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Introduction

Ambitious goals are being set for ramping up sustainable aviation fuels (SAF) to help reduce greenhouse gas (GHG) emissions from aviation. Those include the US SAF Grand Challenge to increase supplies to 3 and 35 billion gallons per year by 2030 and 2050, respectively¹ and the European objective of deriving 63% of fuels for air transport from sustainable sources by 2050.² Significant biomass resources exist in the US,³ and with the development of those feedstock supplies and synergistic conversion processes, sufficient feedstocks will be available to meet the 2050 SAF volumetric goals.⁴ The most efficient conversion and use of feedstocks towards meeting SAF specifications and GHG goals is an important consideration. Various conversion technologies, including gasification and the conversion of bio-derived ethanol to jet are promising near-term options.⁴ Pyrolysis based processes promise higher yields but have inherent challenges towards

Prediction of sustainable aviation fuel properties for liquid hydrocarbons from hydrotreating biomass catalytic fast pyrolysis derived organic intermediates[†]

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Catalytic fast pyrolysis (CFP) of wood, woody residues, and agricultural waste has the potential to produce organic liquid intermediates that can be hydroprocessed to sustainable aviation fuels (SAF). In this article we present results of property predictions for two SAF fuels produced from hydroprocessing of CFP-based oxygenated organic intermediates followed by distillation to obtain the jet boiling range cuts. The two intermediates, in turn, were produced using ZSM-5 and Pt/TiO₂ catalysts. The objective is to assess the applicability of recently developed prediction methods and other well-established methods to CFP-derived SAF fuels, focusing on five properties that are important for jet fuel certification, namely density, flash point, net heat of combustion, freezing point, and distillation temperatures (10% distilled and final boiling point). For flash point and net heat of combustion, several of the methods tested gave comparable results with mean absolute errors (MAE) that are within the reproducibility limit of the data. The API and ASTM D7215 methods provided the best results for flash point, while the ASTM D3338 correlation gave the lowest MAE for net heat of combustion. For density at 15 °C, the MWA method of Shi *et al.* gave the lowest MAE of 0.0033 g cm⁻³, however, it is still larger than the reproducibility limit of 0.0005 g cm⁻³. The accuracy of the prediction methods for freezing point could not be assessed since the experimental data of the two SAF fuels were given as <-70 °C. The distillation temperatures were not well predicted by the method tested.

meeting the stringent requirements for aviation fuels, stemming primarily from the high oxygen content, corrosiveness, and chemical instability of the bio-oil intermediate. Thousands of different compounds are released during pyrolytic breakdown of biomass⁵ and funneling them towards specific molecules is difficult. In addition, variations in feedstock compositions and process conditions also manifest themselves as variations in the pyrolysis products. Towards the goal of getting a more stable and consistent slate of products, catalysts are used to deoxygenate and stabilize pyrolysis vapors in a process known as catalytic fast pyrolysis (CFP).6,7 Further hydrotreatment of condensed CFP oils is necessary to produce hydrocarbons for SAF, and this step provides a critical boost towards compositional compatibility and predictability of fuel quality. Enabling SAF from CFP will require careful study and understanding of product properties relevant for jet fuel certification. CFP oils contain aromatic structures (hydrocarbons, phenolics and higher hydroxyaromatics) that could be converted to cycloparaffins, which have been identified as desirable SAF components.4 With suitable hydrotreating conditions, including temperature, it is possible to produce SAF with high fractions of cycloparaffins.8 The composition of the product differs from most other bio-based SAF fuels; high concentrations of cyclohexanes, octahydroindenes, and

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decahydronaphthalenes are present in this product. Towards the goal of advancing the future adoption of CFP-derived SAF, the aim of this work is to employ model-based methods from literature to evaluate and connect recently available experimental information⁸ with composition-based prediction methods. These property predictions and related understanding from experimental speciation can help accelerate the understanding and adoption of SAF from CFP and other conversion processes and help inform process and catalyst designs; prediction of key product properties using known compositional measurement techniques that require small sample quantities can facilitate those efforts. In this work we focus on evaluation of properties with available experimental measurements (density, flash point, net heating value, freezing point, distillation temperatures for 10% distilled, and final boiling point) and plan to discuss additional properties in future publications.

Catalytic fast pyrolysis for fuels production

A simple block flow diagram of the CFP process,6,7 including the hydroprocessing and product recovery steps is shown in Fig. 1. Biomass contains significant amounts of oxygen (~40 wt% in dry wood), carbon, and hydrogen, as well as nitrogen and sulfur. Other components, including inorganic mineral matter in varying proportions, dictated by the biomass source and type, are also present. Efficient removal of all other atomic constituents while preserving carbon and hydrogen for making hydrocarbon molecules in the correct jet-fuel boiling range is the key for high yields of SAF. After thermal breakdown of biomass into solid char, permanent gases, and oxygenated vapors during fast pyrolysis, subsequent deoxygenation during catalytic upgrading of pyrolysis vapors is important for producing and retrieving (after condensation) a stabilized organic intermediate or CFP oil. This CFP oil can then be further hydroprocessed to hydrocarbon fuels and blendstocks via removal of the remaining oxygen, sulfur, and nitrogen. Note that nitrogen and sulfur are usually minor species (<0.2 wt%) in woody biomass, and efficient deoxygenation is the primary chemical challenge for CFP processes using woody feedstocks.

CFP processes can provide high hydrocarbon yields with effective chemistry to preserve a large fraction of the carbon in biomass in larger molecules and not forming CO, CO_2 , and other light gases during catalytic upgrading; many processes propose the use of hydrogen for removing most of the oxygen *via* hydrodeoxygenation, thus reducing coke, CO, and CO_2 formation.⁹ In addition to process challenges, especially with the design and maintenance of efficient catalysts with high

target yields, making aviation-compatible molecules and the removal of contaminants are other challenges that need to be addressed to realize the full potential of this technology towards SAF production.

For the two CFP-derived fuel sample analyses used in this article, carbon efficiencies from biomass to hydrotreated products in the SAF boiling range were calculated to be 11% and 13% using ZSM-5 and Pt/TiO₂ CFP catalysts, respectively. These overall carbon efficiencies are similar to those that can be estimated from other published work, e.g., 7-12% in the SAF range based on 25-42% overall carbon efficiencies and 28% of the distillation cuts in the jet boiling range.10 Carbon efficiencies are known to vary with oxygen contents of CFP oils; data points follow a pattern of reduced carbon efficiencies with increased deoxygenation during CFP.11 There are no standardized CFP oils and research efforts are currently focused on finding optimal catalysts and conditions for increasing the yields of specific fuels and products; our current work falls within the continuum of yields and oxygen contents,11 trying to address gaps and questions regarding the production of SAF via CFP.

Properties of interest for SAF

Aviation fuels and their properties have been studied and documented.12 Since SAF candidate products need to make near-term impact without disrupting the existing infrastructure or requiring any modifications to aircraft engines, SAF properties are held to similar standards as regular jet fuel. In cases where SAF from specific processes cannot meet the full set of aviation fuel properties, e.g., farnesane (2,6,10-trimethyldodecane, a C15 molecule derived from sugar fermentation), specific blend limits (10% for farnesane) are set to allow use within the current aviation fuel infrastructure.4 As a guide to the reader, some key properties, and their specifications of two commonly used jet fuels are reproduced in Table 1. The significance of these properties during aviation fuel use is discussed in the literature.4 Note that we will be focusing on a subset of the listed properties (density, flash point, net heating value, freezing point, distillation temperatures for 10% distilled, and final boiling point).

Importance of understanding the impact of chemical composition on properties

The importance and benefits of predicting properties of aviation fuels from chemical compositions was discussed by Vozka and Kilaz.¹³ Since chemical compositions can be measured

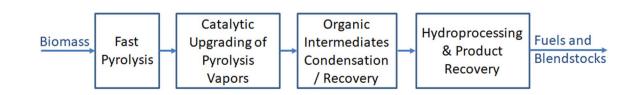


Fig. 1 Simplified block flow diagram of catalytic fast pyrolysis (CFP) process including hydroprocessing and product recovery steps.

Table 1 Specifications for select jet fuel properties¹²

Property	Jet A ^{<i>a</i>} , 2018	JP-8 ^b , 2018	
Density at 15 °C (g cm ^{-3})	0.775 to 0.84	0.775 to 0.84	
Flash point (°C)	38 min	38 min	
Viscosity (cSt) ^c	8 at −20 °C, max	8 at −20 °C, max	
Freezing point (°C)	-40 max (-47 for jet A-1)	-47 max	
Aromatics (vol%)	25, max	25, max	
Heat of combustion, (MJ kg^{-1})	42.8, min	42.8, min	
Sulfur content (mass%)	0.3	0.3	
Mercaptan sulfur (mass%)	0.003	0.002	
Acid number	0.1	0.015	
Distillation (°C)			
IBP^d			
10%	205	205	
20%			
50%			
90%			
FBP^d	300	300	
Smoke point (mm)	18 min	18 min	

^{*a*} Jet fuel commonly used in commercial aviation. ^{*b*} Jet fuel commonly used by the US military. ^{*c*} 1 cSt = 1 mm² s⁻¹. ^{*d*} IBP, initial boiling point, FBP, final boiling point.

using known methods and relatively small quantities of fuel, the ability to predict fuel properties from chemical compositions can enable rapid feedback during the development of conversion processes for alternative fuels production. In this context CFP processes that can vary significantly because of the numerous choices of feedstocks, process configurations, catalysts, and operating conditions can benefit from the identification of appropriate prediction methods and any necessary related developments to help accelerate the potential adoption of SAF from CFP through better understanding of fuel property variations due to chemical compositional changes of the fuel products. Although the eventual long-term testing is necessary for any SAF pathway certification,4 available experimental data corroborating property prediction methods can help build confidence that we have a good understanding of what it will take for the CFP family of processes to be compatible with SAF production.

Property predictions in literature

Yang *et al.*¹⁴ reviewed physicochemical properties of bio-jet fuels considering their chemical compositions. They suggested the inclusion of chemical compositions, including the classes of hydrocarbons and carbon chain lengths because they significantly influence fuel performance.

Vozka and Kilaz¹³ provided an overview of chemical composition–property correlation techniques from 1955 to 2019 covering four categories of analytical measurement techniques: nuclear magnetic resonance (NMR) spectroscopy, infrared (IR) spectroscopy, Raman spectroscopy, and gas chromatography (GC, one and two-dimensional). These techniques and their predictive capabilities were compared using measures of the uncertainty for future predictions. Many characteristics were considered, including smoke point, hydrogen content, net heat of combustion, freezing point, density at 15 °C, flash point, viscosities at -20 °C and -40 °C, calculated cetane index, and acid number. It was observed that chemical compositions from two-dimensional gas chromatography ($GC \times GC$) measurements yielded the best correlations for most property predictions in comparison with other measurement techniques. For smoke point, the most successful correlation was based on NMR results, although it was noted that $GC \times GC$ method was not tested. The viscosity prediction was most successful using GC, implying that $GC \times GC$ would work as well. For density, the GC \times GC method of Shi *et al.*¹⁵ gave a coefficient of determination (R^2) of 0.9999, while that of Vozka *et al.*¹⁶ yielded an R^2 of 0.9967 for their respective sets of aviation fuels evaluated; it was noted that the Shi et al. method predicted density at 20 °C instead of 15 °C. For freezing point, the Shi *et al.* method gave an R^2 of 0.9866; only the GC \times GC evaluation was reported.¹³ For net heat of combustion (NHC), Shi *et al.* was also the only $GC \times GC$ based evaluation reported, with an R^2 of 0.9993. The Shi *et al.* GC \times GC based method predicted flash point with an R^2 of 0.9968. Vozka and Kilaz13 also reviewed correlations based on fuel properties. These correlations are predominantly ASTM methods, which are based either on the other fuel properties or on chemical composition. For example, in the ASTM D7215 (ref. 17) method, flash point is calculated as a function of simulated distillation data, which, in turn, is determined from chemical composition obtained from GC or GC \times GC. The ASTM D3338 (ref. 18) method for estimating net heat of combustion of aviation fuels uses density and average distillation data as inputs.

Wang *et al.*¹⁹ published a comprehensive review of the effects of fuel composition and hydrocarbon molecular structure on the fuel physicochemical properties, including density, NHC, low-temperature fluidity (viscosity and freezing point), flash point, and thermal-oxidative stability. They reviewed many correlations and developed correlations for estimating density, NHC, and viscosity using hydrogen/carbon ratios and molecular weights.

Flora *et al.*²⁰ performed a computational study of optimum blending ratios for conventional and alternative fuels that may be certifiable. The work described blending rules for calculating properties of binary fuel blends from different fractions of the two parent fuels. The properties included density, viscosity, flash point, aromatic concentration, NHC, and derived cetane number (DCN).

Since the $GC \times GC$ composition-based prediction methods of Shi et al.,15 as reviewed by Vozka and Kilaz13 provided good prediction results for many specification properties of interest for SAF, they are further described here. Shi et al.15 developed four statistical algorithms for the prediction of density at 20 °C, freezing point, flash point, and NHC of aviation fuels using detailed hydrocarbon composition obtained from the GC × GC-MS/FID measurements of 17 petroleum-based and synthetic aviation fuels. The four methods include the weighted average (WA) method, partial least squares analysis (PLS), genetic algorithm (GA), and modified weighted average (MWA) method. The correlations were based on the total content (wt%) of ten hydrocarbon classes for each carbon number, which could range from C7 to C19. A short description of all four Shi et al. methods can be found in the ESI.†

Further discussion of property predictions in this work

Predictive correlations that make use of chemical composition are of primary interest in this work since detailed composition determined using GC \times GC with a mass spectrometer and a flame ionization detector (GC \times GC-MS/FID) are available for two potential SAF-compatible fuels. The following sub-sections discuss prediction of the properties including liquid density, flash point, NHC, freezing point, distillation temperatures for 10% distilled, and FBP. They are the focus of this article because of the availability of experimental data.

Liquid density

According to Vozka *et al.*,¹⁶ the density of SAF is an important property and a key indicator of fuel quality. Density also determines the fuel load for a certain volume of fuel and hence the range of the aircraft. It is necessary for key operational requirements such as flow calculations, fuel gauging, metering device adjustments, and fuel thermal expansion calculations.²¹ Chemical composition impacts density;¹⁵ the latter increases with carbon number within a class of hydrocarbons. For a certain carbon number, the density decreases in the order aromatics > cycloparaffins > alkanes. Carbon number and hydrocarbon class are thus important indicators of density.

As discussed in the preceding section, Shi *et al.*¹⁵ developed four statistical algorithms to estimate density at 20 °C using data for 17 petroleum-based and synthetic aviation fuels. They reported that the PLS method gave the best results with an R^2 of 0.9999 and a mean absolute percentage error (MAPE) of 0.06%.¹⁵ As pointed out by Vozka *et al.*,¹³ the limitation of Shi's work was the small number of the fuels tested. In addition, the density correlations were for the temperature of 20 °C instead of 15 °C, which is the temperature required by aviation fuel standards.

To overcome the temperature problem of Shi's density correlation, Vozka et al.16 developed correlations for density at 15 °C predicted from chemical composition. Composition data obtained from GC \times GC-time-of-flight (TOF)-MS and GC \times GC-FID were used to construct a composition matrix with seven hydrocarbon classes;16,22 each of those classes included carbon numbers ranging from C7 to C20. They used 50 samples that included both alternate fuel blending components and petroleum-derived fuels. In addition, they used the density of a single compound to represent each carbon number for their calculations instead of the average density used in the method of Shi et al.15 In their work, they used the PLS method and the regularized support vector machines (SVM) method, which uses high dimensional spaces. For the PLS and SVM methods, two alternative correlations were developed, one using the composition matrix and another one using the product matrix. Therefore, a total of four correlations were presented: PLS composition, PLS product, SVM composition, and SVM product. The SVM composition method gave an R^2 of 0.9967 and the lowest MAPE of 0.1%. A short description of the Vozka et al. density method can be found in the ESI.†

Liquid density $\rho_{\rm L}$ of mixtures can be approximated by using specific volume $\nu_{\rm L}$ with a linear mixing rule; excess volume contribution is neglected.²³

$$\nu_{\rm L} = \sum_{i} w_i \nu_{{\rm L},i} \Rightarrow \rho_{\rm L} = \left(\sum_{i} \frac{w_i}{\rho_{{\rm L},i}}\right)^{-1} \tag{1}$$

where w_i is mass fraction of component *i*. According to Gmehling,²³ this simplified density calculation typically leads to small errors. Reiter *et al.*²⁴ used the equation towards the generation and evaluation of surrogate diesel fuels. Dahmen and Marquardt²⁵ used it in their work on formulation of biofuel blends.

Flash point

The flash point is the lowest temperature at which the vapors above a flammable liquid will ignite upon the application of an ignition source, *i.e.*, spark or flame.²⁶ At the flash point temperature, just enough liquid has vaporized to bring the vapor-air space over the liquid above the lower flammability limit.²⁷ Therefore, it is an important property for hydrocarbon fuels used in aviation and is used to determine their volatilization and flammability characteristics.¹⁵

The minimum flash point is generally 38 °C for kerosenetype fuels, with higher minimums allowed with specific agreements with the purchaser.

The low-end boiling point temperatures of the distillation curve dictate the flash point.¹² In the API technical databook,²⁸ flash point ($T_{\rm F}$) is correlated with the ASTM D86 (ref. 29) 10% temperature (t_{10}) for petroleum fractions:

$$T_{\rm F}(^{\circ}{\rm C}) = \frac{0.555556}{-0.013449 + \frac{1.583039}{[t_{10}(^{\circ}{\rm C}) + 273.15]} + 1.903 \times 10^{-3} \ln[t_{10}(^{\circ}{\rm C}) + 273.15]} - 273.15.$$
(2)

A simpler correlation was later proposed:²⁶

$$T_{\rm F}\,(^{\circ}{\rm C}) = -64.542 + 0.70704t_{10}\,(^{\circ}{\rm C}). \tag{3}$$

The ASTM D7215 method¹⁷ predicts flash point equivalent to the test methods ASTM D56 (ref. 30) and D93 (ref. 31) from simulated distillation data obtained using ASTM D2887.³² The equation for the D56 test method was developed using over one hundred samples of petroleum-derived diesel and jet fuel using PLS regression and is given as:

$$T_{\rm F} (^{\circ}{\rm C}) = -55.5 + 0.164 T_{\rm IBP} (^{\circ}{\rm C}) + 0.095 T_5 (^{\circ}{\rm C}) + 0.453 T_{10} (^{\circ}{\rm C})$$
(4)

where T_{IBP} is the initial boiling point temperature, and T_5 and T_{10} are the temperatures at which 5 and 10 wt% of the sample are recovered, respectively. T_{IBP} , T_5 and T_{10} are data from the simulated distillation measurement. Eqn (4) was used by Vozka *et al.*³³ to predict flash point of blends of hydroprocessed esters and fatty acids (HEFA) feedstocks sourced from camelina (CAME), tallow (TALL), and mixed fat (MFAT) with Jet A. Since the ASTM D7215 equation was not developed using mixtures of Jet A/A-1 with alternative fuel blending components such as HEFA, a new correlation was developed to improve the results using flash point data of the blends.³³ The revised parameters for the ASTM D56 equation are:

$$T_{\rm F} (^{\circ}{\rm C}) = -39.244 + 0.246 T_{\rm IBP} (^{\circ}{\rm C}) - 0.058 T_5 (^{\circ}{\rm C}) + 0.428 T_{10} (^{\circ}{\rm C}) (5)$$

Net heat of combustion (NHC)

NHC is a combustion performance specification for aviation fuels representing the energy contained in the fuel. It is the released energy obtained from the complete combustion of fuel, with products being in the gaseous state.²⁶ NHC can be reported on gravimetric or volumetric basis, and it determines the distance and payload of a flight.¹⁹ Civil aircrafts, usually constrained by payload weight, use fuels with high gravimetric NHC. Flight vehicles such as rockets and missiles are limited by the small design sizes of their fuel tanks to leave sufficient space for avionics; in such cases a higher volumetric NHC is desirable.^{38,39}

For aviation and SAF applications, gravimetric NHC is of interest. According to ASTM D3338,¹⁸ NHC of jet fuels should range from 40.19 to 44.73 MJ kg⁻¹. For convenience and brevity, the remainder of the paper shall refer to gravimetric NHC simply as NHC.

NHC can be calculated from chemical composition, such as n-paraffins, iso-paraffins, cycloparaffins, aromatic content, or the other properties, such as density and distillation temperatures. A list of these methods was given by Wang *et al.*¹⁹ Since NHC of aviation fuels are strongly influenced by the hydrogen content and the total number of carbon atoms, they correlated NHC of 64 fuels to hydrogen-to-carbon ratio (H/C) and molar mass (M).

The ASTM D3338 method utilizes distillation data, aromatic content (vol%), and density. It is the officially permitted calculation method and is listed in many jet fuel specifications worldwide:³³

$$NHC(MJ kg^{-1}) = \frac{[5528.73 - 92.6499A + 10.1601T + 0.314169AT]}{\rho_{15}} + 0.0791707A - 0.00944893T - 0.000292178AT$$

$$+ 35.9936 \qquad (6)$$

Additional alternative blending mixes (Fischer–Tropsch hydroprocessed synthesized paraffinic kerosene (FT-SPK), HEFA, synthesized iso-paraffins from hydroprocessed fermented sugars (SIP), and alcohol-to-jet synthetic paraffinic kerosene (ATJ)) with Jet A were later used to further validate the equation.³⁴

Flash point can also be predicted or calculated from detailed chemical composition.^{15,33,35–37} In the work of Shi *et al.*,¹⁵ flash point was correlated using the same four methods described in the density section above and the ESI.[†] Of the 17 aviation fuels tested, the MWA method was reported to provide the best results with an R^2 of 0.9968 and MAPE of 1.24%.¹⁵

where NHC is net heat of combustion (MJ kg⁻¹), sulfur-free basis, *A* is aromatics vol%, ρ_{15} is density (kg m⁻³) at 15 °C, and *T* (°C) is volatility. The latter is the average of 10%, 50%, and 90% points determined according to the D86 method.²⁹ Eqn (6) was derived from 241 fuels, most of which conforming with aviation gasoline and jet fuel specifications; stepwise linear regression methods were used for the derivation.

Vozka *et al.*^{33,34} applied the D3338 method to alternative aviation fuels and reported that it was well suited for HEFA/Jet A mixture and other alternate blending component mixtures. Results were compared with data using ASTM D4809;⁴⁰ the differences were below the reproducibility and the repeatability of the experimental values.

NHC can also be calculated from the detailed chemical composition^{15,37} of a fuel, such as composition obtained from GC \times GC data. In the simplest form, NHC is related to composition *via* the following equation.

$$NHC_{m} = \sum_{i} w_{i} NHC_{i}$$
(7)

where w_i is mass fraction of component *i*, NHC_m is the NHC of the mixture and NHC_{*i*} is NHC of the constituent component *i*. Vozka *et al.*³³ used this equation to predict NHC of blends of HEFA-CAME, -TALL, and -MFAT with Jet A using NHC values of Jet A and the neat HEFA blend components. Flora *et al.*²⁰ used it for blends of Jet A/FT-SPK (Syntroleum GTL), Jet A/HEFA-SPK (Sasol Iso-Paraffinic Kerosene (IPK)), Jet A/Gevo ATJ, Jet A/ LanzaTech ETJ, and Jet A/Farnesane. Here, ETJ and GTL stand for the ethanol-to-jet and gas-to-liquid fuels, respectively.

Shi *et al.*¹⁵ correlated NHC with detailed composition of 17 aviation fuels obtained from GC \times GC-MS/FID measurements by four statistical algorithms as described in the density section and the ESI.[†] The MWA method was reported to give the best results with R^2 of 0.9993 and mean absolute error (MAE) of 0.0102 MJ kg⁻¹.

Freezing point

Freezing point is a vital low-temperature fluidity characteristic for aviation fuels, especially in low temperature environments such as at high altitudes. Freezing point is the temperature at which the last wax crystal melts, when warming a fuel that has previously been cooled until wax crystals form.^{41,49} The freezing points of Jet A and Jet A-1 are specified to be -40 °C and -47 °C, respectively (Table 1). Typically, the requirement is to maintain the fuel with a temperature of at least 3 °C above its freezing point.^{20,42}

Freezing point is influenced by molecular structure and symmetry. The freezing point of hydrocarbons is affected by carbon number^{33,43} and hydrocarbon class.^{14,33,44} Yang et al.¹⁴ listed three key factors that significantly impact the freezing point of bio-jet fuel; they are iso-paraffins content, alkylated aromatics content, and the carbon chain length of bioparaffins. n-Paraffins have the highest freezing point within a carbon number. Increasing iso-paraffin content in aviation fuels was found to lower freezing point.14,43,44 A definite trend was not found for cycloparaffins,43 however, generally low freezing points were observed for alkyl-substituted cycloparaffins.45,46 Certain aromatic compounds can reduce the freezing point of bio-jet fuels⁴⁷ and blends of jet fuels with a gas-to-liquid-derived synthetic paraffinic kerosene (GTL SPK).48 Average carbon number of a fuel also influences its freezing point; fuels with lower average carbon number have lower freezing points.19

Cookson *et al.* correlated freezing point with chemical composition detected by high-performance liquid chromatog-raphy (HPLC), GC, and ¹³C NMR.^{49–52} One equation related freezing point to the total amount of n-paraffins:⁴⁹

$$FP(^{\circ}C) = 60.7w(n) - 62.0$$
(8)

FP (°C) =
$$85.5w$$
 (C12 to C14) - 60.3 (9)

The third equation used the total amount of n-paraffins, branched and cyclic paraffins and aromatics.^{49,52}

FP (°C) =
$$-0.8w(n) - 63.8w(BC) - 55.9w(Ar)$$
 (10)

The freezing point data of kerosene fuels used to develop eqn (8)–(10) ranged from -50 °C to -32 °C.

Distillation temperatures were later introduced.⁵¹ For the equation below, the freezing point data of jet fuels used ranged from -70 °C to -33.5 °C.

FP (°C) =
$$81.1w(n) + 53.6w(Ar)$$

+ $0.255T_{10}$ (°C) + $0.338T_{90}$ (°C) - 206.2 (11)

In eqn (8)–(11), w(n), w(BC), and w(Ar) are weight fractions of n-paraffins, branched + cyclic paraffins, and aromatics, respectively. w(C12-C14) is weight fraction of the C12 to C14 n-paraffins. T_{10} and T_{90} were obtained from GC or simulated distillation method.⁵¹

Vozka *et al.*³³ evaluated the four Cookson equations for predicting the freezing point of blends of Jet A and three HEFAs: CAME, TALL, and MFAT. They reported that since the freezing point values used in the development of the Cookson equations were in the range of -50 °C to -32 °C and the fuels used originated from different sources, these equations could not accurately predict the freezing points of the HEFA blends. The experimental freezing points of Jet A/HEFA blends range from -59 °C to -48.5 °C; with the lowest value lying outside the correlation limit of -50 °C. They applied the Cookson equations to predict freezing point of their mixtures anyway and concluded that none of the equations produced accurate results for TALL/Jet A blends, while eqn (8) predicted the freezing point within 2 °C of experiments for CAME/Jet A and most of MFAT/Jet A blends.

Shi *et al.*¹⁵ also correlated freezing point with detailed composition of 17 aviation fuels obtained from GC × GC-MS/FID measurements by different statistical algorithms as discussed previously. The freezing point data used ranged from -67 °C to -37.5 °C. The MWA method was reported to give the best results with R^2 of 0.9866 and MAE of 0.82 °C.

Wang *et al.*¹⁹ and Vozka *et al.*¹³ listed many other methods used to estimate freezing point of aviation and SAF fuels. They are not considered in this work.

Distillation temperature

According to ASTM D7566,⁵³ distillation temperatures of aviation fuels refer to the D86 distillation temperatures, which can be obtained using the ASTM D86 (ref. 29) or ASTM D2887 (ref. 32) procedures. For the D2887 procedure, the simulated distillation data are first obtained, then converted to the D86 temperatures using the correlation given in Appendix X4 of the ASTM D2887 standard.³² For this correlation, the test method D86 correlated data is calculated from the test method D2887 data using the following equation,

$$a_{n} = a_{0} + a_{1} \cdot T_{n-1} + a_{2} \cdot T_{n} + a_{3} \cdot T_{n+1}$$
(12)

where t_n is the *n*th boiling point temperature of test method D86 correlated, a_i is the *i*th coefficient given in Table X4.1 of the ASTM D2887 standard, and T_n is the *n*th boiling point temperature of D2887.

The distillation temperatures of an aviation fuel can be predicted by calculating the equivalent D2887 temperatures using the compositions (weight percent) and normal boiling point temperatures of the constituent compounds in the fuel. A spline fit of the sorted normal boiling point temperatures *versus* the cumulative weight percent of the components is used to obtain the equivalent D2887 temperatures. Eqn (12) is then used to calculate the distillation temperature from the D2887 temperatures.

Experimental measurements for jet fuel cuts of hydrotreated CFP oils

CFP-derived liquid hydrocarbons were produced *via* pilot-scale or bench-scale CFP to generate CFP oil, followed by hydrotreating (HT) of the oil. The steps mirror the conceptual process described above (Fig. 1). The hydrotreated product was further distilled to obtain the jet fuel boiling range cuts used for compositional analysis and property measurements. Those results were used in this work to evaluate how well the composition-based, and ASTM prediction methods discussed in the preceding section perform compared to the experimental data.

The jet range fuels used in this work were produced from hydrotreating of two different CFP intermediates that had been produced using different CFP catalysts, ZSM-5 and Pt/TiO2. The first CFP intermediate was produced in the thermochemical pilot-scale process development unit at NREL54 with a feed rate of 10 kg h^{-1} of pine in an *ex situ* configuration at 550 °C, over a ZSM-5 catalyst from Equilibrium Catalyst, Inc.,55 and the second CFP intermediate was produced in a bench-scale ex situ fluidized-bed pyrolyzer and fixed bed upgrading reactor combination with a feed rate of 150 g h^{-1} of a mixture of pine and forest residues over a 0.5 wt% Pt/TiO2 catalyst.9 Both CFP oils were hydroprocessed in a continuous trickle bed reactor in a staged process with a maximum temperature of 385 °C and a pressure of 125 bar over a sulfided NiMo/Al2O3 catalyst without plugging issues.8 The hydrotreated product was further fractionated via distillation in a micro spinning band distillation apparatus, and the cut obtained at 145-245 °C was evaluated for jet fuel properties. For convenience, the two resultant jet fuels will be referred to as "ZSM-5 CFP, HT fuel" and "Pt/TiO2 CFP, HT fuel", respectively, for the remainder of the paper.

Compositions of these jet range fuels were obtained using GC \times GC-MS/FID method on a LECO Pegasus 4D system.⁵⁶ Components were identified based on mass spectral matching using LECO ChromaTOF® software⁵⁶ and the NIST 2014 Mass Spectral Library,⁵⁷ characteristic spectral features, and retention times. The GC \times GC/FID results were quantified based on calculated response factors (using effective carbon number) and

Table 2 Mass percentage of compound groups found in the jet fuel cuts from the distillation of the hydrotreated products from CFP oils obtained using two different catalysts (ZSM-5 and Pt/TiO₂)

	ZSM-5 CFP,	Pt/TiO ₂ CFP,
Class	HT fuel	HT fuel
n-Paraffins	1.84	2.57
Iso-paraffins	2.74	2.13
One-branched monocycloparaffins	16.74	31.33
Multi-branched monocycloparaffins	36.32	30.51
Di- and tri-cycloparaffins	37.51	29.63
Alkylbenzenes	2.02	2.26
Cycloaromatics	2.82	1.57

normalized to 100%; note that the boiling range of the jet fuel cut is in a temperature range which fully evaporates in the GC inlet and elutes through the column for detection. Details of the analysis are included in the ESI.[†] Table 2 shows the normalized mass percentage of compounds found by class. Both fuels were dominated by cycloparaffins. The Pt/TiO2 CFP, HT fuel contained more than twice the amount of mono-branched monocycloparaffins and about 5 wt% less of the other types of cycloparaffins compared to the ZSM-5 CFP, HT fuel. For other compound classes, the overall compositions of the two fuels were similar. Detailed compositions of compounds identified for ZSM-5 CFP, HT fuel and Pt/TiO2 CFP, HT fuel can be found in Tables S2 and S3, respectively in the ESI.[†] The heaviest compound identified in ZSM-5 CFP, HT fuel was C13H22 ((3aa, $6a\alpha$, $9a\alpha$, $9b\beta$)-perhydrophenalene), while that for Pt/TiO₂ CFP, HT fuel was C₁₄H₂₆ (1,1'-(1,2-ethanediyl)biscyclohexane). No C7 compounds were identified in the Pt/TiO₂ CFP, HT fuel.

Measured properties for the fuels included liquid density at 15 °C, freezing point, NHC, D86 distillation temperature at 10%, D86 distillation FBP, and flash point. The measurements are summarized in Table 3. The measured properties are well within the specification limits of aviation fuels shown in Table 1.¹²

Pure properties and methods used for mixture property predictions

We collected pure component properties of the compounds (Tables S2 and S3[†]) identified in the ZSM-5 CFP, HT fuel and Pt/ TiO₂ CFP, HT fuel because pure component properties inform mixture property predictions in most of the methods. The purecomponent properties collected included liquid density at 15 ° C, freezing point, NHC, flash point, and normal boiling point; these were necessary for the mixture properties reported in this paper.

Thermodynamic properties of pure compounds were found with the NIST ThermoData Engine (TDE)⁵⁹ software. In this software, the dynamic data evaluation concept is implemented to obtain thermodynamic and thermophysical properties of compounds and mixtures. The available data are used as input parameters for single-property or multi-property equations. If the experimental property values are not available, the gaps are Table 3 Measured properties of the jet fuel cuts from the distillation of the hydrotreated products from ZSM-5 and Pt/TiO₂ CFP oils

	Density ^a at 15 °C (kg m ⁻³)	Freezing point ^{<i>a</i>} (°C)	NHC ^b (MJ kg ⁻¹)	Distillation ^c , 10% (°C)	Distillation ^c , FBP (°C)	Flash point ^a (°C)
ASTM D7566 ⁵³ / D4054 ⁵⁸	775 to 840	Max -40	>42.8	≤205	≤300	>38
ZSM-5 CFP, HT fuel	834	<-70	43	174.3	249.2	49.6
Pt/TiO ₂ CFP, HT fuel	833	<-70	43	170.2	256.8	46.6
Reproducibility ^d	0.5	0.8	0.234	2.4	11.8	3.1

^{*a*} Measured at an ambient pressure of 0.08 MPa. ^{*b*} Refers to t = 25 °C and p = 101.3 kPa. ^{*c*} Estimated D86 temperature obtained from D2887 data using eqn (12). ^{*d*} Expanded uncertainties for 0.95 level of confidence. Estimated as reproducibility values taken from the standards. For distillation data, the values are for D2887 data.

filled with the predicted ones. Details on the TDE procedures for pure compounds have been published.⁶⁰⁻⁶²

The pure-compound freezing points were found as the crystal-liquid triple-point temperatures. This quantity was evaluated by TDE, and the values were reported only if experimental results were available. To fill gaps in the experimental data, saturated liquid densities were predicted by the Yamada-Gunn method using the modified Rackett equation.⁶³

NHC of compound $C_aH_bO_c$ is the enthalpy (isobaric heat of reaction) of its quantitative combustion in oxygen where the products are gaseous H₂O and CO₂. The specific net heats of combustion at T = 25 °C and P = 0.1 MPa were calculated with the equation:

$$NHC = \frac{a\Delta_{f}H(CO_{2}(g)) + 0.5b\Delta_{f}H(H_{2}O(g)) - \Delta_{f}H(C_{a}H_{b}O_{c})}{M}$$
(13)

where $\Delta_f H(CO_2(g)) = -393.51 \text{ kJ mol}^{-1}$ and $\Delta_f H(H_2O(g)) =$ -241.826 kJ mol⁻¹ were recommended by CODATA⁶⁴ and *M* is the molar mass of the target compound. The enthalpy of formation of the compound, $\Delta_f H(C_a H_b O_c)$, was found with one of three procedures described below. If experimental data were available, the $\Delta_{f}H$ value in the liquid phase evaluated by TDE was used. Otherwise, the gas-phase value was estimated using the group-contribution procedure by Verevkin et al.65 The enthalpies of vaporization evaluated by TDE were subtracted from the estimated value to obtain the liquid-phase values. For some compounds, the experimental data appeared to be unreliable and/or the above predictive procedure could not be used. In this case, the gas-phase enthalpy of formation was predicted using the aLL5 variation of the ab initio-based protocol by Paulechka and Kazakov⁶⁶ and converted to the liquid phase value as described above.

The normal boiling point temperatures were derived from the temperature-dependent vapor pressures interpolated by the Wagner 25 equation.⁶⁷

$$\ln\left(\frac{p}{p_{\rm c}}\right) = \frac{T_{\rm c} + 273.15}{T + 273.15} \left(A_1\tau + A_2\tau^{1.5} + A_3\tau^{2.5} + A_4\tau^5\right)$$
(14)

where $\tau = 1 - \frac{T + 273.15}{T_c + 273.15}$, *T* and *T*_c are the temperature and critical temperature in °C, respectively, and *p*_c is the critical pressure. *T*_c values were found using the experimental data or

predictive procedures.^{68–72} The critical pressure and vapor pressures were obtained from simultaneous regression of multiple properties (p, p_c , enthalpy of vaporization, liquid and gas heat capacities, *etc.*). The source of the input p_c values was either experimental or one of the predictive methods.^{68–71,73} The predicted p_c values given in Table S4† are the ones used as the inputs. The evaluated p_c values are close, but not necessarily equal, to them.

For the compounds considered here, the experimental vapor pressures were not measured over the whole temperature range from the triple point to the critical point or were not available at all. To fill the gaps, predictions with the Ambrose–Walton⁷³ method were used. An acentric factor is required in this method. In some cases, it was found from the critical parameters and the normal boiling point temperature predicted with the Nannoolal–Rarey–Ramjugernath–Cordes⁷⁴ or Constantinou–Gani⁷² methodologies.

Flash points were calculated by a two-step procedure. The flash-point pressure of a liquid $C_a H_b O_c$ was estimated using the equation:⁷⁵

$$P_{\rm fp}(\rm kPa) = \frac{101.3}{8N}$$
(15)

where N = a + (b - 2c)/4 is the number of moles of O₂ required for complete combustion of 1 mol of the compound. Then, the flash-point temperature was found using eqn (14).

Selected properties of the compounds are presented in Table S4 in the Excel file of the ESI.†

Results and discussion

Properties of ZSM-5 CFP, HT fuel and Pt/TiO_2 CFP, HT fuel were predicted using composition-based correlations as well as the ASTM methods discussed above and compared to experimental measurements shown in Table 3. Results are discussed in this section.

Liquid density

Liquid density at 15 °C was predicted using the statistical methods of Shi *et al.*¹⁵ including WA, PLS, GA, and MWA methods; the methods of Vozka *et al.*¹⁶ with 98 predictors including PLS product, PLS composition, SVM product, and

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SVM composition methods; and eqn (1). Although the methods of Shi et al. were developed using data at 20 °C, we included them for comparison. In applying these methods for density prediction (as well as flash point, freezing point, and NHC), the composition matrices of the CFP-derived SAF fuels were constructed using the weight % data in Tables S2 and S3[†] and by mapping each compound in the table to one of the 10 classes defined in the methods. The resulting composition matrices are summarized in Tables S5 and S6.† The density matrix from the Shi et al. paper¹⁵ with average density values for each carbon number in the class was used instead of the density value for each individual component of the fuel. For the methods of Vozka et al.,16 the composition and density matrices were constructed in the same way as the Shi's methods with compounds mapped to one of the seven classes defined in the methods, but with a representative density value for each carbon number in the class, as recommended by the authors. The resulting composition matrices for the fuels are shown in Tables S7 and S8 of the ESI.[†] When applying eqn (1), the density value for each individual component shown in Table S4 of the ESI[†] was used. The prediction results using these methods shown in table are compared with the experimental data from table. For each method, error, mean absolute error (MAE), percent error, and mean absolute percent error (MAPE) are reported (Table 4).

The MWA, PLS_P, and eqn (1) methods predicted liquid density at 15 °C with MAPE of ~0.5%, with the MWA method having the lowest MAPE of 0.39%. The GA method predicted negative density for both fuels. The negative density prediction likely indicates that the method was developed using too few fuel samples (17), given the large number of model coefficients involved (20) compared to the PLS or MWA methods, which used 10 coefficients each. All four methods of Shi *et al.* underpredicted density at 15 °C of both fuels, except MWA, which overpredicted density of Pt/TiO₂ CFP, HT fuel. Since these methods were developed using density data at 20 °C, the underprediction of data at the lower temperature of 15° is reasonable.

While the SVM_C method had been shown to provide the most accurate results (MAPE of 0.0984%) for 50 samples of petroleum-derived fuels and alternative fuel blending components,16 it underpredicted liquid density of the two CFP-derived SAF fuels, resulting in MAPE of 5.31%. It is likely that the large discrepancy in accuracy is caused by the especially high contents of cycloparaffins present in the two fuels (>90 wt%) compared to those used in the development of the SVM_C method. The PLS product and PLS composition methods performed better than the SVM product and SVM composition methods for these fuels with MAPE of 0.51% and 0.84%, respectively. The simple weighted average method of eqn (1) also provided good predictions with MAPE of 0.52%. Similar results of \sim 0.5% by the three best methods (MWA, PLS_P, and eqn (1)) can be considered quite good. However, they are not good enough since the MAEs of ~ 0.004 g cm⁻³ are larger than the reproducibility of the measured density data of 0.0005 g cm^{-3} .

Flash point

Flash points of the CFP-derived SAF fuels were predicted using the methods of Shi *et al.*,¹⁵ the API correlation (eqn (2),²⁸), the ASTM D7215 method (eqn (4),¹⁷), and the modified ASTM method (eqn (5),³³). For the Shi *et al.* methods, the same composition matrices used in the density predictions were used. The flash point property matrix from the paper¹⁵ with average flash point value for each carbon number in the class was used. The revised coefficients for the MWA method as shown in Table S10 of the ESI were used since the values presented in table 9 of the original paper¹⁵ for flash point were incorrect because they were swapped with those for freezing point method.⁷⁶

For eqn (2), the ASTM D86 10% temperature, t_{10} , was calculated using the method described in the distillation temperature section above, and from composition of the

Table 4 Comparison between experimental (exp.) density data (g cm $^{-3}$) at 15 °C and predictions for the two jet range fuels produced from hydrotreated ZSM-5 and Pt/TiO₂ CFP oils

Fuel name E		Prediction								
	Exp.	WA ^{<i>a,b</i>}	PLS ^{<i>a,b</i>}	$\mathrm{GA}^{a,b}$	MWA ^{<i>a,b</i>}	PLS_P	PLS_C	SVM_P	SVM_C	Eqn (1)
ZSM-5 CFP, HT fuel	0.8345	0.8208	0.8187	-0.5380	0.8336	0.8415	0.8439	0.7784	0.7666	0.8363
Pt/TiO ₂ CFP, HT fuel	0.8332	0.8138	0.8099	-1.5786	0.8389	0.8348	0.8377	0.8123	0.8126	0.8263
		Error								
ZSM-5 CFP, HT fuel		-0.0137	-0.0158	-1.3725	-0.0009	0.0070	0.0094	-0.0561	-0.0679	0.0018
Pt/TiO ₂ CFP, HT fuel		-0.0194	-0.0233	-2.4118	0.0057	0.0016	0.0045	-0.0209	-0.0206	-0.0069
	MAE ^c	0.0165	0.0195	1.8921	0.0033	0.0043	0.0070	0.0385	0.0443	0.0044
		% Error								
ZSM-5 CFP, HT fuel		-1.64	-1.89	-164.47	-0.1	0.83	1.1	-6.72	-8.14	0.21
Pt/TiO ₂ CFP, HT fuel		-2.33	-2.80	-289.46	0.68	0.20	0.54	-2.51	-2.47	-0.83
	$MAPE^{d}$	1.98	2.34	226.96	0.39	0.51	0.84	4.62	5.31	0.52

^{*a*} Estimation methods of Shi *et al.*¹⁵ WA = weighted average, PLS = partial least squares, GA = genetic algorithm, MWA = modified weighted average. Estimation methods of Vozka *et al.*¹⁶ PLS_P = partial least squares product, PLS_C = partial least squares composition, SVM_P = support vector machine product, SVM_C = support vector machine composition. ^{*b*} Estimation methods of Shi *et al.*¹⁵ are for density at 20 °C but are included for comparison. ^{*c*} MAE = mean absolute error. ^{*d*} MAPE = mean absolute percent error.

constituent compounds of the fuels shown in Tables S2 and S3[†] and normal boiling point temperatures in Table S4.[†] The distillation temperatures T_{IBP} , T_5 , and T_{10} required in the ASTM D7215 method, eqn (4) and its modification, eqn (5) were calculated from composition shown in Tables S2 and S3[†] and normal boiling point temperatures in Table S4.[†] A spline fit of the sorted normal boiling point temperatures *versus* the cumulative weight percent of the components is used to obtain the required temperatures. Prediction results using these methods are compared to the experimental data in Table 5.

Four prediction methods, MWA, API, ASTM 1, and ASTM 2, gave MAE within the reproducibility of the measured flash point data of 3.1 °C. The API method (eqn (2)) gave the lowest MAE and MAPE among the methods tested. It underpredicted flash point of the ZSM-5 CFP, HT fuel by 3.4 °C, but predicted the property of the Pt/TiO₂ CFP, HT fuel exactly, resulting in an MAE of 1.7 °C and an MAPE of 3.4%. The ASTM D7215 method (ASTM 1) gave a slightly larger MAE of 1.8 °C and MAPE of 3.8%, with a smaller error for ZSM-5 CFP, HT fuel, and a larger error for the Pt/TiO₂ CFP, HT fuel. The ASTM 2 method provided the next best results with MAE and MAPE of 2.0 °C, and 4.2%, respectively. The recommended MWA method of Shi et al. (with a reported MAPE of 1.24%)15 had the lowest MAE and MAPE among the four methods of Shi et al. However, it overpredicted flash point of Pt/ TiO₂ CFP, HT fuel compared to the API, ASTM 1, and ASTM 2 methods, hence larger MAE (2.9 °C) and MAPE (6.1%). The significantly larger errors of the MWA method for the two CFPderived SAF fuels compared to literature¹⁵ may be caused by the significantly higher contents of cycloparaffins present in these fuels (>90 wt%) compared to the values of \sim 50 wt% for most of the fuels used in the development of the method.¹⁵ The WA method provided the next best results among the Shi et al. methods. The PLS method predicted flash points that are significantly lower than the minimum of 38 °C allowed for aviation fuels (see Table 1). The GA method significantly overpredicted flash point of ZSM-5 CFP, HT fuel, but underpredicted

Pt/TiO₂ CFP, HT fuel. The erratic behavior may be attributed to overfitting of the data as previously discussed for density.

Net heat of combustion (NHC)

NHC of the two CFP-derived SAF fuels was predicted using the methods of Shi et al.,15 the ASTM D3338 method, eqn (6),18 and the simple weight fraction average of pure component NHC, eqn (7). For the Shi et al. methods, the same composition matrices used in the density and flash point predictions were used. The NHC property matrix from the paper¹⁵ with average property value for each carbon number in the class was also used. For eqn (6), the aromatics volume%, A, was computed from the composition data of aromatics compounds shown in Tables S2 and S3 of the ESI.[†] Liquid density at 15 °C, ρ_{15} , was calculated using the Shi's MWA method, which provided the best prediction of liquid density at 15 °C for these two fuels. The volatility, T, was calculated from the average of test method D86 10%, 50%, and 90% points. When applying eqn (7), the NHC value for each individual component shown in Table S4 of the ESI[†] was used. The prediction results using these methods are compared against the experimental data in Table 6.

The WA method of Shi *et al.*, the ASTM D3338 correlation, and eqn (7) predicted NHC of the two CFP-derived SAF fuels with MAPE well below 1% and with MAE within the reproducibility of the measured data of 0.234 MJ kg⁻¹. The difference between the WA method and eqn (7) lies in the fact that the former uses an average value of NHC for each of the 13 carbon numbers in the 10 compound classes while for eqn (7), NHC for each individual compound was used. However, the predicted NHCs of the fuels themselves are essentially the same. MWA, the best method from Shi *et al.*,¹⁵ overpredicted the data and gave MAPE of 2.2%. The PLS method also overpredicted the data, resulting in an MAPE of 2.5%. The GA method underpredicted NHC with values that are well below the minimum value of 42.8 MJ kg⁻¹ required for aviation fuels (see Table 1).

Table 5 Comparison between experimental data (exp.) and predicted flash points (°C) for the two jet range fuels produced from hydrotreated ZSM-5 and Pt/TiO₂ CFP oils

Fuel name		Prediction	Prediction									
	Exp.	WA ^a	PLS ^a	GA^a	MWA ^a	API	ASTM 1	ASTM 2				
ZSM-5 CFP, HT fuel	49.6	47.1	28.8	172.7	52.2	46.2	47.4	47.9				
Pt/TiO ₂ CFP, HT fuel	46.6	42.1	-1.3	17.4	49.9	46.6	48.0	48.9				
		Error										
ZSM-5 CFP, HT fuel		-2.5	-20.8	123.1	2.6	-3.4	-2.2	-1.7				
Pt/TiO ₂ CFP, HT fuel		-4.5	-47.9	-29.2	3.3	0.0	1.4	2.3				
	MAE^{b}	3.5	34.3	76.2	2.9	1.7	1.8	2.0				
		% Error										
ZSM-5 CFP, HT fuel		-5.0	-41.9	248.2	5.1	-6.8	-4.4	-3.3				
Pt/TiO ₂ CFP, HT fuel		-9.6	-103	-62.7	7.1	0.0	3.1	5.0				
	$MAPE^{c}$	7.3	72.3	156	6.1	3.4	3.8	4.2				

^{*a*} Estimation methods of Shi *et al.*¹⁵ WA = weighted average, PLS = partial least squares, GA = genetic algorithm, MWA = modified weighted average. API = eqn (2);²⁸ ASTM 1 = ASTM D7215 - eqn (4);¹⁷ ASTM 2 = modified ASTM D7215 - eqn (5).^{33 *b*} MAE = mean absolute error. ^{*c*} MAPE = mean absolute percent error.

Table 6 Comparison between experimental data (exp.) and predicted NHC (MJ kg⁻¹) for the two jet range fuels produced from hydrotreated ZSM-5 and Pt/TiO₂ CFP oils

Fuel name		Prediction	Prediction							
	Exp.	WA ^a	PLS ^a	GA^a	MWA ^a	ASTM	Eqn (7)			
ZSM-5 CFP, HT fuel	42.99	43.01	43.63	37.21	43.53	42.95	43.03			
Pt/TiO ₂ CFP, HT fuel	43.02	43.11	44.53	34.48	44.38	43.03	43.13			
		Error								
ZSM-5 CFP, HT fuel		0.02	0.64	-5.78	0.53	-0.04	0.04			
Pt/TiO ₂ CFP, HT fuel		0.09	1.51	-8.54	1.35	0.01	0.11			
	MAE^{b}	0.05	1.08	7.16	0.94	0.02	0.07			
		% Error								
ZSM-5 CFP, HT fuel		0.04	1.5	-13.4	1.2	-0.09	0.1			
Pt/TiO ₂ CFP, HT fuel		0.2	3.51	-19.9	3.14	0.02	0.24			
·	$MAPE^{c}$	0.1	2.50	16.7	2.2	0.05	0.2			

^{*a*} Estimation methods of Shi *et al.*¹⁵ WA = weighted average, PLS = partial least squares, GA = genetic algorithm, MWA = modified weighted average. ASTM = ASTM D3338 - eqn (6).¹⁸ ^{*b*} MAE = mean absolute error. ^{*c*} MAPE = mean absolute percent error.

The ASTM D3338 correlation gave the lowest MAE (0.02 MJ $\rm kg^{-1})$ and MAPE (0.05%) for these fuels.

Freezing point

Freezing points of the two CFP-derived SAF fuels were predicted using the methods of Shi *et al.*¹⁵ and Cookson *et al.* (eqn (8)–(11)). For the Shi *et al.* methods, the composition matrices from the density predictions were used. The freezing point property matrix from the paper¹⁵ with average freezing point value for each carbon number in the class was also used. The revised coefficients for the MWA method as shown in Table S10 of the ESI† were required since the values presented in Table 9 of the original paper¹⁵ were swapped with those for the flash point method.⁷⁶ For the Cookson *et al.* correlations, eqn (8)–(11), the parameters w(n), w(BC), and w(Ar) were computed from the weight fractions of n-paraffins, branched + cyclic paraffins, and aromatics, respectively, of the CFP-derived SAF fuels (Table 2). w(C12 to C14) was determined from the sum of the weight fractions of all C12 to C14 n-paraffins. T_{10} and T_{90} were determined using the same procedure employed in computing the distillation temperatures for flash point prediction *via* the ASTM D7215 method (eqn (4)).

Prediction results using these methods are compared with the experimental data in Table 7. Since the experimental freezing points of both fuels were reported as <-70 °C, actual errors and % errors could not be determined. However, a value of -70 °C allowed the comparison of predicted results against this upper limit.

All the methods tested predicted meaningful freezing points for the two fuels, without any erratic results as observed for some of the density and flash point methods. In addition, no predicted values are above the specification limits of -40 °C for Jet A and -47 °C for Jet A-1. For the ZSM-5 CFP, HT fuel, the WA, eqn

Table 7 Comparison between experimental data (exp.) and predicted freezing points (°C) for the two jet range fuels produced from hydrotreated ZSM-5 and Pt/TiO₂ CFP oils

Fuel name		Prediction									
	Exp.	WA ^a	PLS ^a	GA ^a	MWA ^a	Eqn (8)	Eqn (9)	Eqn (10)	Eqn (11)		
ZSM-5 CFP, HT fuel	<-70	-64	-88	-75	-88	-61	-60	-62	-94		
Pt/TiO ₂ CFP, HT fuel	<-70	-71	-106	-69	-89	-60	-60	-62	-95		
		Error ^b									
ZSM-5 CFP, HT fuel		6	-18	-5	-18	9	10	8	-24		
Pt/TiO ₂ CFP, HT fuel		-1	-36	1	-19	10	10	8	-25		
	MAE ^c	4	27	3	19	9	10	8	24		
		% Error	b								
ZSM-5 CFP, HT fuel		9	-26	-7	-26	13	15	11	-34		
Pt/TiO ₂ CFP, HT fuel		-1	-51	2	-27	14	14	12	-36		
	$MAPE^{d}$	5	38	4	27	13	14	11	35		

^{*a*} Estimation methods of Shi *et al.*¹⁵ WA = weighted average, PLS = partial least squares, GA = genetic algorithm, MWA = modified weighted average. ^{*b*} Errors and errors% were computed assuming that the experimental freeze point was -70 °C. ^{*c*} MAE = mean absolute error. ^{*d*} MAPE = mean absolute percent error.

Table 8Comparison between experimental data (exp.) and predicteddistillation temperatures (°C) for the two jet range fuels produced fromhydrotreated ZSM-5 and Pt/TiO2 CFP oils

	t_{10}		$t_{ m FBP}$		
Fuel name	Exp.	ASTM ^a	Exp.	ASTM ^a	
ZSM-5 CFP, HT fuel	174.3	164.9	249.2	224.5	
Pt/TiO ₂ CFP, HT fuel	170.2	165.4	256.9	217.1	
		Error		Error	
ZSM-5 CFP, HT fuel		-9.4		-24.7	
Pt/TiO ₂ CFP, HT fuel		-4.8		-39.7	
	MAE^{b}	7.1		32.2	
		% Error		% Error	
ZSM-5 CFP, HT fuel		-5.4		-9.9	
Pt/TiO ₂ CFP, HT fuel		-2.8		-15.4	
	MAPE ^c	4.1		12.7	

 a ASTM D2887 conversion procedure described above. $^{^{32}\ b}$ MAE = mean absolute error. c MAPE = mean absolute percent error.

(8)–(10) predicted freezing points that are higher than the -70 °C limit of the data by more than 5 °C, with eqn (8) and (9) overpredicting by 9 and 10 °C, respectively. The remaining methods predicted freezing points that are lower than the upper limit of -70 °C. The GA method and eqn (8)–(10) overpredicted the freezing point data of the Pt/TiO₂ CFP, HT fuel. However, both the WA and GA methods predicted freezing points that are close to the upper value of -70 °C for this fuel. The remaining three methods (PLS, MWA, eqn (11)) predicted freezing points that are well below the upper limit of -70 °C. Since the actual value of freezing point was not reported, it is not possible to judge the relative merits of each prediction method. However, it is reasonable to state that the methods that predicted freezing points below and close to -70 °C are better than those that overpredicted it.

Distillation temperature

Distillation temperatures t_{10} and t_{FBP} of the two CFP-derived SAF fuels were calculated using the method described in the distillation temperature section above, and compositions and normal boiling points of the constituent compounds of the two fuels shown in Tables S2–S4 of the ESI.† Prediction results are summarized in Table 8. The procedure underpredicted t_{10} of the ZSM-5 CFP, HT fuel and Pt/TiO₂ CFP, HT fuel by 9.4 °C and 4.8 ° C, respectively. The resulting MAE of 7.1 °C is well above the reproducibility limit of 2.1 °C. For t_{FBP} , the method underpredicted the data by 24.7 °C and 39.7 °C, respectively, resulting in MAE of 32.2 °C, which is also well above the reproducibility limit of 11.8 °C. The significant underprediction is likely caused by the estimated normal boiling points of the compounds in the heavy ends of the fuels.

Conclusion

This work showed that flash point, and NHC of jet range fuels from two different CFP catalysts can be well predicted using many of the available prediction methods with mean absolute errors that are within the reproducibility limit of the property data. The API correlation (eqn (2)), and the ASTM D3338 correlation (eqn (6)) provided the lowest mean absolute error for flash point, and NHC, respectively. Density at 15 °C can be predicted with mean absolute percent error of $\sim 0.5\%$, which is quite good, but not accurate enough. For freezing point, all the methods tested predicted meaningful results for the two fuels. However, the accuracy of the prediction methods could not be assessed since the experimental data of the two fuels were given as $<-70^{\circ}$ C. To facilitate model evaluation, any future measurements of freezing point data for CFP-derived SAF fuels should report actual numerical values. The t_{10} and t_{FBP} distillation temperatures were not well predicted by the method tested, which significantly underpredicted the experimental data. The shortcomings are likely caused by the uncertainty of the estimated normal boiling points of the compounds in the fuels.

While results for density, flash point, and NHC are encouraging, the evaluation was based on only two sets of data. Additional data points would be useful in improving the confidence in the applicability of the methods. The scarcity of data was due to the efforts and resources necessary for producing the required samples, along with the associated analytical measurements, within the time window of the initial experimental work. There are ongoing efforts at NREL to produce additional CFP-derived SAF samples for compositional and property analysis. These data, when they become available, will be useful in further evaluation effort and to update of the knowledge gained in this work. Our future work will also report on the prediction of some additional properties shown in Table 1 to help advance the development of CFP-derived SAF.

Author contributions

SW: literature data collection, mixture property predictions modeling, technical coordination, writing – initial draft and subsequent review/edits. EP: data collection, estimation and compilation of pure properties, writing/review/edits. KI: experimental work and providing related data, writing/review/edits. EC: analytical measurements of CFP products and related review/edits. CM: NIST coordination and group lead, review/ edits. AD: conceptualization and coordination as NREL project lead for thermochemical analysis, writing/review/edits.

Conflicts of interest

The authors declare that they have no known competing financial interests or personal relationships than could have appeared to influence the work reported in this paper.

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