

PREDICTION OF THE FCC FEEDSTOCKS CRACKABILITY

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This paper presents a statistical model for prediction of feedstock's crackability (potential to generate valuable products on catalytic cracking process), based on experimental reactivity data by microactivity test (MAT - Microscale Fixed Bed Reactor) and detailed physicochemical characterization. A minimum amount of experimental tests corresponding to feed properties (typically available at refinery) is used to build a more complete description of feedstocks including chemical composition and hydrocarbon distribution. Both measured and calculated physicochemical properties are used to predict the yields of main products at several MAT reaction severities. Different well known functions correlating yields and conversion (previously tested with experimental data MAT) allows the evaluation of maximum point of gasoline yield. This point is used like a crackability index and qualitative point comparison of feedstock's potential. Extensive feedstocks data base from Instituto Colombiano del Petróleo (ICP) with a wide range of composition were used to build the model, including the following feeds: 1. Light feedstocks - Gasoils of refinery and laboratory cuts from different types of Colombian crude oils and 2. Heavy feedstocks - Residues or feedstocks combined (blending of gasoil [GO], atmospheric tower bottom [ATB], demetallized oil [DMO] and demetallized oil hydrotreated [DMOH] in several proportions) from the four fluid catalytic cracking units (FCCU) at Ecopetrol S.A. refinery in Barrancabermeja - Colombia. The results of model show the prediction of valuable products such as gasoline for different refinery feedstocks within acceptable accuracy, thus obtaining a reliable ranking of crackability.

Keywords: *catalytic cracking, crackability, predicting properties, statistical model.*

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En este trabajo se presenta un modelo estadístico para la predicción de la craqueabilidad de cargas (potencial para generar productos valiosos en el proceso de ruptura catalítica), basado en datos experimentales de reactividad, obtenidos por pruebas de microactividad (MAT - Reactor de Lecho Fijo de Laboratorio) y caracterización fisicoquímica detallada. Se utiliza una cantidad mínima de mediciones experimentales disponibles en una refinería promedio, a partir de las cuales se construye una descripción más completa de la carga. Esta descripción fisicoquímica es usada para predecir los rendimientos de los principales productos en un amplio rango de severidades. Diferentes funciones que relacionan los rendimientos con la conversión (previamente probadas con los datos experimentales MAT), permiten evaluar el punto de máximo potencial a gasolina y comparar cualitativamente la craqueabilidad de un conjunto de cargas. Una extensa base de datos del Instituto Colombiano del Petróleo (ICP) con un amplio rango de composiciones fue usada para construir el modelo, la cual incluyó: 1. Cargas livianas - Gasóleos de refinería y cortes de laboratorio de diferentes tipos de crudos Colombianos y 2. Cargas pesadas - Residuos o cargas combinadas (mezcla de gasóleo [GO], fondo atmosférico de vacío [ATB], aceite desmetalizado [DMO] y aceite hidrotratado desmetalizado [DMOH] en diferentes proporciones) de las cuatro unidades de ruptura catalítica (FCCU) de la refinería de ECOPETROL S.A de Barrancabermeja-Colombia. Los resultados muestran una aceptable precisión en la predicción del potencial de distintas cargas de refinería, respecto a productos valiosos como la gasolina, obteniéndose una clasificación confiable de su craqueabilidad.

Palabras Clave: *ruptura catalítica, craqueabilidad, predicción de propiedades, modelo estadístico.*

INTRODUCTION

The process of catalytic cracking is one of the most important conversion processes in a petroleum refinery. It plays a key role in the production of valuable fuels such as high octane gasoline from feedstocks with high boiling point and low economic value. The effect of the processed feedstock quality on the yields is very important and has been frequently studied in order to predict the potentiality of the feedstock in terms of its chemical composition. Although different methods have been proposed to achieve this aim, one can distinguish two major trends reported in the literature: the first one uses a pseudo reactive model whose kinetic parameters represent conversion velocity among actual yields, according to boiling point products and nature feeds (lumps). The objective in this case is to correlate feedstock characteristics with those parameters. Voltz, Nace, & Weekman (1971) and Ancheyta, López, & Aguilar (1998) used a three-lump model and correlated its parameters with structural properties in order to calculate the yield of cracking vacuum gasoils. The second and more common alternative consists in the direct correlation of properties with yields. The advantage in this case is the possibility to infer the impact of feedstock in products in one single step. Some authors like Bollas, Vasalos, Lappas, Latridis & Tsioni (2004) use a simple model reactor to involve cracking conditions (residence time and temperature) furthermore to correlate crackability with the feed characteristics.

Fisher (1990) conducted a pioneer study to understand the influence of different types of carbon and aromatic structures on the catalytic cracking products, thus proposing the well-known concept of precursors. By performing mass spectrometry studies on different vacuum gasoils, he found a direct correlation between the paraffin, cycloparaffin, and monoaromatic content with the gasoline and conversion yield at the maximum point gasoline yield, obtained in MAT tests. He concluded that this point can be utilized as a comparative parameter for feedstock quality. Similarly, Lerner & Himpsl (1997) observed that the condensation degree of aromatic compounds plays a direct role in feedstock crackability. The relation between the reaction parameters from the three-lump models and the ratio aromatic to naphthenic structures and paraffinic to naphthenic carbon published by Voltz, Nace, & Weekman, (1971) and Ancheyta, López, & Aguilar, (1998) respectively, coincide with the conclusions of Fisher (1990) and Lerner, & Himpsl (1997). Sheppard, Al-Alloush, Green,

Zagula, Young, & Wisecarver (2003) used the hydrogen to carbon plus sulfur ratio as a key factor of gasoline yield, corrected by hydrocarbon and sulfur compounds that boil below the reaction temperature (determined by mass spectroscopy studies) and affected by total nitrogen and metals. The coke yield was expressed in function of the residual microcarbon content (MCR) and on exponential equation of basic nitrogen, sulfur, and metals. The gas, LCO (Light Cycle Oil) and HCO (Heavy Cycle Oil) yields also were predicted with multiple linear correlations of MCR, metals, sulfur, nitrogen, and hydrogen. Mariaca, Rodriguez, & Maya (2003) proposed the reactive hydrogen like crackability index, determined by mass spectroscopy on vacuum gasoils. This was correlated in linear form with saturated content, hydrogen, and Kw factor (Watson's characterization factor) and expressed the cracking yields by inverse polynomial equation.

Improvement in instrumental techniques and new data processing methods have made possible the direct use of spectroscopic signals (which respond in a unique manner to the composition of a petroleum fraction) in order to infer feed properties and variation of products characteristics in conversion processes. Baldrich, & Novoa (2007) presented a recent application of visible UV spectroscopy to infer structural properties of medium distillates and light gasoils (performed by high-resolution mass spectrophotometry) with excellent predicting results. However, the instrumental techniques are strongly limited for light and medium feedstocks analysis, their availability, cost and usefulness.

Instrumental measurements are indispensable in the above mentioned works. However, parallel efforts have been made to correlate bulk properties with basic structural and molecular characteristics, in order to predict the feed crackability. This aim is better applicable to the scope of a refinery, because bulk properties are easy to measure and they are typically used in the control of processes. Shi, Xu, Cheng, Hu & Wang (1999) related gasoline and vacuum coke residues crackability by a new characterization index (KH), defined by the molecular weight, density, and atomic H/C ratio of the feedstock. Bollas *et al.* (2004) determined an index crackability and coking tendency as a function of carbon atoms number subject to cracking (non-aromatic carbon obtained by the ndM empirical method) and nitrogen atoms number. The feedstock was divided in pseudo components in order to reconstruct light and heavy fractions that exhibit better correlation with the indexes mentioned above, and

whose sum represented the characteristics of the whole feed. Ng, Wang, Fairbridge, Zhu, Yang, Ding, & Yui (2004) reintroduced the concept of precursors and correlated in linear form, the maximum gasoline yield with the saturated and monoaromatic compounds and LCO yield with the diaromatic and triaromatic compounds. Xu, Gao, Zhao, & Lin (2005) developed empirical correlations to predict gasoline and coke yields as a function of the SARA (Saturates-Aromatics-Resins-Asphaltenes) fraction composition in vacuum deasphalted residues.

In this work, bulk properties, and structural characteristics are correlated with MAT yields, for numerous gasoils and combined feedstocks. The objective of this work was to develop a physicochemical and crackability prediction for studying and evaluating the impact of feedstock composition variability in the FCC process. Reactivity data are related with simple linear functions between yields and conversion in order to: get a graphical description of feedstock reactivity and establish the maximum gasoline yield point. This point is utilized like a crackability index to define the potential feed quality.

Crackability is defined as the maximum potential of a feedstock to generate valuable products. This characteristic is closely related to feedstock structural composition due to the fact that each hydrocarbon family tends to react in similar way, having a preferential contribution to certain yield. In all cases, the increment of molecular size or carbons number facilitates the cracking of molecule. In the catalyst takes place multiple reactions like cracking, hydrogen transfer, isomerization, dehydrogenation, polymerization and condensation. Even the catalytic cracking of a linear paraffinic molecule is complex. Some authors as Dewachtere, Santaella & Fromen (1999) use sophisticated kinetic models, which involve a great number of compounds on chemical reaction networks to approach the effect of feed composition on yields and products quality. However, it is possible to highlight some general characteristics for the three main hydrocarbon families present in petroleum. Paraffin molecules produce olefins and smaller paraffins. Linear and low molecular weight paraffins do not react easily and their crackability increases with the presence of tertiary carbons, and decreases with quaternary carbons. Naphthenic molecules tend to crack to gasoline in large proportion. The quality (Octane Number) of this gasoline is higher than the gasoline generated by paraffins, with an important content of aromatic molecules as a result of naphthenic ring dehydrogenation. Aromatic molecules are considered as hard-breaking

structures. However, alkylic chains more frequent in low condensed aromatics (mono and diaromatics) can be easily detached from the aromatic ring. Aromatic molecules without lateral chains are important generators of coke and gas as the rings condensed number increases in their structure. Large molecules such as resins and asphaltenes exhibit a direct trend toward the generation of heavy fractions and coke. Therefore, the feedstocks for catalytic cracking undergo a pre-treatment process to withdraw these compounds.

The presence of contaminants, such as nitrogen, is also related to feedstock crackability due to its neutralizing effect of catalyst acidity. Metals such as Vanadium and Nickel act as strong poisons that destroy the zeolite structure. However, unlike nitrogen, their effect is not instantaneous and is reflected at long term, destroying the crystalline catalyst network. Sulfur affects mainly the distribution and quality of products, although it does not affect crackability in a significant manner. In general, the increment of these elements in the feedstock is related to the increase of aromatic structures because they represent their structural composition indirectly and can therefore correlate their reactivity. Bulk properties such as density, index refraction, and distillation curve also reflect the feedstock structural character, particularly when they are used conjointly in the classification factors (KUOP factor, Correlation Index, ndM Method, molecular weight - see annex).

When the catalytic cracking products of two feeds are compared, the relative differences between yields are maintained (the qualitative tendency not the quantitative difference of values), independently of the unit in which they are evaluated. Of course this crackability concept is true only if the catalyst and reaction conditions used on two feedstocks were the same. This suggests that there is some relation between yields obtained in different units. Ng *et al.* (2004) found a linear correlation between the yields obtained in a Riser and MAT yields at constant conversion ($\text{Riser-Yield} = b + m * \text{MAT-Yield}$). Figure 1 illustrates the crackability of two gasoils (CDU = gasoil from a crude distillation unit and VB = gasoil from a visbreaker unit), evaluated in three different units from ICP (MAT- microscale fixed bed reactor, DCR - fluidized bed pilot unit and FFB - fluidized fixed bed laboratory reactor) with the same catalyst. It is evident that relative yields between the two feedstocks are maintained, even though reactions severities are not comparable from one unit to the other.

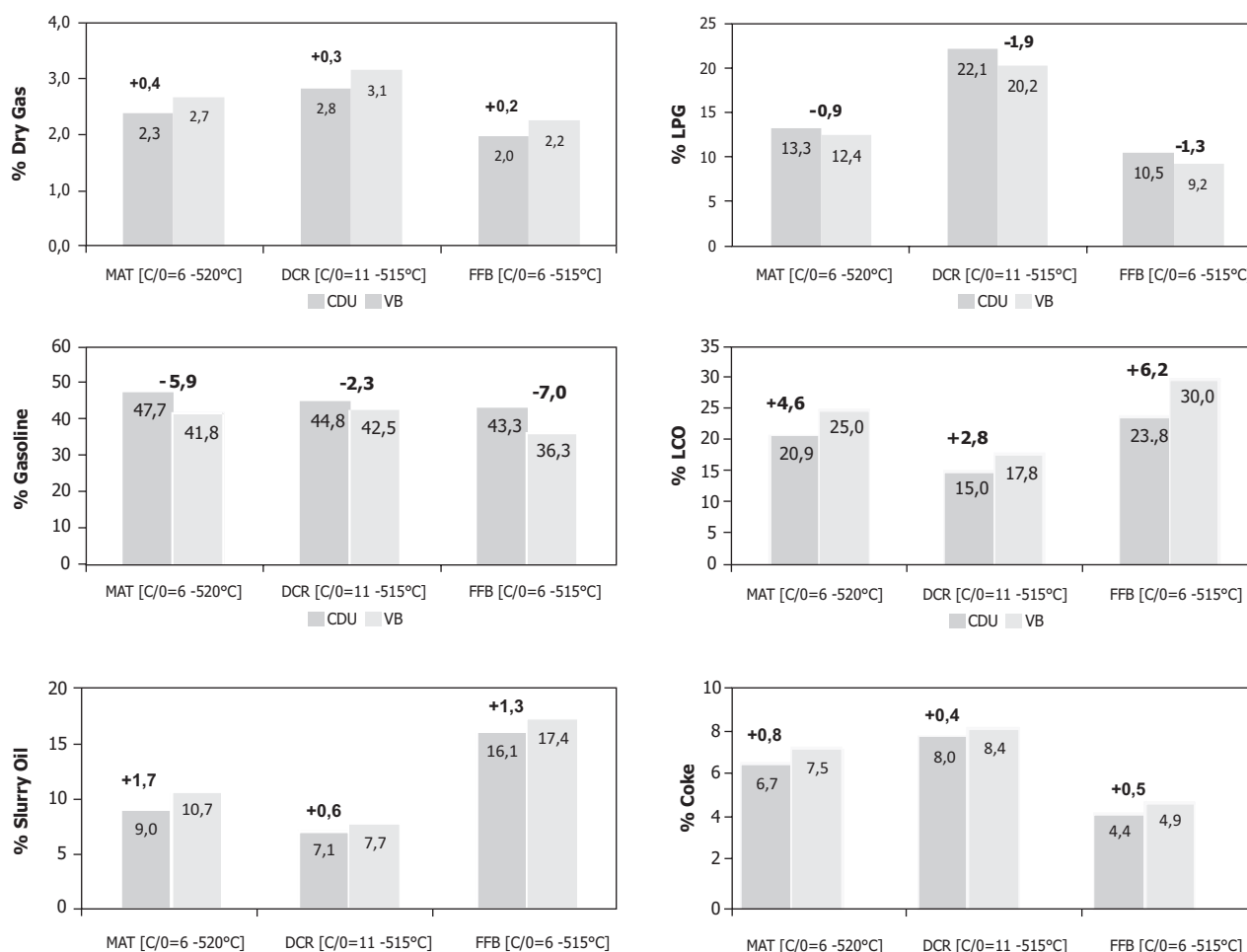


Figure 1. Catalytic cracking feedstocks yield in different units

EXPERIMENTAL WORK

The feedstocks used for model development were divided in two groups (light and heavy feedstocks). The first group corresponded to gasoils obtained through sampling procedures in refinery and laboratory cuts from Colombian crude oils of different nature (from highly paraffinic crude oil - Cusiana to highly aromatic crude oil - Castilla, Figure 2). Among refinery gasoils, most of them virgins (atmospherics and vacuum), were included gasoils from visbreaking and hydrotreating process. As a result, 82 gasoils with detailed physicochemical information constituted the database of light feedstock. The second group corresponded to residues selected from

ICP database. This database covers detailed five year physicochemical information of both individual and combined feeds of the four Barrancabermeja refinery FCC units. Principal component statistical analysis (PCA) was applied to identify those feeds with large physicochemical variability. Additionally, five typical individual feeds were selected: light and heavy vacuum gasoils (LVGO, HVGO), atmospheric tower bottom (ATB), demetallized oil (DMO), which correspond to lighter fraction from deasphalted vacuum tower residue, and hydrotreated DMO (DMOH). As a result, 21 residues with several compositions constituted the database of heavy feedstocks. Table 1 shows the wide range of physicochemical characteristics of gasoils and residues used in the construction of the model.

Table 1. Properties variation range of feedstock base model

PROPERTY		GASOILS					
		MIN	MAX	MIN	MAX		
Bulk	°API	11,5	33,4	14,9	26,4		
	MCR (%wt)	0,008	3,89	0,13	5,70		
	Sulfur (ppm)	1.210	23.360	3.030	14.804		
	Basic Nitrogen (ppm)	49	990	210	1.170		
	Total Nitrogen (ppm)	127	3.288	774	3.693		
	Nickel (ppm)	0,001	6,66	0,13	12,28		
	Vanadium (ppm)	0,003	7,92	0,58	15,98		
Structural	Simulated Distillation	IBP	121	429	122	261	
		T50%	265	572	412	596	
		FBP	454	645	559	718	
	Aromatic UV - Shell	Mono	2,00	8,03	3,25	5,69	
		Di	1,92	6,32	2,72	4,39	
		Tri	1,31	6,42	2,76	3,90	
		Tetra	0,42	6,23	2,02	5,40	
		Total	9,35	21,81	12,94	17,27	
		SARA	Saturated	27,2	79,3	26,7	65,4
		Aromatic	14,1	57,0	28,0	60,7	
		Resin	0,5	15,5	3,2	15,5	
	Classification	ndM Method	% Aromatic C	12,6	27,5	15,8	23,7
			% Paraffinic C	32,5	74,5	49,4	64,3
% Naphthenic C			10,5	41,4	16,8	34,8	
KUOP Factor		11,0	12,5	11,6	12,2		
Correlation Index		25,8	75,2	45,4	55,5		
MW (g/mol)		187	583	328	664		

The different characteristics of feedstocks ensured a strong structural variability, reflected in its hydrocarbon distribution (paraffinic, naphthenic and aromatic - Figure 3). It promoted marked differences in the yield products distribution and permitted the model to take into account different types of feedstock that a FCCU is able to process.

Reactivity experimental design used for the evaluation of feedstocks crackability (Table 2) was the result of previous experimental analysis using as starting point the conclusions of sequential design of experiments technique conducted by Reina & Osorio (2005) in the same unit. The MAT unit operating conditions

(Cat/Oil ratio, injection time and feedstock flow) were optimized keeping repeatability values less than 1% in the gasoline yield variation and mass balance between 98 and 102%. The objective was to obtain a product yield curve profile in a wide reaction conditions range. However, the temperature was fixed at 515°C for gasoils and 530°C for residues. Two base equilibrium catalysts (ECAT) from the FCCU Barrancabermeja refinery were selected, one for gasoils and other one for residues, considering good industrial performance in terms of activity and coke selectivity. In conclusion, with the MAT experimental design, it was possible to obtain complete product yield selectivity curves according with the feedstock reactivity.

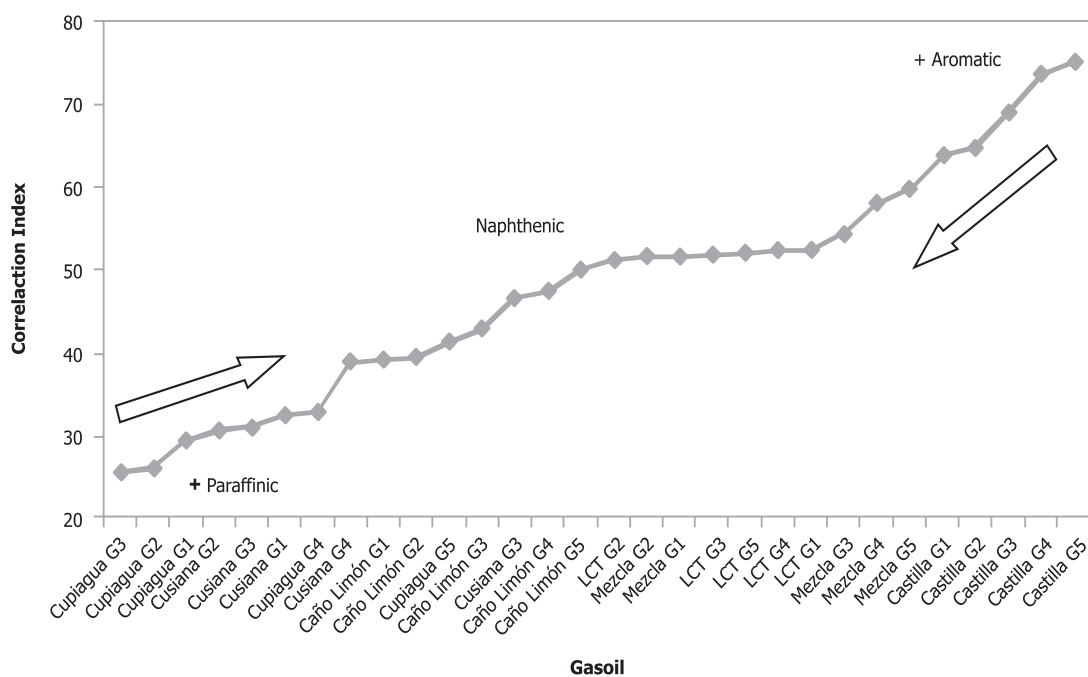


Figure 2. VGO Hydrocarbon distribution from different types of Colombian crude oils as a function of correlation index (IC)

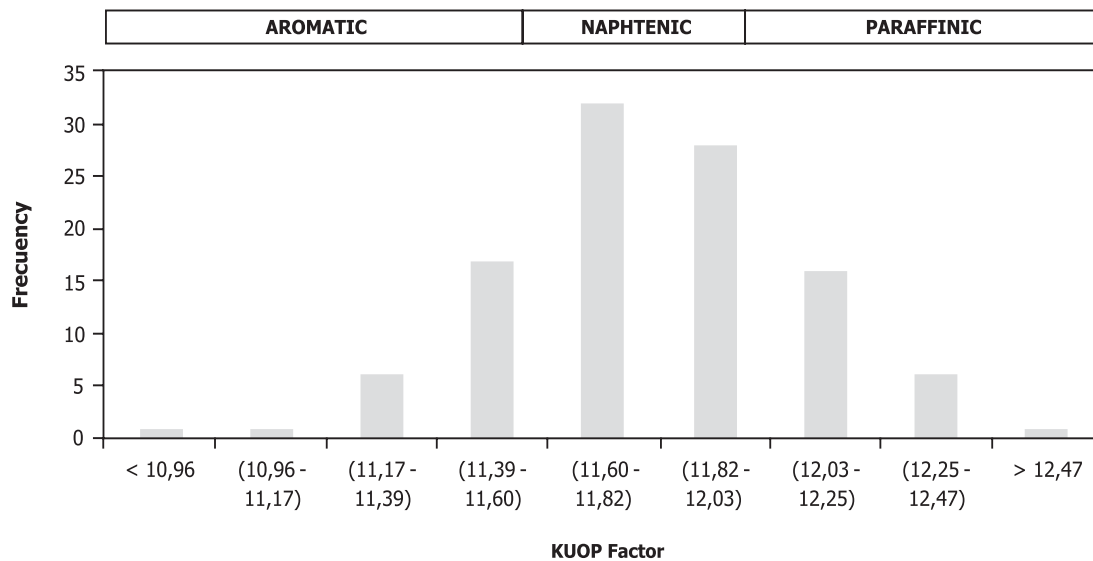


Figure 3. KUOP distribution factor of feedstocks database

Table 2. Experimental design - MAT reactivity tests

Temperature		Gas oils = 515°C - Residues = 535°C					
CONDITION	'A	B	C	D	E	F	'G
Catalyst (g)	3	5	5	5	5	5	5
Feedstock (g)	2,00	1,67	1,25	1,00	0,83	0,71	0,63
Ratio C/O	1,5	3	4	5	6	7	8
Injection Time - t(s)	60	20	20	20	20	20	40
² whsv (1/h)	40,0	60,0	45,0	36,0	30,0	25,7	11,3
Flow (g/s)	0,033	0,083	0,063	0,050	0,042	0,036	0,016

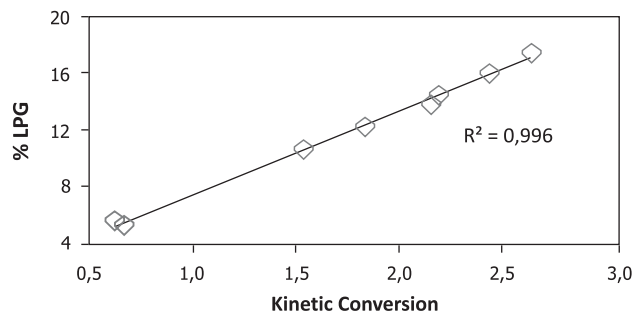
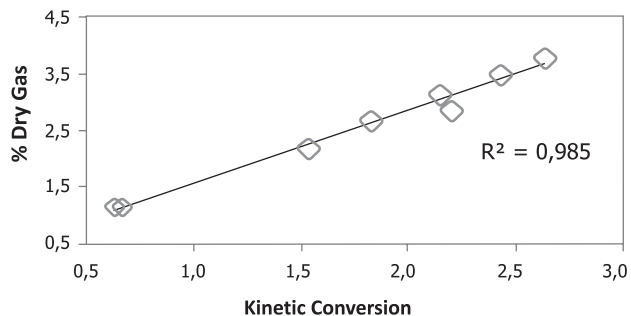
1. Experiments A y G were duplicated. 2. whsv stands for "weight hourly space velocity" = $3600 / [t \cdot C/O]$

Linear functions between product yields and feedstock conversion (Table 3), some of them frequently used to check the coherence of experimental results, are utilized as interpolation tool. Figure 4 shows an example of the MAT tests for a light gasoil, exhibiting excellent adjustment for each function. The gasoline conversion curve is fit by the linear function between

its selectivity and the kinetic conversion. Its optimum point corresponds to the maximum gasoline yield that can be obtained from the feedstock at the MAT unit. The conversion at this point can be replaced in the other functions to estimate the yields else. This maximum point is used for the construction of a feedstocks crackability ranking.

Table 3. Linear Function of Catalytic Cracking Yields [$y = b + mx$]

y (Product Yield Parameter)		x (Conversion Parameter)	
Coke	CK	Kinetic conversion	$X/(100-X)$
Gasoline Selectivity	GSL/X	Kinetic Conversion	$X/(100-X)$
LCO Selectivity	$LCO/(100-X)$	Conversion	X
Slurry Oil	SO	Inverse conversion	$(1-X/100)^n$
LPG	LPG	Kinetic Conversion	$X/(100-X)$
Dry Gas	DG	Kinetic Conversion	$X/(100-X)$



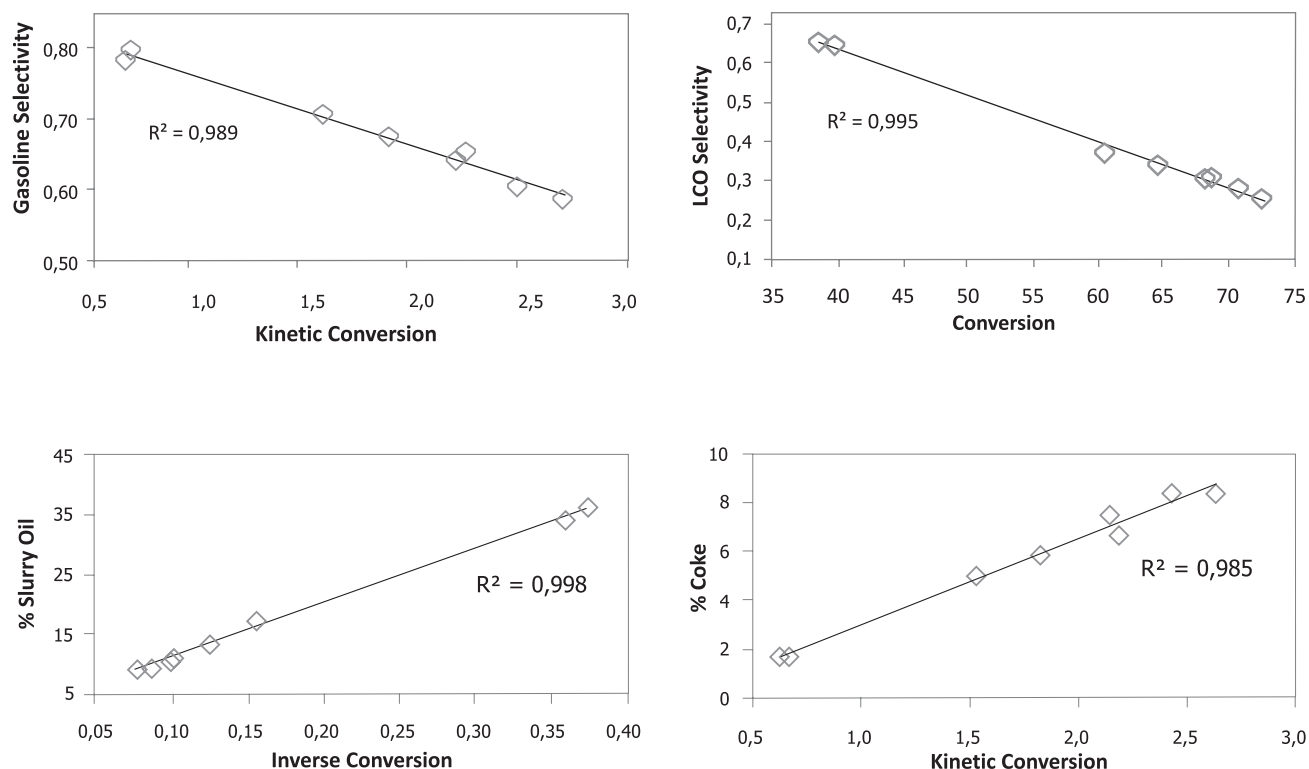


Figure 4. MAT results for light refinery gasoil (MAT tests [◊], Regression line [-])

RESULTS

Prediction of Physicochemical Properties

Correlations were developed for gasoils and combined feedstocks separately, by multiple linear regressions. The objective was to predict full physicochemical characteristics from a minimum number of basic properties. A set of equations was established to estimate bulk (Aniline point, MCR, Viscosity and Refraction Index), structural (Shell aromatics and SAR distribution) and composition characteristics (Nitrogen and Sulfur), from only two basic feedstock properties (API and Distillation Curve). Table 4 shows the mean absolute errors (MAE) obtained with the correlations and experimental repeatability values of each laboratory test (R. Exp.). The correlation for sulfur exhibits the largest mean error. Considering that sulfur concentration in feedstocks to FCCU is about 8.000 ppm, the estimation relative error would be 11%, which is acceptable. Regarding the remaining correlations, a

better adjustment to the experimental data was achieved with relative errors below 5%, and absolute average error close to experimental repeatability value for each test.

Due to the importance of the aromatic fraction and its distribution (mono, di, tri and tetra aromatics) in feedstock crackability, prediction and validation results of this property are presented. Figure 5 compares the estimation from the FCC-SIM[®] rigorous commercial simulator of the catalytic cracking process by KBC Profimatics (Copyright © KBC Advanced Technology plc Version 2004-1.5) with the correlation developed. Better adjustment of the correlation to experimental data (values are showed in ascending order and graphic scales are identical) is clearly observed. Figure 6 shows the validation conducted with gasoils outside development database (validation database). Good predictive results were attained, including two foreign gasoils from imported crude oils (ABO and Escravos gasoils -Six first points on the graphics).

Table 4. Correlations for physicochemical properties prediction

PROPERTY	Gasolins			Residues		
	Correlation	MAE	R ²	Correlation	MAE	R ²
Sulfur - S (ppm)	$f(\text{IC}, \text{IR}, \text{API})$	924	0,94	$f(\text{IC}, \text{IR}, \text{API})$	706	0,79
Basic Nitrogen - NB (ppm)	$f(\text{IR}, \text{IR}/\text{API})$	42	0,95	$f(\text{API}, \text{IC})$	34	0,93
Total Nitrogen - NT (ppm)	$f(\text{NB})$	146	0,94	$f(\text{NB})$	142	0,89
Refraction Index - IR	$f(\text{API})$	0,0018	0,98	$f(\text{API})$	0,0009	0,98
MCR (%wt)	$f(\text{NB}, T_{50\%}, T_{90\%}, K_{\text{UOP}}, \text{MW}, \text{Tetra})$	0,13	0,90	$f(\text{API}, \text{NB}, \text{IR}, K_{\text{UOP}})$	0,17	0,86
Aniline Point - PA (°C)	$f(\text{IC}, \text{API})$	2,1	0,96	$f(\text{IC}, \text{API})$	1,5	0,96
Aromatic - Arom (%wt)	$f(\text{NB}, \text{API})$	2,5	0,90	$f(\text{NB}, \text{API})$	2,9	0,82
Saturated - Sat (%wt)	$f(\text{API}, \text{NB}, \text{Arom})$	1,6	0,96	$f(\text{API}, \text{NB}, \text{Arom})$	1,8	0,97
Resin - Res (%wt)	$f(\text{NB}, \text{IR}, \% \text{Cr})$	1,0	0,79	$f(\text{IR}, \text{Sat}, \text{Arom})$	0,4	0,92
Monoaromatic - Mono (%wt)	$f(\text{API}/\text{PA}, \text{IC} \cdot \text{PA}, \text{PA})$	0,37	0,78	$f(\text{API}/\text{PA}, \text{IC} \cdot \text{PA}, \% \text{Cp}, \text{API}, \text{IC})$	0,26	0,86
Diaromatic - Di (%wt)	$f(\text{API}/\text{PA}, \text{IC} \cdot \text{PA}, \text{PA}) - \text{Mono}$	0,55	0,86	$f(\text{API}/\text{PA}, \text{IC} \cdot \text{PA}, \text{API})$	0,43	0,82
Triaromatic - Tri (%wt)	$f(\text{IR}, \text{IC}, K_{\text{UOP}}) - \text{Mono}$	0,18	0,85	Total - Mono - Di - Tetra	0,37	0,88
Tetaromatic - Tetra (%wt)	$f(\text{KUOP}, \text{API}, \% \text{Ca})$	0,52	0,72	$f(\text{Total}, \text{S}, \text{PA}, \text{IR})$	0,23	0,93
Totalaromatic - Total (%wt)	Mono + Di + Tri + Tetra	0,40	0,80	$f(\text{API}, \text{IC}, \text{IR})$	0,57	0,89
Viscosity 80°C - V80 (cSt)	$\exp[f(T_{50\%}/\text{API}, \text{IC}/\text{MW})]$	1,14	0,90	$\exp[f(\text{API}, \text{IC}, T_{50\%}, \text{MW}, \text{IC}/\text{MW})]$	1,18	0,97
Viscosity 100°C - V100 (cSt)	$\exp[f(T_{50\%}/\text{API}, \text{IC}/\text{MW})]$	1,09	0,98	$\exp[f(\text{API}, \text{IC}, T_{50\%}, \text{MW}, \text{IC}/\text{MW})]$	1,01	0,98

1. The operator means: $(P_1, P_2, \dots, P_n) = a_1 + b_1 \cdot P_1 + b_2 \cdot P_2 + \dots + b_n \cdot P_n$, where $a_1, b_1, b_2, \dots, b_n$ are different constants for each correlation and P_1, P_2, \dots, P_n are properties correlated.

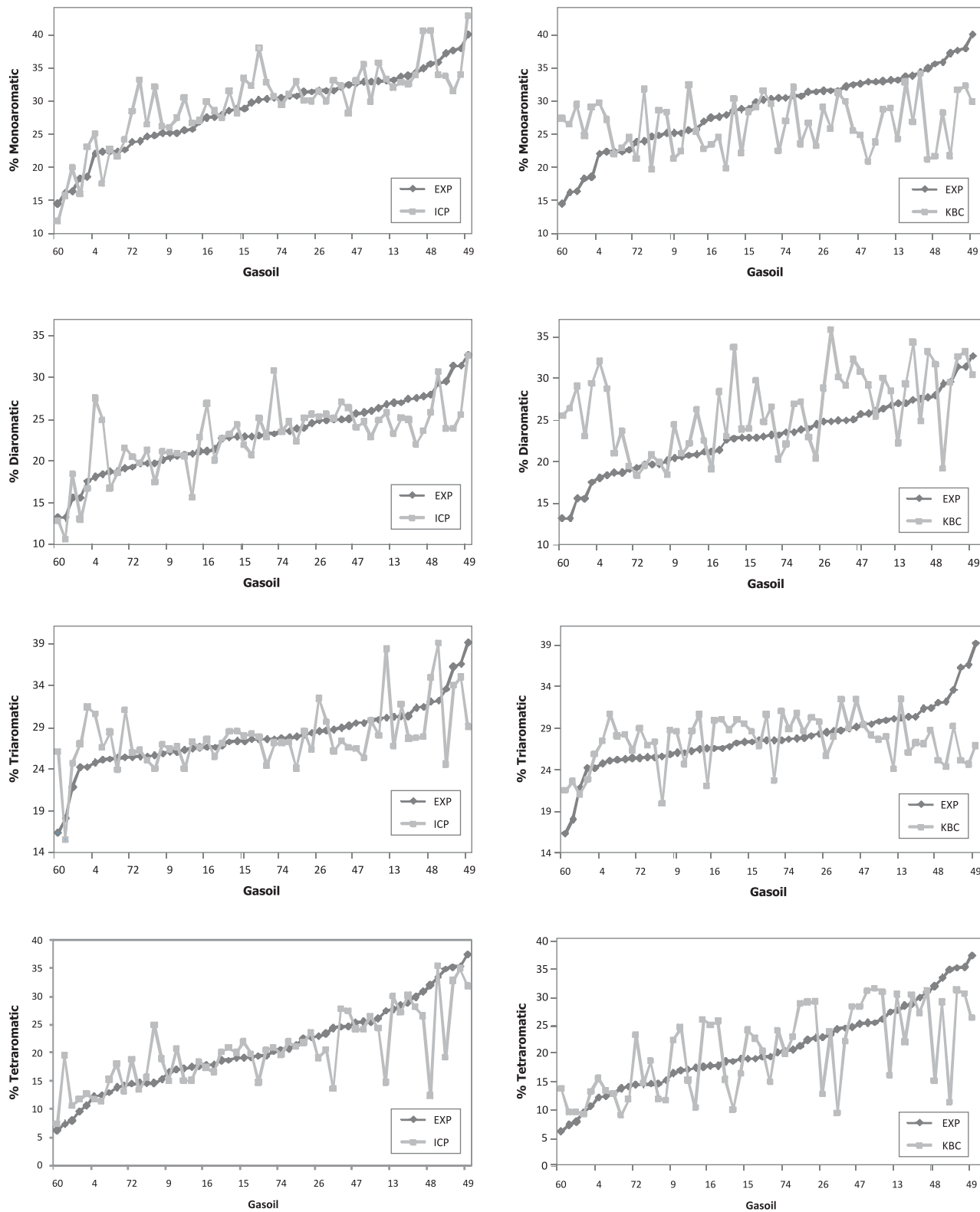


Figure 5. Aromatic hydrocarbon distribution prediction of gasoils (EXP: experimental development database, ICP: prediction of correlation developed and KBC: prediction of FCC process simulator)

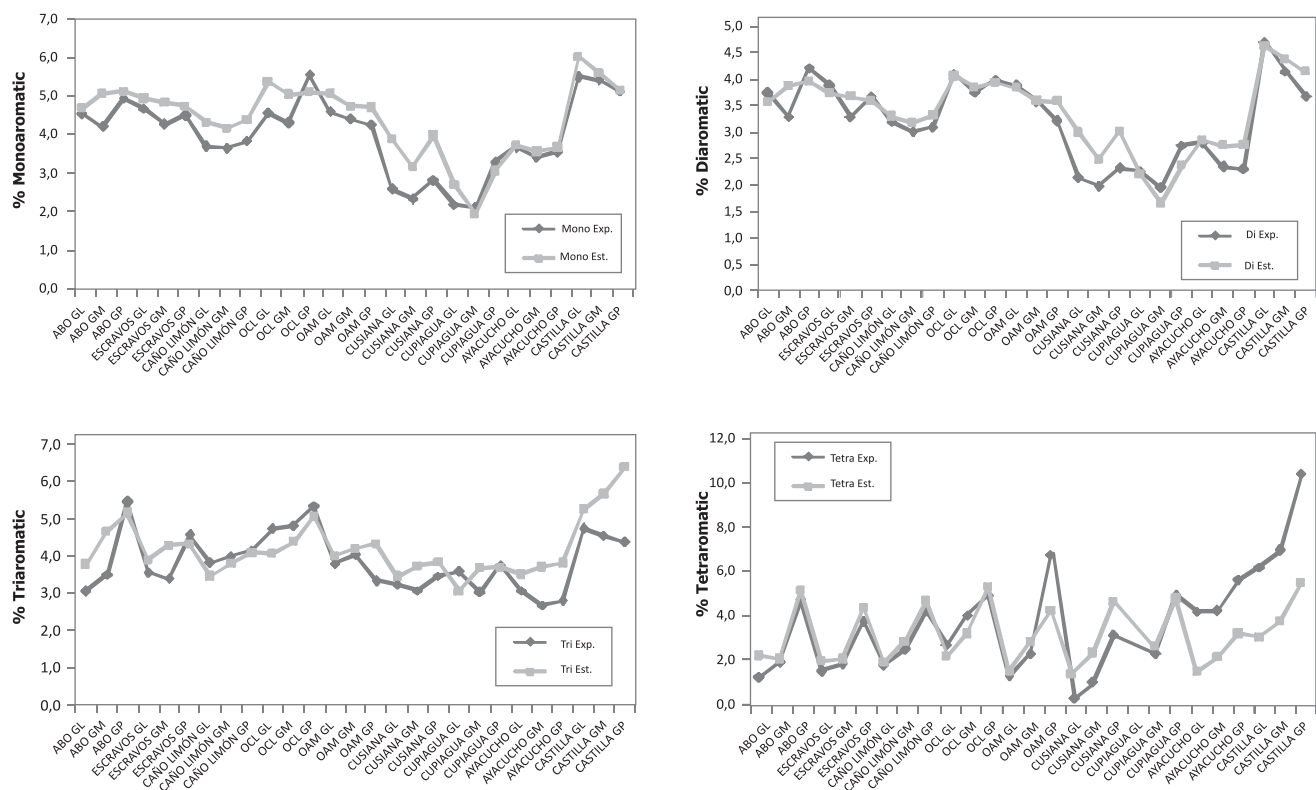


Figure 6. Aromatic carbon distribution prediction of gasoils (Exp.: experimental validation database and Est.: prediction of correlation developed)

Prediction of Crackability

Yields estimated at different reaction severities (Table 2) in the MAT unit were directly correlated to feedstock properties by multiple linear regressions (Table 5). The correlated properties and mathematical formulation obtained for each yield correlation were kept independently of reaction condition. Therefore, only the constants reflected the effect of severity changes. This indicates that the correlated properties have strong influence on the corresponding yields, despite their selection was only consequence from statistical criteria. Differences between correlated properties for gasoils and residues are reasonable because of high boiling points and mixed composition of residues in contrast with the more defined origin of gasoils and its lower boiling points. However, the correlation for gasoline prediction results in function of condensed aromatic compounds with 3 and 4 rings, and basic nitrogen for both cases. This result seems to contradict the influence of saturated precursors found by Fisher (1990) and Ng *et al.* (2004), and it partially agrees with

Voltz *et al.* (1971) and Ancheyta *et al.* (1998) regarding the aromatics and nitrogen, respectively. Results obtained by Xu *et al.* (2005) also show an acceptable trend with SARA distribution. However, prediction is not sufficiently good, reporting a mean absolute error higher than 2% in gasoline for a small set of feedstocks (6 deasphalted residues), as compared to 0,86% for the correlation obtained for the wide range composition of feedstocks (98 altogether).

The aromatic distribution has an important impact in general feedstock conversion. As Lerner & Himpel (1997) concluded, there is a great influence of the feedstock condensation degree on crackability and it is more applicable for residues due to its highest poliaromatics content. This conclusion is observed in the participation of each type of aromatic structure in correlations: monoaromatic (1 correlation), diaromatic (3 correlations), triaromatic (8 correlations) and tetraaromatics (11 correlations).

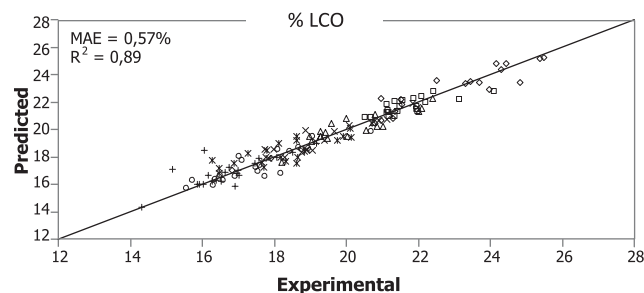
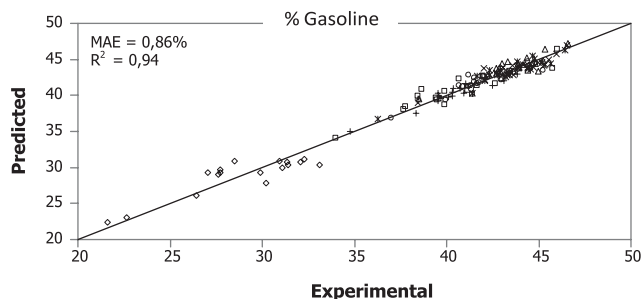
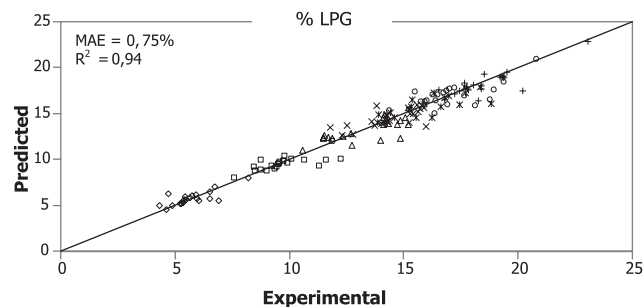
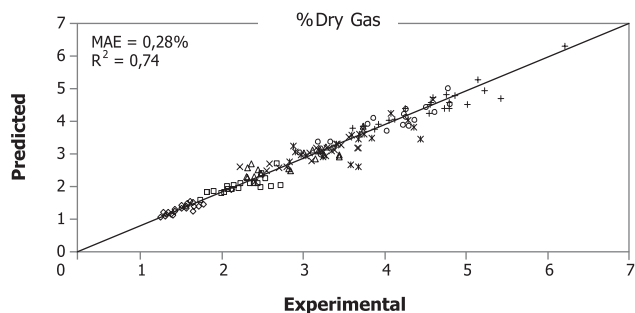
The influence of classification factors was very low and no correlation between the ndM classification and crackability was observed, as Ancheyta *et al.* (1998) and Bollas *et al.* (2004) found it. For gasoils, whose MCR content is very low, aromatics and resins influenced coke prediction severely, while residues were better correlated with tri and tetraaromatics, MCR, molecular weight, boiling point, and Total Nitrogen. All these compounds are indicators of a feedstock coking tendency.

The only heteroatom that had an impact on crackability was nitrogen, particularly its basic character content. Sulfur and metals did not exhibit any relation to catalytic reactivity. Figure 7 shows parity plots for the prediction of different yields of combined feedstocks. An acceptable correlation and very low prediction average absolute error is observed, comparable to the experimental repeatability of MAT tests.

Table 5. Correlations for catalytic cracking yield MAT prediction

Yield	Gasoils ^{1,2}	Residues ^{1,2}
Dry Gas (DG)	Dry Gas (DG) = Xn - (Di, Tri, Tetra, NB)	(Tetra, API, T50%, MW)
LPG	LPG (Di, Tri, Tetra, NB) - GSL	(Tri, Tetra, T50%, API)
Gasoline (GSL)	Gasoline (GSL) (Tri, Tetra, NB)	(Tri, Tetra, T50%, NB)
LCO	LCO (Mono, Di, Tri, Tetra, IC, T10%)	(Tri, Tetra, T50%, IC)
Slurry Oil (SO)	Slurry Oil (SO) = 100 - X - LCO	(Tri, Tetra, T50%, NB)
Coke (CK)	Coke (CK) (Arom, Res, %Cn)	(Tetra, MCR, NT, T50%, MW)
Conversion (X)	Conversion (X) = Xn - CK	= DG + LPG + GSL + CK
Net Conversion (Xn)	Net Conversion (Xn) (Di, Tri, Tetra, NB)	= X - CK

1. The operator means: $(P_1, P_2, \dots, P_n) = a_1 + b_1 \cdot P_1 + b_2 \cdot P_2 + \dots + b_n \cdot P_n$, where $a_1, b_1, b_2, \dots, b_n$ are different constants for each correlation and P_1, P_2, \dots, P_n are properties correlated. 2. The correlation of specific yield keeps its mathematical formulation¹ for each reaction condition of table 2, but the constants $a_1, b_1, b_2, \dots, b_n$ are different for each condition.



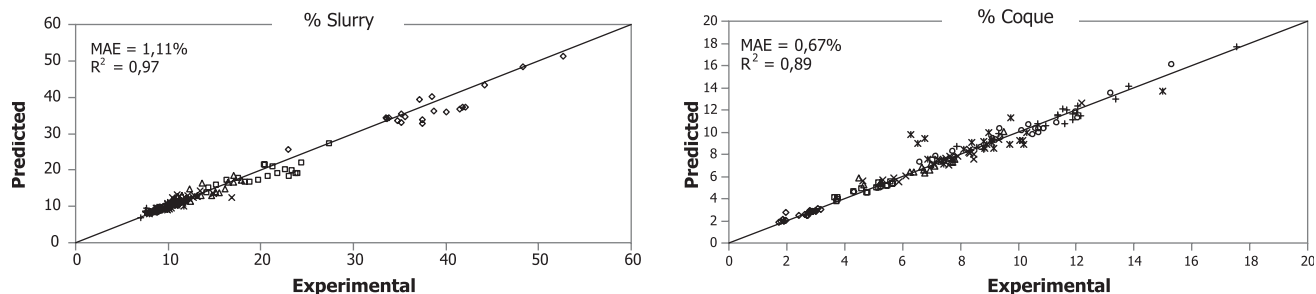


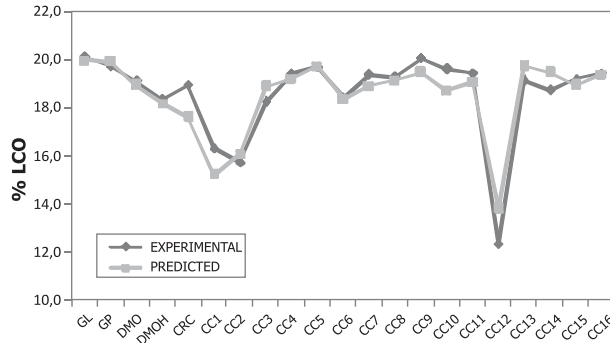
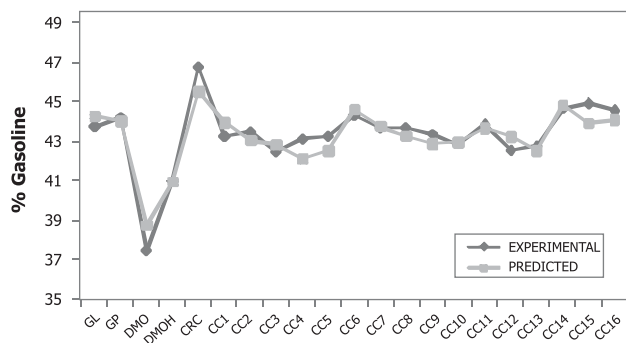
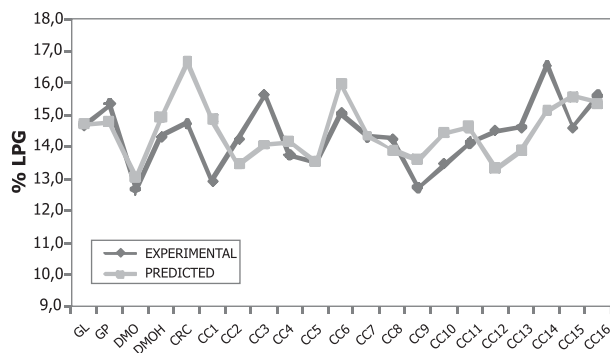
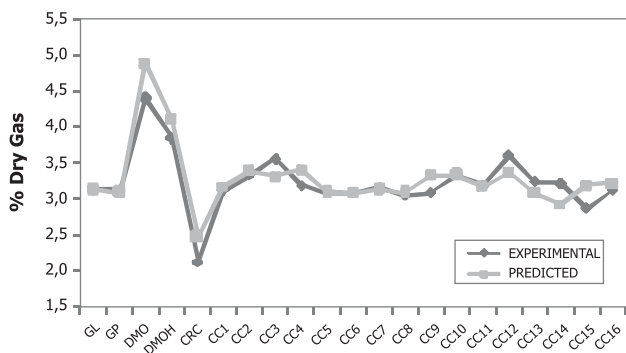
Figure 7. MAT yields prediction of residues
(Reaction condition, referred in Table 2): A [◇] B [□] C [△] D [X] E [*] F [○] G [+]

Evaluation of maximum yield gasoline was conducted on each feedstock, according to the following procedure:

- The slope and intercept of the linear relation between gasoline selectivity and kinetic conversion was calculated.
- A theoretical gasoline selectivity curve was obtained replacing, in the previously equation, conversion values between 0% and 100%.
- The maximum point of gasoline yield is calculated from gasoline selectivity curve optimization.
- The conversion obtained was used in the other selectivity curves to calculate the yield of re-

maining products (dry gas, LPG, LCO, slurry oil and coke).

Figure 8 shows the yields at the maximum gasoline point, estimated from MAT experimental data and from calculated values applying the developed correlations. The model responds satisfactorily to the experimental trend observed and highlighted with excellent fit Gasoline, LCO and Coke prediction. Particularly, for the first five points of each graph that correspond to individual feedstocks, whose characteristics represent extreme of possible feedstocks to FCCU.



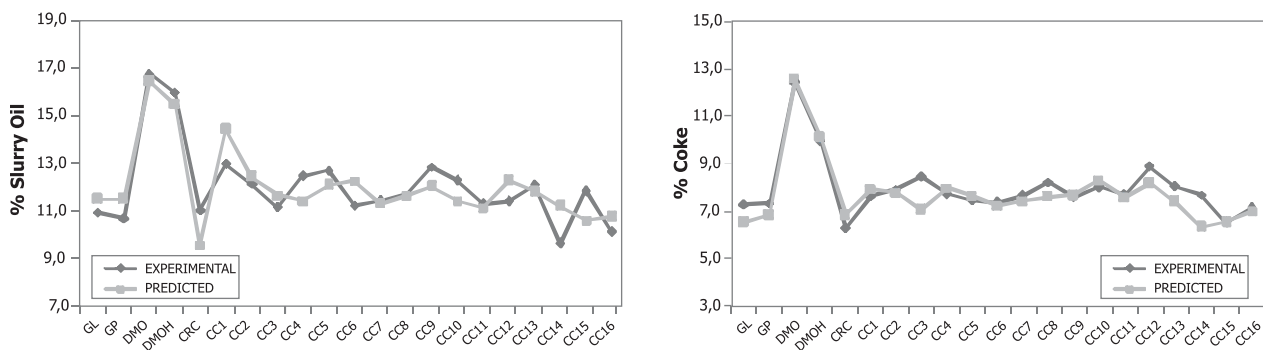


Figure 8. Crackability prediction of residues (EXPERIMENTAL: experimental development database and PREDICTED: prediction of correlation developed)

The concept of crackability used in this work and the statistical model (physicochemical estimation correlations, crackability estimation correlations, and estimation of maximum gasoline potential) were validated with two different feedstock groups: the first group corresponds to a set of processed gasoils of historical background FCCU Cartagena Refinery (ECOPETROL S.A.) and an imported gasoil (API = 26, NB = 210 ppm, MeABP = 790 °F, and Total Aromatics = 11,3%). The second group consists of different and more recently combined feedstocks outside of development database (validation database). These feeds were evaluated at the ICP FCC pilot plant (DCR), at different severities. Gasoils were evaluated with a different catalyst to the one used in the MAT reactivity tests of developed model, while the same catalyst was used with the combined feedstocks.

With only two basic available properties in refinery (distillation curve and API gravity), the remaining characteristics and the MAT yields of each feedstock were calculated. The procedure to evaluate maximum yield

of gasoline was then applied on experimental DCR pilot plant data and estimated MAT data by statistical model. Figure 9 compares the results. It is observed that the relative variation of gasoline potential at the pilot plant is very similar to the variation estimated by the model in both cases. It is important to highlight that the use of a different catalyst for gasoils did not affect such trend and values proximity is only coincidental. Even for imported gasoils, whose nature and origin is unknown and almost certainly is different with respect to gasoils of Colombian crude oils, the relative gasoline potential is well predicted. On the other hand, combined feedstocks (evaluated with the same base catalyst used in the development model) show Riser gasoline yields higher than estimated values for MAT, but alike gasoils the relative variation is very similar. These results are consistent with the crackability concept and evaluation proposed in this work, in which the maximum gasoline yield is a key parameter to evaluate the potential of feedstocks on catalytic cracking process, independent from the inherent conditions of unit and catalyst.

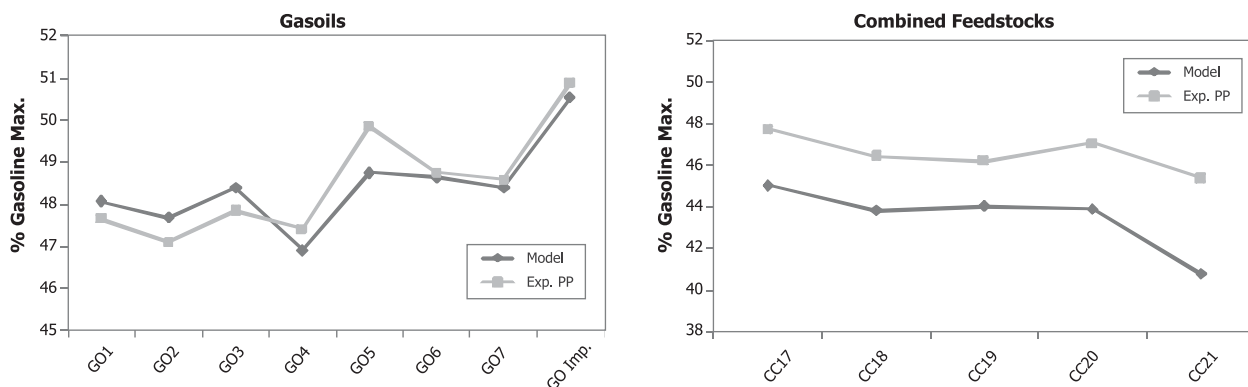


Figure 9. Crackability prediction of gasoils and residues (EXPERIMENTAL: experimental validation database and Model: prediction of statistical model developed)

Applied cases

The following three illustrative cases show the influence predicted by the model of different feeds on catalytic reactivity.

Case I: Five full range vacuum gasoils from Colombian crude oils, with widely different characteristics, showed a trend that is coherent to their nature (Table 6 and Figure 10). For the Cupiagua crude oil gasoil, higher gasoline yield is observed since this is a highly paraffinic feedstock (KUOP = 12,35 and %Cp = 71,34). On the contrary, the gasoil from the Castilla crude oil exhibits lower gasoline yield, due to its high aromatic nature (KUOP = 11,25 and %Ca = 25,2). The gasoil from the Cusiana crude oil is highly paraffinic and reach similar conversions as well as Cupiagua. However, its higher nitrogen content affects its gasoline selectivity. The remaining gasoils (OCL and Caño Limon) are characteristic of intermediate - type crude oil (naphthenic oils), as it was estimated by the model.

Table 6. Physicochemical properties of virgin gasoils

Property	C. Limón	OCL	Cusiana	Cupiagua	Castilla
API	23,5	19,5	25,6	29,6	15,1
S (ppm)	6.620	8.370	3.750	1.410	20.340
N (ppm)	1.640	1.030	590	430	1.770
KUOP	11,94	11,6	12,09	12,35	11,25
%Ca	17,16	20,26	15,16	11,17	25,2
%Cn	20,68	26,26	19,38	17,48	53,38
%Cp	62,16	52,78	65,45	71,34	46,62

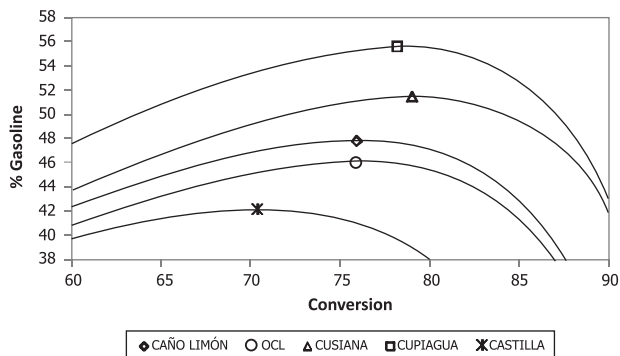


Figure 10. Crackability evaluation of virgin gasoils

Case II: A virgin gasoil from crude distillation unit (CDU) is compared to gas oil from the visbreaking process on vacuum tower residua (VB) and with a blend of 80% weight of CDU + 20% weight VB (Figure 11). It is then coherent that the maximum gasoline point and its conversion for the virgin gasoil are greater than visbreaking gasoil point. Obviously, the blend should be in between. Besides, it is observed that even though the VB gasoil fraction is only 20%, it exerts a much greater influence over blend crackability, affecting its selectivity to gasoline severely.

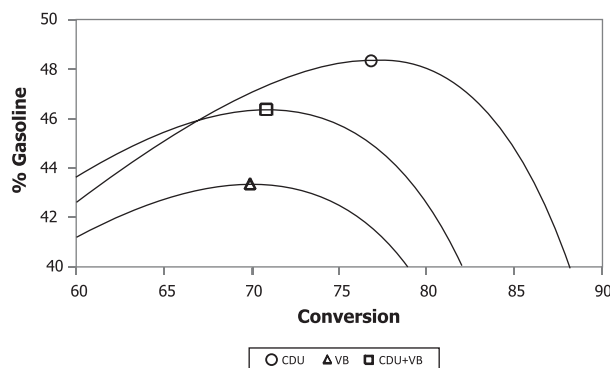


Figure 11. Crackability evaluation of gasoils from visbreaking process on vacuum residua

Case III: A demetallized residue (DMO) is hydrotreated at three different temperatures. Its quality for catalytic cracking increases with the hydrotreatment severity because of replacement of contaminant elements such as sulfur, nitrogen and metals by hydrogen and because of saturation of aromatic rings. Therefore, the gasoline potential of the feedstock also increases with the severity of the hydrotreatment (Figure 12), where DMOH 1, 2, and 3 were processed at 330, 350 and 370 °C temperatures, respectively.

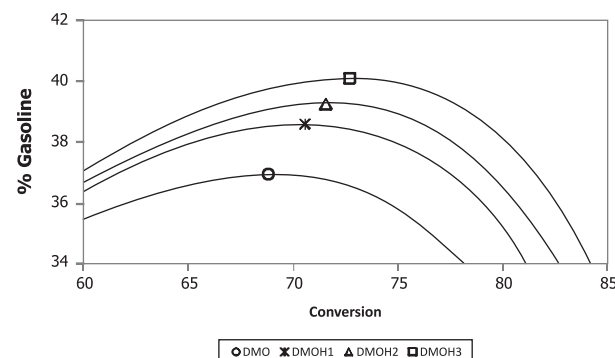


Figure 12. Crackability evaluation of hydrotreated residues

CONCLUSIONS

- A statistical model for physicochemical properties and crackability prediction of feedstocks for fluid catalytic cracking process was completed. The model exhibits acceptable description and agreement with the experimental data of a broad set of feedstock from different origins and characteristics. Basic properties such as distillation curve and density allow calculating structural and composition properties such as the content of basic nitrogen, distribution of aromatic carbon and SAR distribution, among others. The correlations developed suggest that these properties are strongly related to the capability of feedstocks to catalytic cracking, especially the condensation degree or the presence of 3 or more aromatic ring structures.
- The use of linear functions between conversion and products is a simple but effective way to obtain the maximum gasoline potential point. This parameter was correctly estimated by the model and its use as a crackability index was found independent from reaction conditions, catalytic unit and catalyst. In conclusion, even an absolute devoid of detailed physicochemical and catalytic reactivity data (typical at refinery because of logistic and economic inconveniences), can be solved by models based on basic feedstock properties.

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ANNEX

Characterization factor (KUOP). [Referred by Algelt & Boduszinski (1994)]: This factor determines

the dominant hydrocarbon family in the composition of a petroleum fraction. Values equal to or greater than 12 indicate paraffinic nature. Values equal to or less than 11,5 indicate aromatic nature and values of between 11,5 and 12 indicate a naphthenic or intermediate nature.

$$KUOP = (CABP + 459,7)^{1/3}/SG \quad (1)$$

SG = Specific gravity at 60 °F

CABP = Cubic Average Boiling Point in °F

Correlation Index (IC). [Referred by Speight (1991)]: This index was developed by the U.S.B. of Mines, based on the graph of specific gravity and the reciprocal value of the boiling point (K) of pure hydrocarbons. It indicates the type of hydrocarbon prevailing in a feedstock, using a wider range of values than the KUOP factor and marked by the most paraffinic structure (the hexane with a IC of 0) and the most aromatic structure (the benzene with a IC of 100). The cyclohexane have a IC value of 50.

$$IC = 473,7 \times SC - 456,8 + 456,8 / MeABP \quad (2)$$

MeABP = Mean Average Boiling temperature in K

ndM Method. [Referred by Algelt & Boduszinski (1994)]: It estimates the aromatic carbon (%Ca), paraffinic carbon (%Cp), and naphthenic carbon (%Cn) distribution from the refraction index (n), density (d), and molecular weight (M). The following correlations must be used with measurements at 70 °C.

$$\% Ca = Ax + 3660/M \quad (3)$$

$$\% Cr = By - 3S + C/M \quad (4)$$

$$\% Cn = \% Cr - \% Ca \quad (5)$$

$$\% Cp = 100 - \% Cr \quad (6)$$

$$x = 2,42(n - 1,46) - (d - 0,828) \quad (7)$$

$$y = (d - 0,828) - 1,1(n - 1,46) \quad (8)$$

$x > 0$, A = 410

$x < 0$, A = 720

$y > 0$, B = 775 y C = 11500

$y < 0$, B = 1440 y C = 12100

S = Sulfur content (% weight)

Molecular Weight (MW). According to the UOP 375-86 method.

$$MV = \text{anti log}[I \times M + J + (L / M)] \quad (9)$$

$$I = -0,000067214393D2 - 0,0013189667D + 0,002322975 \quad (10)$$

$$J = 1,496307D2 - 2,4028499D + 2,7013135 \quad (11)$$

$$L = -166,84095D2 - 240,43988D - 92,008149 \quad (12)$$

M (MeABP) = Mean Average Boiling Point in °F

C (CABP) = Cubic Average Boiling Point in °F

D = Specific gravity at 60 °F

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