

Rapid IR Fuel Screening

(Previously: Shock Tube and Flow Reactor Studies of the Kinetics of Jet Fuels)

Stanford University

PI: Ronald K. Hanson

PM: Anna Oldani

Cost Share Partner: Stanford University

Research Approach:

Develop statistical models that **correlate the physical and chemical properties of a fuel** (e.g., boiling point, heat of combustion, flash point, etc.) **with its vapor-phase FTIR spectrum.**

Apply these models to **predict the physical and chemical properties of next-generation SAFs** and fuel components.

Objective:

Develop a **compact, low-volume prescreening tool** for the prediction of physical and chemical properties of sustainable aviation fuels (SAFs) using Fourier-Transform Infrared (FTIR) spectrometry and advanced statistical analysis methods.

Project Benefits:

FTIR prescreening approach will **make SAF design and approval process less costly and more efficient.**

This low-volume (<1 mL) method yields insights that are **complementary to other prescreening approaches** (e.g., GCxGC).

Major Accomplishments (to date):

- An FTIR spectrometer facility has been modified to enable wide-spectrum IR measurements (2-16 μm)
- Separate **models have been trained for key physical/chemical properties** using the expanded dataset of pure hydrocarbons/real fuels
- The **models accurately predict the properties of all fuels** in the training dataset

Future Work / Schedule:

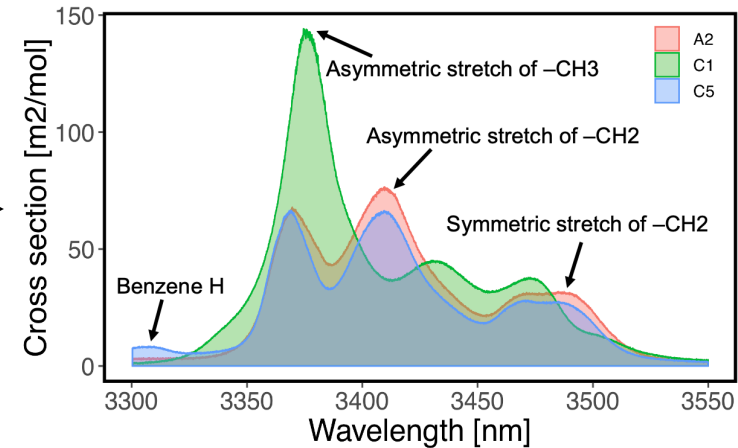
- **Continue expanding the training dataset** to include more fuels (>C8) and additional property data
- Optimize models and begin utilization to **predict the properties of SAFs** and SAF components
- Compare IR analysis approach with alternate prescreening methods

Introduction

Motivation:

- The vapor-phase IR absorption spectrum of a hydrocarbon fuel contains quantitative information about the fuel's molecular structure and functional groups
- Statistical models can be used to infer the physical and chemical properties of fuels from this spectral information

Fuel structure is evident in the IR spectra; shape and height of absorption features reflects the type and number of functional groups, and can be **correlated with physical/chemical properties**



Approach:

- Develop a training dataset based on broad (2-16 μm) vapor-phase spectra of fuels and their key physical and chemical properties
- Build statistical models using the training data to directly correlate fuel properties with their FTIR spectra; four different models under consideration
- Use the optimized models to predict the physical and chemical properties of next-generation SAFs and SAF components

FTIR Spectrum
[2-16 μm]

Model 1: Regularized Linear Model

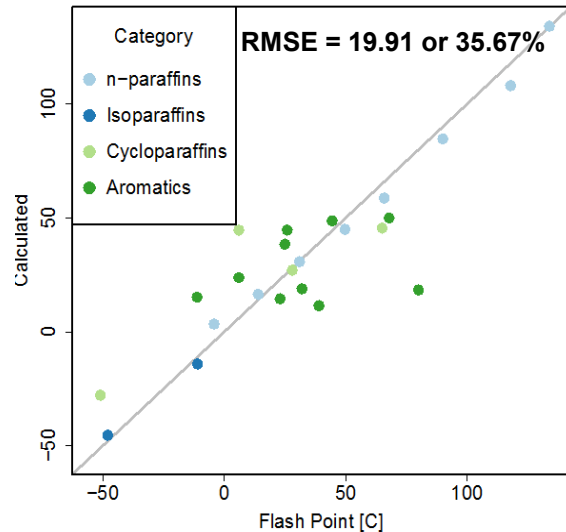
Properties

- Introduced the benefits of utilizing the extended 2-16 μm spectral range vs. 3.4- μm region only

Property prediction example: Flash Point ($^{\circ}\text{C}$)

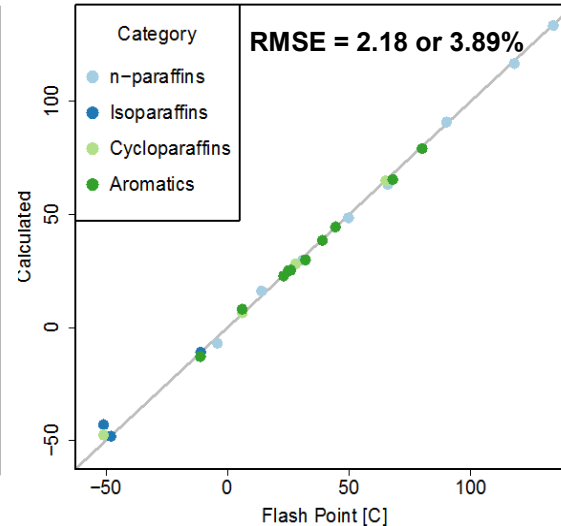
3.4 μm region only

CVE = 24.42 or 43.58%, $R^2 = 0.81$

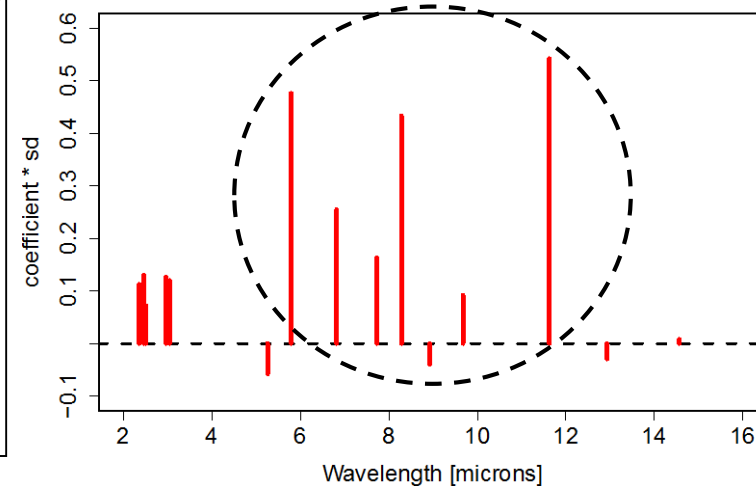


2-16 μm region

CVE = 7.89 or 14.11%, $R^2 = 0.99$



Relative contribution of selected wavelengths



- Previous wide-spectrum analysis relied exclusively on spectra available in public Pacific Northwest National Lab (PNNL) database
- Research need:** measurement of the 2-16 μm spectra of **real fuels and pure hydrocarbons** not included in the PNNL database

Recent Progress

- The **FTIR experimental setup was modified** to enable measurement of gas-phase fuel spectra across the full 2-16 μm region
 - Heated-cell windows replaced to enable wide-spectrum transmission, purge system constructed

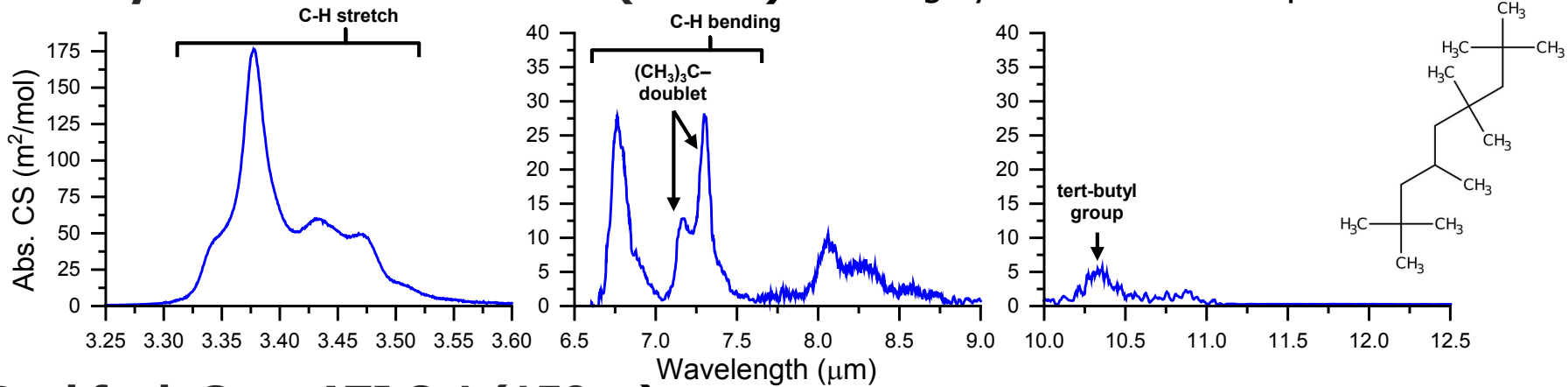


- Stanford has now measured 15 pure hydrocarbons and real fuels, expanding our 2-16 μm training dataset to 71 fuels:

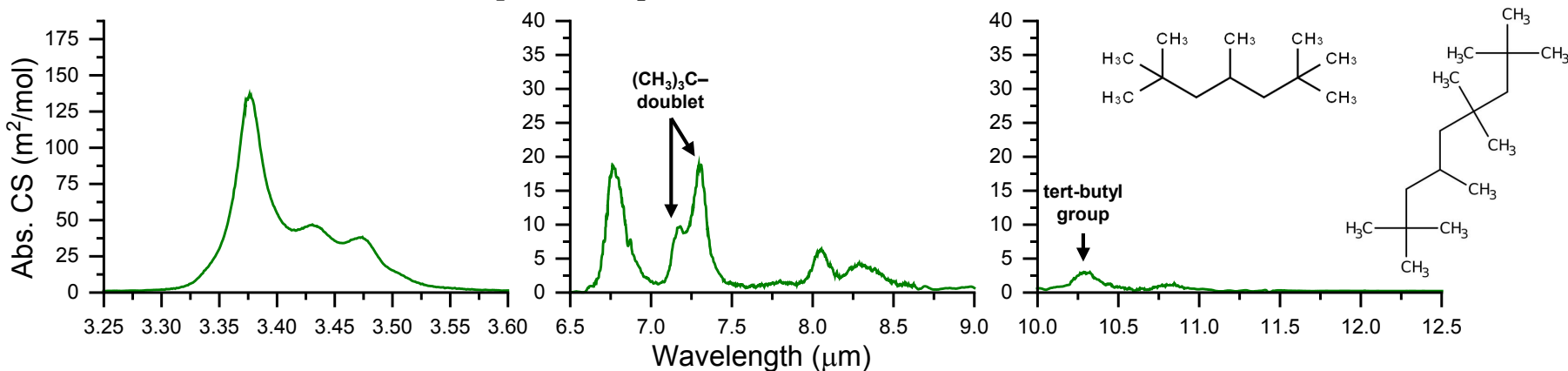
Fuel category	Stanford-Measured Spectra	PNNL Spectra	
n-paraffins	2	8	} C5-C16 neat hydrocarbons
Isoparaffins	2	5	
Cycloparaffins	3	4	
Aromatics	1	15	
Multi-comp. blends	0	24*	
Real fuels	7	0	*weighted combination of individual components
Total	15	56	71 Fuels

Example of Newly Measured Spectra

Pure Hydrocarbon: Isocetane (150 °C) → highly branched C16 isoparaffin



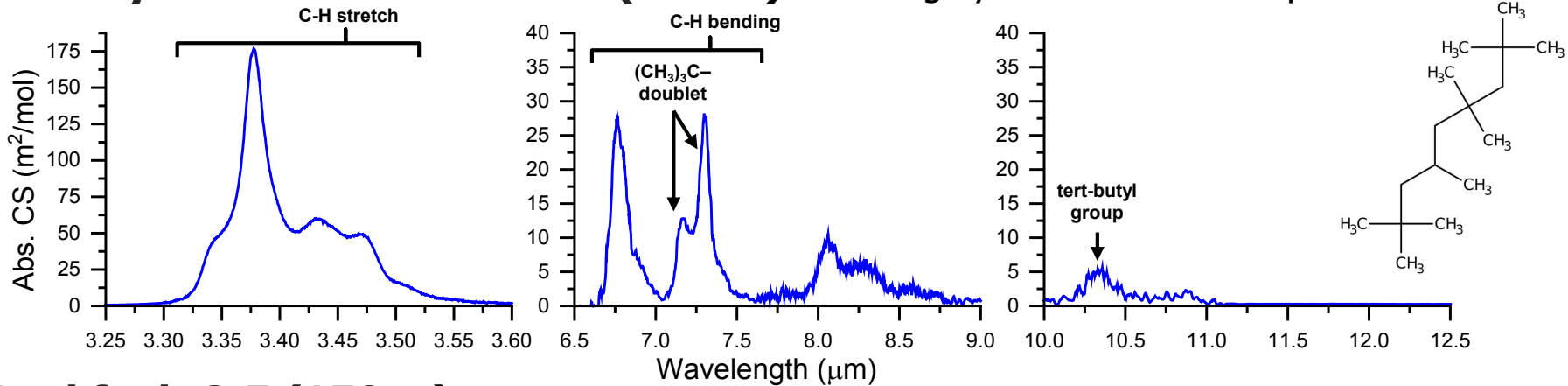
Real fuel: Gevo ATJ C-1 (150 °C) → predominantly isocetane, 2,2,4,6,6-pentamethylheptane



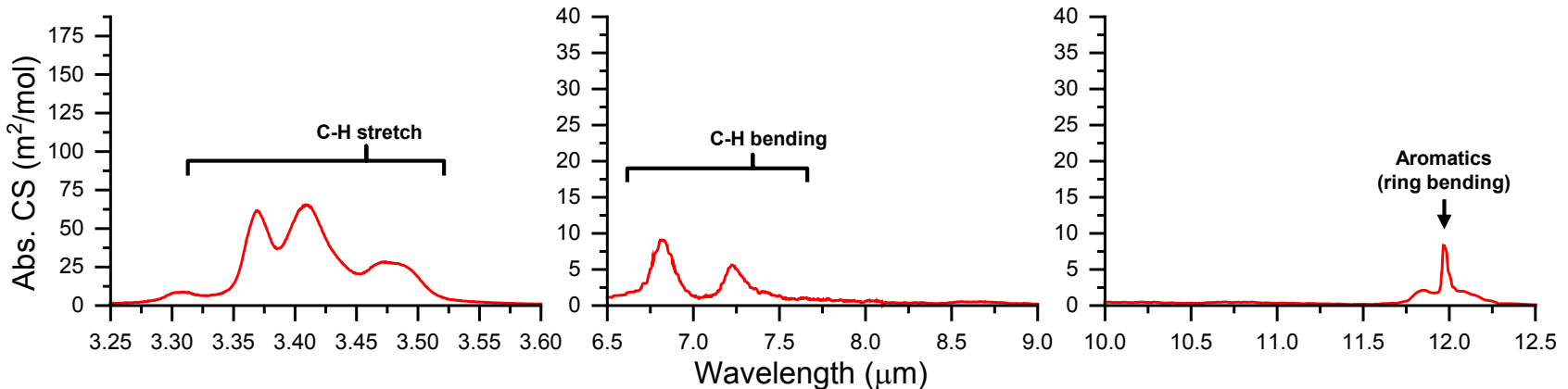
- Using multiple fuel concentrations enables **good signal-to-noise ratio** across the entire 2-16 μm range
- As expected, prominent **features seen in the isocetane spectrum also appear in the C-1 spectrum**

Example of Newly Measured Spectra

Pure Hydrocarbon: Isocetane (150 °C) → highly branched C16 isoparaffin



Real fuel: C-5 (150 °C) → 73% C10 isoparaffins, 27% trimethylbenzene



- Using multiple fuel concentrations enables **good signal-to-noise ratio** across the entire 2-16 μm range
- **Clear differences observed between the isocetane and C-5 spectra** (shape, height, location of spectral features)

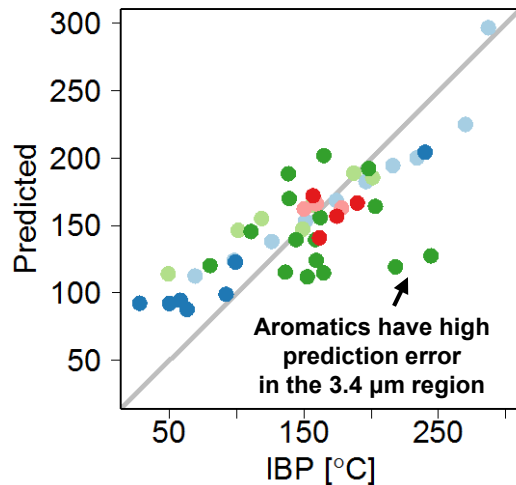
Model Performance (2-16 μm)

- The prediction accuracy of the previous 3.4 μm model and the new 2-16 μm model are compared for **initial boiling point (IBP)**

Earlier Model

3.4 μm region (N = 24, $\alpha = 1$)

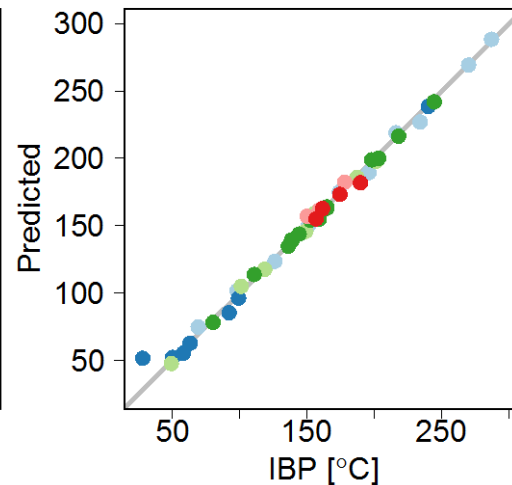
MAE = 26.98 or 17.72%, 47 pts



Current Model

2-16 μm region (N = 496, $\alpha = 0.9$)

MAE = 2.96 or 1.95%, 47 pts



CATEGORY

- n-paraffin
- Isoparaffin
- Cycloparaffin
- Aromatics
- Jet A
- C fuel

Training cross-validation error

Spectral region	CVE [°C]	CVE (%)	R ² value
3.4 μm	28.82	18.93%	0.59
2-16 μm	8.55	5.61%	0.99

- Using the expanded training dataset and the **full 2-16 μm region greatly improves the prediction accuracy** due to the relatively large contribution of wavelengths outside the 3 μm region to the variation in IBP
- The high R² value indicates that the model fits the training data well, and the low cross-validation error (CVE) highlights the potential of this approach for making **highly accurate future property predictions**

Summary and Future Work



- IR spectral analysis offers a **robust, low-volume way to predict the physical and chemical properties of fuels**
- Modification of our FTIR spectrometer setup has **enabled measurement of gas-phase, 2-16 μm spectra** of relevant hydrocarbon fuels and SAFs
- Regularized linear models trained on the expanded training dataset show **excellent predictive performance for all the properties considered**

Next steps:

- **Continue use of FTIR to expand the current training dataset** to include more pure hydrocarbons (larger than C8) and real fuels
- Continue optimization of models, use them to **predict the properties of SAFs** and SAF components, and compare model performance with alternate methods
- ASCENT collaboration (J. Heyne, U. Dayton): investigate the **impact of isomeric structure on ignition properties**