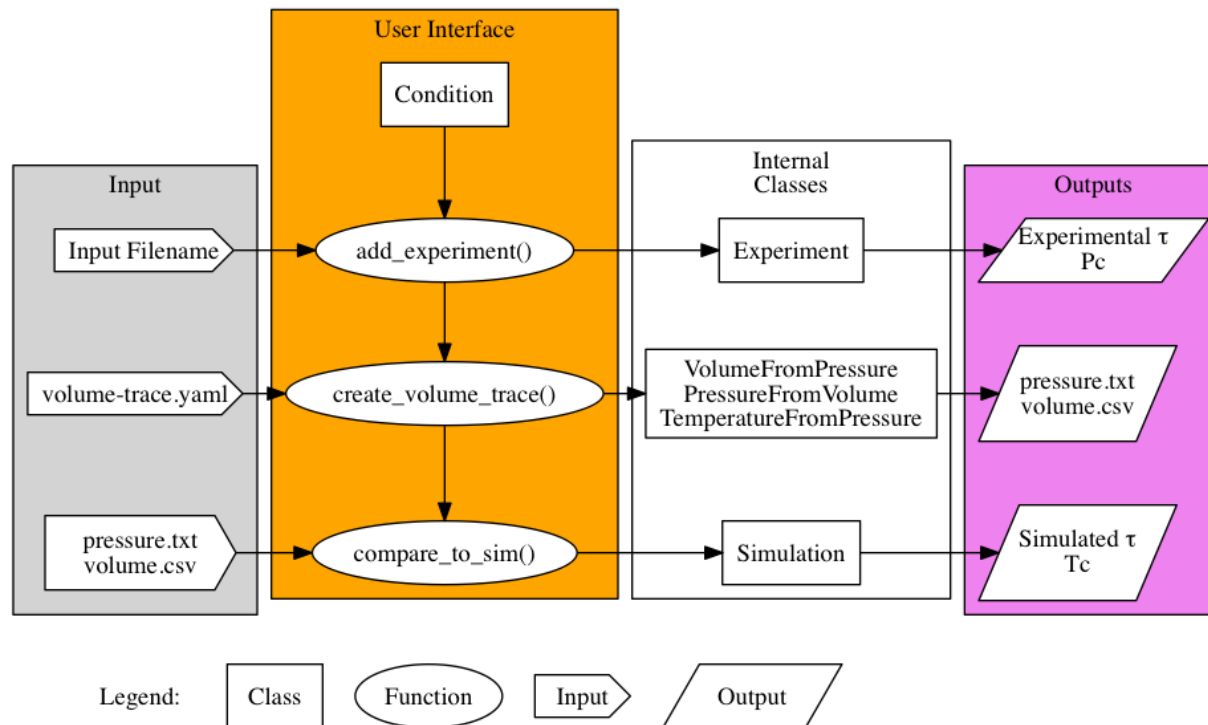
Modeling Facility Effects

- The reaction chamber is modeled as undergoing adiabatic compression followed by adiabatic expansion, to reproduce the non-reactive pressure trace
- The temperature at the EOC is found by applying the compression/expansion process to the law of conservation of energy

$$c_v \frac{dT}{dt} = -P \frac{dv}{dt} - \sum_k u_k \frac{dY_k}{dt}$$

Modular Design

- Enables modifications for different file formats with consistent choices of filtering criteria, etc.



Scientific Python Software

- SciPy for filter construction and convolution (fftconvolve was ~100x faster than NumPy's convolve)
- Cantera (<https://github.com/Cantera/cantera>) to calculate thermodynamic information about the reactor
- Matplotlib for plots
- Documentation is available online (<http://bryanwweber.github.io/UConnRCMPy/>), generated by Sphinx

Demo

SciPy 2016 UConnRCMPy demo

```
In [1]: import uconnrcmpy as ucr
import os
from pathlib import Path
import yaml
%matplotlib
os.listdir('.')
```

Using matplotlib backend: MacOSX

```
Out[1]: ['.DS_Store',
'.ipynb_checkpoints',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1626.txt',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1633.txt',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1640.txt',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1646.txt',
'00_in_02_mm_373K-1285t-100x-19-Jul-15-1620.txt',
'demo.ipynb',
'NR_00_in_02_mm_373K-1278t-100x-19-Jul-15-1652.txt',
'species.cti']
```

Create the Condition

```
In [2]: cond_00_in_02_mm = ucr.Condition()
```

Start adding experiments using the input field

```
In [3]: cond_00_in_02_mm.add_experiment()
```

Filename: 00_in_02_mm_373K-1282t-100x-19-Jul-15-1626

Finish the reactive experiments using the Path argument to add_experiment()

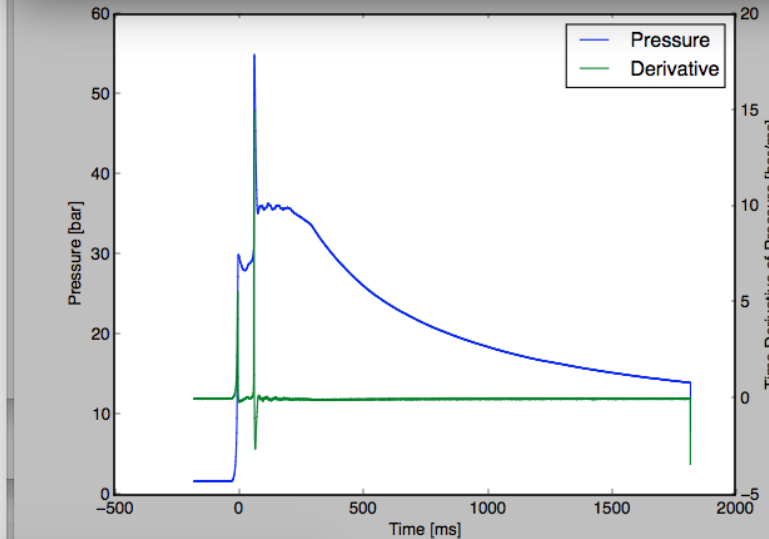
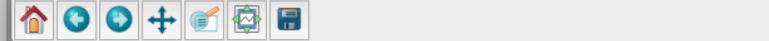
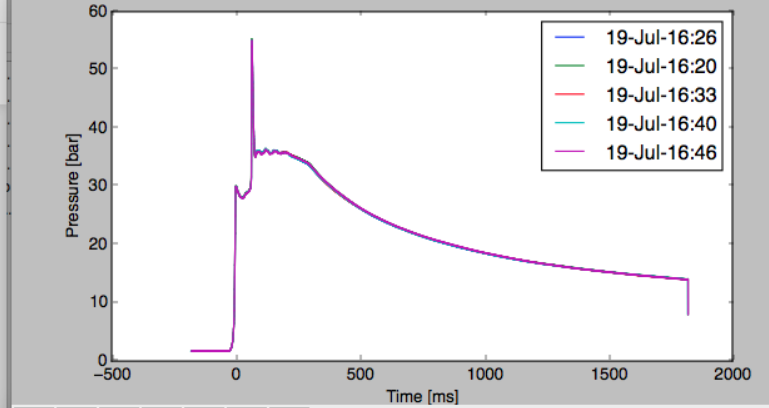
```
In [4]: cond_00_in_02_mm.add_experiment(Path('00_in_02_mm_373K-1285t-100x-19-Jul-15-1633.txt'))
cond_00_in_02_mm.add_experiment(Path('00_in_02_mm_373K-1282t-100x-19-Jul-15-1626.txt'))
cond_00_in_02_mm.add_experiment(Path('00_in_02_mm_373K-1282t-100x-19-Jul-15-1626.txt'))
cond_00_in_02_mm.add_experiment(Path('00_in_02_mm_373K-1282t-100x-19-Jul-15-1626.txt'))
```

Determine which is the representative experiment and add it to volume-trace.yaml

```
In [ ]: reacfile = '00_in_02_mm_373K-1282t-100x-19-Jul-15-1633.txt'
y = {'reacfile': reacfile, 'comptime': 33}
with open('volume-trace.yaml', 'w') as yaml_file:
    yaml.dump(y, yaml_file, default_flow_style=False)
```

Add the non-reactive experiment, still using add_experiment()

```
In [ ]: cond_00_in_02_mm.add_experiment()
```



f_x	Time											
	A	B	C	D	E	F	G	H	I	J	K	L
1	Time	P0	T0	Pc	Tig	First Stage	Tc	in Spacers	mm Shims	Right Tank		
2	1307	878	413	29.92496204	11.88		895.5559595	2	0	O2	252	
3	1314	880	413	29.96528205	11.68		903.4760968	2	0	N2	2399	
4	1322	880	413	30.04718894	11.58		896.9368688	2	0	DME	2399	
5	1329	880	413	30.00451748	11.67		897.6781982	2	0	C3H8	2500	
6	1336	880	413	29.99434871	11.44		894.7526087	2	0	Pd	100	
7										Ta	23.3	
8				Mean	11.65							
9				Std. Dev.	0.16							
10				%	1.38							
11												

This non-reactive trace matches well, so we can proceed. First, define the remaining quantities in volume-trace.yaml

```
In [8]: nonrfile = 'NR_00_in_02_mm_373K-1278t-100x-19-Jul-15-1652.txt'
y = {'**y', 'nonrend': 400, 'reacend': 80, 'nonrfile': nonrfile}
with open('volume-trace.yaml', 'w') as yaml_file:
    yaml.dump(y, yaml_file, default_flow_style=False)
```

Then we need to create the volume trace used for modeling

```
In [9]: cond_00_in_02_mm.create_volume_trace()
```

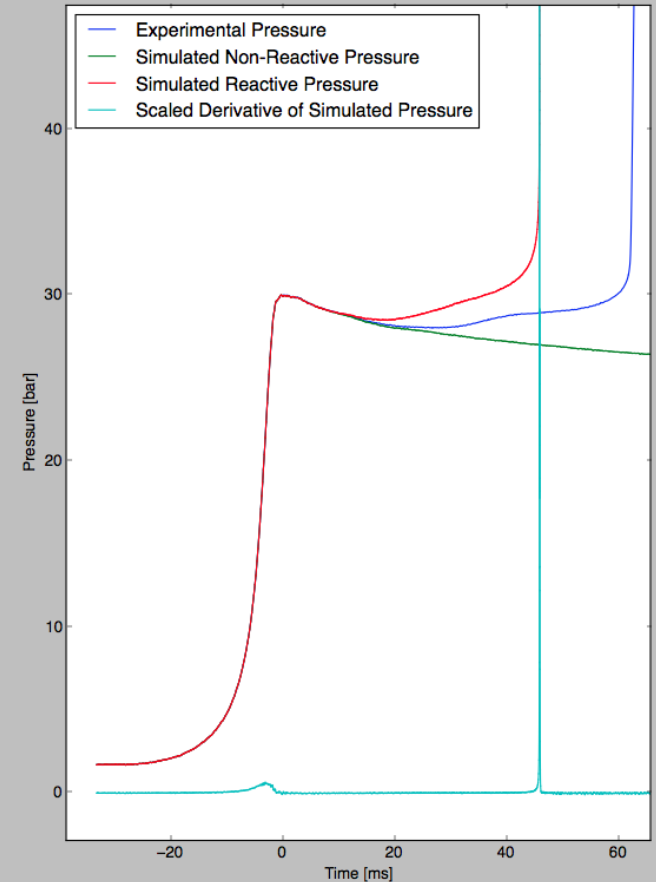
Then run the simulation to determine T_C

```
In [12]: cond_00_in_02_mm.compare_to_sim(run_reactive=True, run_nonreactive=True)
```

```
Are you sure you want to overwrite the nonreactive simulation? Input y or n: n
Nothing was done
Are you sure you want to overwrite the reactive simulation? Input y or n: y
759      46.116253
```

```
In [13]: os.listdir('.')
```

```
Out[13]: ['.DS_Store',
'.ipynb_checkpoints',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1626.txt',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1633.txt',
'00_in_02_mm_373K-1282t-100x-19-Jul-15-1640.txt',
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'00_in_02_mm_373K-1285t-100x-19-Jul-15-1620.txt',
'demo.ipynb',
'NR_00_in_02_mm_373K-1278t-100x-19-Jul-15-1652.txt',
'species.cti',
'Tc_P0_T0_373K_pressure.txt',
'volume-trace.yaml',
'volume.csv']
```



Future Work

- Automatic calculation of optimal low-pass filter cutoff frequency
- Automatic calculation of optimal moving-average smoothing filter width
- Improved detection of the EOC
- Unit testing!
- See <https://github.com/bryanwweber/UConnRCMPy/issues>

Acknowledgements

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Email: bryan.w.weber@gmail.com

Github: [@bryanwweber](https://github.com/bryanwweber)

Web: bryanwweber.com

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