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Adsorption of pharmaceuticals onto activated carbon fiber cloths

Modeling and extrapolation of adsorption isotherms at very low concentrations

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Abstract

Activated carbon fiber cloths (ACFC) have shown promising results when applied to water treatment, especially for removing organic micropollutants such as pharmaceutical compounds. Nevertheless, further investigations are required, especially considering trace concentrations, which are found in current water treatment. Until now, most studies have been carried out at relatively high concentrations (mg.L⁻¹), since the experimental and analytical methodologies are more difficult and more expensive when dealing with lower concentrations (ng.L⁻¹). Therefore, the objective of this study was to validate an extrapolation procedure from high to low concentrations, for four compounds (Carbamazepine, Diclofenac, Caffeine and Acetaminophen). For this purpose, the reliability of the usual adsorption isotherm models, when extrapolated from high (mg.L⁻¹) to low concentrations (ng.L⁻¹), was assessed as well as the influence of numerous error functions. Some isotherm models (Freundlich, Toth) and error functions (RSS, ARE) show weaknesses to be used as an adsorption isotherm model with Marquardt's percent standard of deviation was evidenced as the best combination model, enabling the extrapolation of adsorption capacities by orders of magnitude.

Keywords

Extrapolation; adsorption capacities; trace concentrations; activated carbon fiber cloths; micropollutants

1. Introduction

The contamination of aquatic systems by pharmaceutical residues is widely reported and constitutes a growing concern and risk for the environment and human health (Das *et al.*, 2014; Gamarra *et al.*, 2015; Huerta-Fontela *et al.*, 2011). For instance, Loos *et al.*, (2009) detected diclofenac in 83 % of their samples in European rivers. The maximum concentration was 11 ng.L⁻¹ (de Jesus Gaffney *et al.*, 2015). Likewise, carbamazepine and acetaminophen were identified and quantified at concentrations of 200 ng.L⁻¹ in surface waters (Segura *et al.*, 2011). Exposure to environmentally concentrations of pharmaceutical residues (Carbamazepine, Diclofenac and Acetaminophen) could alter biomarkers and biochemical processes of C. *fluminea* and duckweed plants (Chen *et al.*, 2014; Kummerová *et al.*, 2015). However, risks for human health were considered negligible by Webb *et al.*, (2003).

Adsorption processes are recognized among the most efficient, promising and widespread water treatments for the removal of pharmaceuticals (Foo and Hameed, 2009). In practice, granular activated carbon (GAC) and/or powdered activated carbon (PAC) are used to eliminate organic pollutants. According to (Rigobello *et al.*, 2013; Sotelo *et al.*, 2014), the adsorption capacities of diclofenac onto granular activated carbon are about 230 mg.g⁻¹ for a residual aqueous concentration of 40 mg.L⁻¹. For carbamazepine, Cai and Larese-Casanova (2014) found capacities of 200 mg.g⁻¹ onto granular activated carbon for a concentration of 0,5 mg.L⁻¹. For a residual concentration of 500ng.L⁻¹, (Yu et al., 2008) found capacities of 1 mg.g⁻¹. More recently, activated carbon fiber cloths (ACFC) have also been studied for water treatment (Faur-Brasquet *et al.*, 2002). The advantages of such textiles are their high specific surface areas, predominantly microporous texture and large adsorption capacities (Ayranci and Hoda, 2005). Therefore, ACFC were more efficient in terms of adsorption rate and selectivity than granular activated carbon for the removal of phenols (Dabrowski *et al.*, 2005). However, to our knowledge, few studies have dealt with the adsorption of pharmaceutical residues onto ACFC (Ayranci and Duman, 2006; Bayram and Ayranci, 2012; Guedidi *et al.*, 2014).

The design of the adsorption process is generally based on the equilibrium data for the particular system: adsorbent vs. adsorbate. Thus, relevant and accurate models for the sorption equilibrium are critical to predict the performances of the treatment. Adsorption isotherms (relationship between the amount adsorbed at equilibrium and the residual aqueous concentration for a given temperature) are usually reported for a narrow range of concentrations. Moreover, high aqueous concentrations (about mg.L⁻¹) (Brasquet *et al.*, 1996) of pollutants are commonly considered since trace concentrations are hardly achievable unless a complex and costly analytical strategy is carried out. The question is whether the adsorption capacities obtained at high concentrations can be easily transposed to realistic trace conditions. As mentioned above, in aquatic compartments, the concentrations of pharmaceutical residues are very low (in the ng.L⁻¹ range) and the determination of adsorption isotherms at such environmental concentrations is difficult, requiring powerful analytical tools such as liquid chromatography coupled to mass spectrometry and time-consuming sample preparation (solid phase extraction).

The accuracy of the extrapolation is greatly dependent on the isotherm model considered as well as the adjustment procedure used for the optimization of the isotherm parameters. In the literature, numerous isotherm models are available, starting from the Freundlich and Langmuir models, which have been widely used and validated to describe equilibrium relationships between various adsorbents and adsorbates (Ho, 2004; Nam *et al.*, 2014; Sotelo *et al.*, 2014; Yu *et al.*, 2008). Nam *et al.* (2014) have shown a better fit of adsorption onto granular activated carbon with the linear form of the Langmuir model, compared to the Freundlich model, for acetaminophen adsorption. These authors used concentrations of pollutants between 20 and 500 ng.L⁻¹, with a mass of granular activated carbon of 1 mg.L⁻¹. Regarding diclofenac, (Nam *et al.*, 2014) have shown a better fit with the Freundlich model while, for carbamazepine, (Yu *et al.*, 2008) have shown better results with the Freundlich model, for an equilibrium concentration of 10-800 ng.L⁻¹. Over the years, a wide variety of equilibrium isotherm models have been developed while considering various assumptions

(heterogeneity of the surface energy, multilayer adsorption, etc.). These isotherm models can be classified as two-parameter (Langmuir, Freundlich, Temkin, Elovich, Dubinin-Radushkevich) or threeparameter models (Redlich-Peterson, Toth, Langmuir-Freundlich, Sips, Radke-Prausnitz). (Limousin *et al.*, 2007; Rouquerol *et al.*, 1999; Ruthven, 1984; Worch, 2012).

In combination with the selection of the relevant model, various options are possible to determine the model parameters: linearization of the equations with a simple least-square regression or a nonlinear regression method. This latter approach has generally been preferred since it provides the most accurate description of experimental data (Foo and Hameed, 2010). Moreover, some models cannot be linearized and the determination of parameters is necessarily based on minimizing an error function, which corresponds to the deviation between the experimental data and the predicted value from the model. The optimized parameters for the isotherm model thus strongly depend on the selected error function. It should be noted that the choice of the error function influences the accuracy of the modeled data, since this function promotes a better fit of isotherms at low or high concentration with different weights for the experimental data.

Recent studies have compared several error functions in combination with the usual isotherm models (Allen *et al.*, 2003; Chan *et al.*, 2012; Foo and Hameed, 2010). Normally, error functions are used to minimize or maximize the error distribution between the experimental equilibrium data and the predicted isotherms, according to the definition of the error function. Although the method of least squares is one of the most widely used techniques with the maximum coefficient of determination, r², some studies have pointed out that other error functions are more relevant, such as the hybrid fractional error function (HYBRID), Marquardt's percent standard of deviation (MPSD), the average relative error (ARE), the sum of the absolute errors (EABS) etc. According to Chan *et al.* (2012), the hybrid error function provided the best overall results. For this study, the Sips model performed the best prediction only if the model was adjusted using the hybrid method with the experimental data.

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Most investigations have been carried out for the adsorption of organic compounds at high concentrations (at the mg.L⁻¹ scale) (Sotelo *et al.*, 2014). Only recently, research programs have focused on the modeling of adsorption isotherms at lower concentrations (μ g.L⁻¹ or less) with the development of analytical tools (Al Mardini and Legube, 2010; Matsui *et al.*, 2003). This trend is more consistent with the emerging pollutants encountered in the environment and should be generalized to achieve a better understanding of the adsorption process used for the production of drinking water. However, at trace concentrations, experimental results are more difficult and more expensive to obtain. In order to avoid this problem, the extrapolation of the isotherm model from high to low concentrations would be an interesting option. In fact, if the model is fitted at high concentrations, the prediction outside the range of measurements (to trace contents) could give relevant results. As previously mentioned, the selection of the isotherm model and the error function plays a key role in predicting adsorption capacities.

This study focused on the relevance and reliability of the extrapolation of adsorption isotherms from high to low concentrations, with the aim of evaluating the ability of numerous isotherm model/error function pairs to achieve this change of scale. For this purpose, three emerging pollutants were chosen and their isotherms of adsorption were determined for one ACFC. Two sets of experiments were carried out at low and high concentrations. Then, the combinations between 13 models and 8 error functions were evaluated.

2. Materials and methods

2.1. Materials

2.1.1. Organic compounds

The targeted compounds – acetaminophen, carbamazepine and diclofenac - were all purchased from Sigma-Aldrich (purity > 98 %). The physicochemical properties and molecular structures of the

compounds are listed in Table **1**. Stock solutions (10 mg.L⁻¹) were prepared by dissolving the commercial standard in ultrapure water (UPW) provided by an ElgaPureLab System (18.2 M Ω .cm). Two sets of experiments were carried out. The first was conducted at high concentrations of pollutant (0.1 – 20 mg.L⁻¹). The second set was achieved at low concentrations (0.1 – 2 µg.L⁻¹), which is the range closer to the levels observed in surface waters (Petrie *et al.*, 2014).

For the high range of concentrations, analyses were performed by Ultra Performance Liquid Chromatography (UPLC) using a Waters ACQUITY H-class system (Waters Assoc., Milford, MA) equipped with a photodiode array detector (PDA). Detection was carried out at 243, 272, 276 and 280 nm with a PDA $e\lambda$ detector (Waters) for acetaminophen, diclofenac and carbamazepine, respectively. 5 µL of sample was injected onto a BEH C18 column (100 x 2.1 mm x 1.7 µm, Waters) thermostated at 35 °C. The binary gradient consisted of a mixture of acetonitrile as mobile phase A, and acetonitrile/water/formic acid (10:90:0.1, v/v/v) as mobile phase B. The separation was initiated at a constant flow of 0.4 ml/min with 90 % B for 1 min, followed by a decrease in B to 10 % within 7 min. This composition was then maintained for 10 min and returned to the initial composition.

For the low range of concentrations, analyses of the selected compounds were performed using UPLC with an Acquity system (Waters) coupled with a tandem mass spectrometer (Quattro Premier, Micromass). The chromatographic system included a 2777 autosampler (Waters) equipped for dual on-line solid phase extraction (on-line SPE) with HLB cartridges. 5 mL of sample was loaded onto the HLB column using a large volume injection loop and a quaternary solvent pump (QSM – Waters). After the loading and cleaning steps, HLB cartridges were connected to the analytical hydraulic circuit. Chromatographic conditions were similar to those described previously.

Mass spectrometry was used with an electrospray ionization source in positive mode with a capillary voltage of 3 kV and nitrogen as the nebulizer and drying gas. The cone gas flow and the desolvation gas flow were set at 50 L h^{-1} and 750 L h^{-1} , respectively. The source temperature and desolvation gas

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temperature were 120 °C and 350 °C, respectively. The multiple reaction monitoring (MRM) mode was used for the quantification of all compounds. Retention time, MRM transitions, cone voltage and collision cell energy are summarized in Table 1.

Compound	Acetaminophen	Carbamazepine	Diclofenac
N° CAS	103-90-2	298-46-4	15307-79-6
Structure	HO	O NH2	CI NH CI OH
MW (g.mol⁻¹)	151.16	236.27	318.13
log Kow	0.46	2.45	4.51
рКа	9.5	14	4.15
Solubility (g.L⁻¹)	14.9	0.02	50.0
Size (A)	7*5*2	10*8*5	10*8*1
Cone Voltage (V)	25	28	22
Collision energy (eV)	19	19	25
Transition	152>110	237.1>194	296.1>250
Retention time (min)	1.49	3.61	4.73

Table 1 : Properties of pharmaceutical residues used in adsorption procedures

2.1.2. Adsorbent

The ACFC was supplied by Dacarb (Asnières-sur-Seine, France). It was soaked as received in ultrapure water to remove any dissolved contaminants and/or fine particles and then dried at 120 °C prior to the experiment. The physical and chemical characteristics of the ACFC are given in Table **2**. The physisorption of nitrogen at 77 K (Autosorb, Quantachrome Instruments) was used to determine the specific surface area (multi-point BET method in the range of relative pressures 0.01 to 0.1), the total pore volume (relative pressure of 0.995), the micropore volume and the average pore size (Quenched Solid State Functional Theory, QSDFT) (Condon, 2006; Rouquerol *et al.*, 1994). In addition, the Boehm titration method was used to quantify the functional surface groups (Boehm, 1994).

The ACFC possesses a highly microporous texture (over 90 % of the total pore volume) and few functional groups in comparison to common granular activated carbons (Lopez-Ramon *et al.*, 1999). This specificity was confirmed by the residual pH (measured after 24 h in ultrapure water), which was close to neutrality and in agreement with the balance between acidic and basic groups. This information is summarized in Table 2.

Table 2 : Physical and chemical properties of the ACFC

Textural properties from N₂ adsorption at 77 K					
Specific surface area (m ² .g ⁻¹)	1615				
Total pore volume (cm ³ .g ⁻¹)	1.02				
Micropore volume (QSDFT) (cm ³ .g ⁻¹)	0.92				
Micropore size (QSDFT) (nm)	< 0.48				
Chemical surface properties	s from Boehm titration				
Residual pH	6.14				
Basic functions (µeq.L ⁻¹)	286				
Carboxylic acid functions (µeq.L ⁻¹)	12				
Lactone functions (µeq.L ⁻¹)	87				
Phenolic functions (μeq.L ⁻¹)	377				

2.2. Adsorption isotherm curves

Kinetics and adsorption isotherms were performed for single components and multi-components in UPW. For the set of experiments at high concentration, each reactor contained 1 L of the aqueous solution with the single targeted compound. Initial concentrations (C₀) ranged from 0.1 mg.L⁻¹ to 20 mg.L⁻¹. On the other hand, 10 L reactors were used for the second set of experiments and the concentrations ranged from 100 ng.L⁻¹ to 2 μ g.L⁻¹ (multi-components). Preliminary experiments demonstrated that inter-component adsorption competition was not significant if the initial concentration was lower than 10 μ g.L⁻¹.

The pH was maintained at 7.5 using hydrogen carbonate buffer and did not vary more than 0.1 pH unit between the initial and final time of the experiment. A constant mass of ACFC (20 mg \pm 0.2) was then incorporated and the reactors were tightly sealed and shaken at 400 rpm. A constant temperature of 25 °C was achieved using a thermostatic bath. Preliminary kinetic data indicated that, for all cases, the adsorption equilibrium was reached within 10 days. After equilibration, the

supernatants were filtered using a 0.2 μ m GHP Acrodisc filter prior to analysis by SPE-UPLC/ MS-MS in order to determine the residual concentration (C_e) and adsorption capacities (Q_e) according to the mass balance.

2.3. Isotherm models

As previously mentioned, numerous models have been formulated to describe adsorption equilibria, considering various assumptions to represent the adsorption phenomenon. Table 3 presents the most common equations for adsorption in the aqueous phase. An ideal model should have four properties: it must be effective, comprehensive, realistic and predictive (Worch, 2012). The choice of the isotherm model is important to understand the state of the adsorbed phase, the interactions between the adsorbent surface and the adsorbed molecules, and the interactions between the adsorbed molecules.

The Freundlich model is widely used for organic pollutants. However, this isotherm model has no limitation to the adsorption capacity at the highest liquid-phase concentrations. On the contrary, the Langmuir isotherm model follows a Henry-type equation at low concentrations and a saturation limit at high concentrations. According to this model, adsorption takes place at specific homogeneous sites within the adsorbent, and once a molecule occupies a site, no further adsorption can take place onto this site. The Langmuir equation can be expressed in its general form (L) or with linearized relationships (L1-L5). Elovich's model is based on kinetic development and supposes a multi-layer adsorption. Three-parameter models have been derived from the Langmuir and Freundlich models. The Sips equation is a combination of these two isotherm equations. At low concentrations, it effectively reduces to a Freundlich isotherm and thus does not follow Henry's law. At high concentrations, the Sips model predicts a monolayer sorption capacity, which is a characteristic similar to the Langmuir isotherm (Foo and Hameed, 2010). The Toth equation is another empirical model developed to improve Langmuir isotherm fittings and useful for describing heterogeneous adsorption isotherm systems, which satisfy both the low- and high-end boundary of the concentration. Adsorption sites are considered to possess low energies.

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Isotherm	Number of parameters	Mathematical expression	Reference
Freundlich (F)	2	$q_e = K_F \cdot C_e^n$	(Freundlich, 1906)
Langmuir (L)	2	$\frac{q_e}{q_e} = \frac{bC_e}{1+bC_e}$	(Langmuir, 1915)
Langmuir 1 (L1)	2	$\frac{1}{a} = \frac{1}{a} \frac{1}{a} + \frac{1}{a}$	(Langmuir, 1915)
Langmuir 2 (L2)	2	$\frac{q_e}{q_m} = \frac{bC_e}{1+bC_e}$ $\frac{1}{q_e} = \frac{1}{C_e}\frac{1}{q_mb} + \frac{1}{q_m}$ $\frac{C_e}{q_e} = C_e\frac{1}{q_m} + \frac{1}{q_mb}$	(Langmuir, 1915)
Langmuir 3 (L3)	2	$q_e = \frac{1}{b} \frac{q_e}{C_e} + q_m$	(Langmuir, 1915)
Langmuir 4 (L4)	2	$\frac{q_e}{C_e} = -bq_e + bq_m$	(Langmuir, 1915)
Langmuir 5 (L5)	2	$\frac{q_e}{C_e} = -bq_e + bq_m$ $\frac{1}{C_e} = bq_m \frac{1}{q_e} - b$	(Langmuir, 1915)
Elovich (Ev)	2	$\frac{q_e}{q_e} = b_e C_e \exp\left(\frac{q_e}{q_e}\right)$	(Elovich and Larinov, 1962)
Langmuir-Freundlich (L-F)	3	$\frac{q_e}{q_m} = b_e C_e \exp\left(\frac{q_e}{q_m}\right)$ $\frac{q_e}{q_m} = \frac{(bC_e)^n}{1 + (bC_e)^n}$	(Sips, 1948)
Sips (S)	3	$\frac{q_{\rm m}}{q_{\rm m}} = \frac{bC_{\rm e}^{n}}{1+bC_{\rm e}^{n}}$	(Sips, 1948)
Linearized Langmuir- Freundlich (LLF)	3	$q_{m} = \frac{q_{m}}{q_{m}}$ $q_{e} = \frac{q_{m}}{\frac{1}{1} + \frac{1}{1}}$	(Limousin et al., 2007)
Tóth	3	$q_{e} = \frac{q_{m}}{\frac{1}{bC_{e}^{n}} + \frac{1}{b}}$ $\frac{q_{e}}{q_{m}} = \frac{bC_{e}}{(1 + (bC_{e})^{n})^{1/n}}$	(Toth, 1971)
Error function name		Mathematical expression	Example of use
Sum of the squared residual	RSS	$\sum_{i=1}^{n} (q_e \cdot \hat{q}_e)_i^2$	(Kumar, 2006)
Relative error	E	$E = \frac{1}{n} \sum_{i=1}^{n} \frac{q_{e,i \text{ mod}} - q_{e,i \text{ exp}}}{q_{e,i \text{ exp}}}$	
Average of absolute error	EABS	$\frac{1}{n}\sum_{i=1}^{n} q_e - q_{e,avg} _i$	(Ng et al., 2003)
Average relative error	ARE	$\frac{1}{n}\sum_{i=1}^{n} \left(\frac{q_e \cdot \hat{q}}{q_e}\right)_i$	(Kapoor and Yang, 1989)
Standard deviation of relative errors	S _{RE}	$\sqrt{\frac{\sum_{i=1}^{n} \left(\left(\frac{q_e}{q_e} - \hat{q}_i\right)_i - ARE \right)_i^2}{n-1}}$	(Boulinguiez <i>et al.,</i> 2008)
Chi-square	χ²	$\sum_{i=1}^{n} \left(\frac{q_e - q_{e,avg}}{q_e} \right)_i^2$	(Boulinguiez <i>et al.,</i> 2008)
Marquardt's percent standard deviation	MPSD	$\sqrt{\frac{1}{n-p}\sum_{i=1}^{n} \left(\frac{q_e - q_{e,avg}}{q_e}\right)_i^2}$	(Marquardt, 1963)
	Hybrid	$\frac{1}{1} \sum_{n=1}^{n} (q_e - q_{e,avg})^2$	(Ng et al., 2003)

Table 3. Adsorption models and error functions for the adjustment of the isotherm adsorption model

2.4. Error functions

The parameters of each isotherm model then have to be adjusted to fit the experimental data (Limousin et al., 2007). Initially, the values of each parameter were guessed and predicted isotherm data $q_{e,avg}$ were computed. Assessment of the goodness-of-fit was then discussed in terms of various error functions. Searching for the best fitted adsorption isotherm using the method of least squares (RSS) is the most widely used technique to predict the optimum isotherm. However, at higher concentration ranges, squares of errors tend to increase, so that this range of concentration has more weight on the fitting procedure than the lower values of concentration/capacities (Foo and Hameed, 2010). As a consequence, for extrapolation towards trace concentrations, the RSS error function is probably inappropriate and the prediction of low values of the isotherm from data obtained at high concentration requires alternative error functions. The hybrid function was developed to improve the RSS fit at low concentrations. Average relative error (ARE) attempts to minimize the fractional error distribution over the entire concentration range. The sum of absolute errors (EABS) approach provides a better fit compared to the RSS model although this function also favors the highest concentrations. The coefficient of determination (r^2) represents the percentage of variability in the dependent variable. Standard deviations of relative errors (s_{RE}) are individually determined to evaluate the overall correlation and the dispersion of its relative errors (Foo and Hameed, 2010). Marquardt's percent standard deviation (MPSD) is similar to a geometric mean error distribution modified according to the number of degrees of freedom of the system. These models are presented in Table 3.

Previously, linear regressions were used to fit isotherms. This procedure was easily applied to many experimental data and the equations were quite simple (Ayoob and Gupta, 2008). However, more recently, many researchers have combined isotherm models with different error functions to show the applicability of linear or non-linear isotherm models to describe the adsorption mechanism. Table 4 summarizes, for adsorption onto activated carbon only, the different combinations of isotherm model/error function used to depict the adsorption of organic and inorganic pollutants.

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Type of activated carbon	Adsorbate	Isotherm models	Error functions	Linear/non -linear	Reference
Activated carbon	Tetrahydrothi(o) phene	Langmuir	r², EABS, RSS, ARE, s _{RE} , Hybrid	Non-linear	(Boulinguiez <i>et al.,</i> 2008)
Activated carbon	Methylene blue	Freundlich, Langmuir, Redlich- Peterson	r², ARE, Hybrid, MPSD, EABS	Non-linear	(Kumar <i>et</i> <i>al.,</i> 2008a)
Activated carbon	Basic red 9	Freundlich, Langmuir, Redlich- Peterson	r², EABS, Hybrid, ARE, SAE, MPSD	Non-linear	(Kumar <i>et</i> <i>al.,</i> 2008b)
Activated carbon	Basic blue 9 dye	Freundlich, Langmuir, Redlich- Peterson	r², χ²	Non-linear	(Jumasiah e <i>t</i> <i>al.,</i> 2005)
Activated carbon	Acid dyes	Freundlich, Langmuir, Sips, Redlich-Peterson	Hybrid, MPSD, ARE, EABS	Linear and non-linear	(Chan <i>et al.,</i> 2012)

Table 4. Some uses of isotherm adsorption models with different error functions

2.5. Adjustment procedure, validation and extrapolation

Optimization was achieved, using the Microsoft Excel solver, for all the combinations of isotherm models with error functions (Table 3). However, the accuracy of the prediction was difficult to compare since the values for the error functions could not be compared with each other. Therefore, in order to compare the ability of the "isotherm model/error function" couples to describe and predict experimental data, the sum of relative error (E) was used as the comparative variable. The methodology is illustrated in Figure 1. First, the parameters of each isotherm model were optimized for the full range of concentrations (FR) in accordance with the error functions. These parameters were then used to determine two average relative errors (E). E_{FR} was related to the average relative error for all experimental data, whereas E_{FR-LC} corresponded to the error of the model for the set of low concentrations based on the FR optimization of parameters. On the other hand, the isotherm parameters were adjusted to the set of high concentrations and the corresponding E_{HC} was calculated. Finally, the accuracy of the extrapolation was determined from E_{HC-LC} , which was the average relative error for the low range of concentrations with parameters determined at high concentrations (Figure 1) (Assoumani *et al.*, 2009).

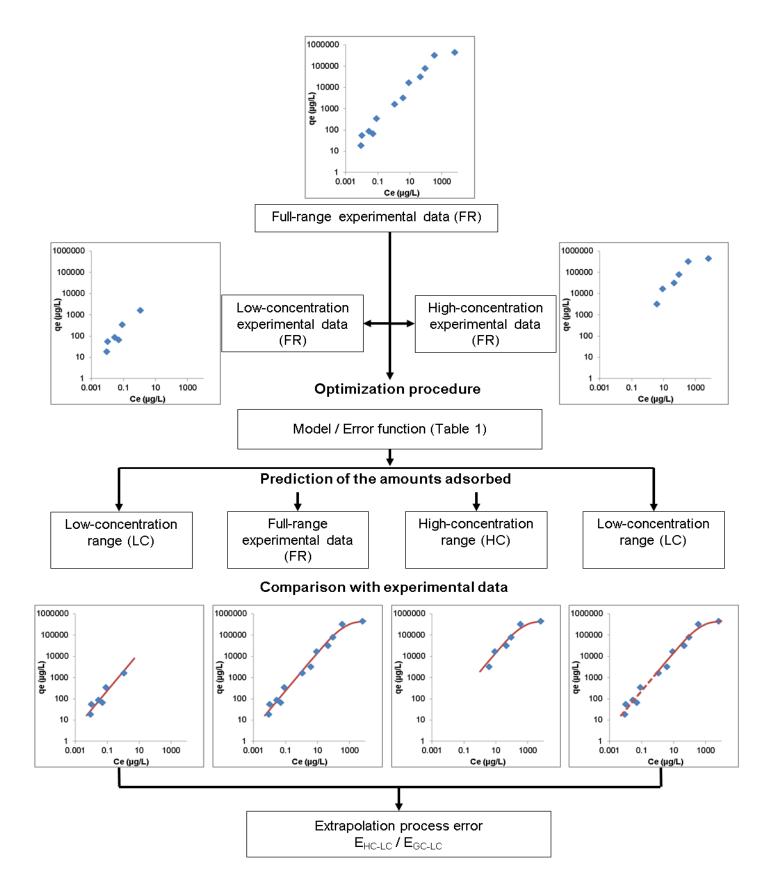


Figure 1. Methodology for the optimization procedure and the extrapolation of the model to the low-range concentrations

3. Results and discussion

3.1 Kinetics of adsorption

Kinetic studies are essential to determine the contact time at which the adsorption equilibrium is reached. Moreover, the change in aqueous concentration against time was used to compute the mass transfer coefficients for each pollutant/adsorbent system. Assuming that the adsorbent particles are spherical and of unique size, (Matthews and Weber, 1977) proposed a kinetic model based on the fact that the overall adsorption is kinetically limited by external and diffusion steps. This homogeneous surface diffusion model (HSDM) supposes that the diffusion has two major limiting steps: diffusion through the external film surrounding the particle and surface diffusion inside the porosity. Thus, two coefficients were evaluated (i) Kf, which represents the external mass transfer, and (ii) Ds, which corresponds to the surface diffusivity. For each adsorbate/adsorbent couple studied, these two mass transfer coefficients were adjusted at high and low concentrations (Table 5) using an iterative optimization procedure (Traegner and Suidan, 1989). Figure **2** shows the kinetics of adsorption for two ranges of concentration, starting at 10 mg.L⁻¹ or 10 µg.L⁻¹. The equilibrium times were highly dependent on the initial concentration whereas the nature of the compounds did not influence the equilibration time. For the lowest concentration, a contact time of 10 days was required to reach equilibrium and this decreased to 50 h when the initial concentration was 10 mg.L⁻¹

1.

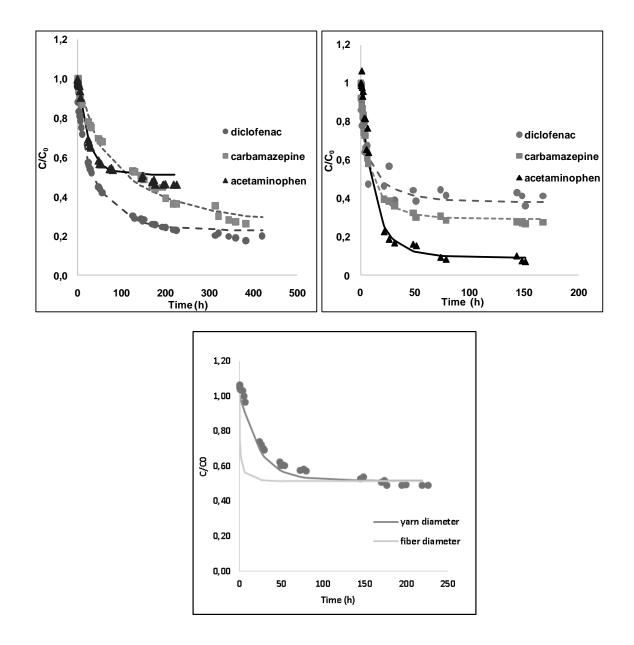


Figure 2. Kinetic curves of adsorption obtained at (a) high concentration ($C_0 = 10 \text{ mg/L}$) and (b) low concentration ($C_0 = 10 \mu \text{g/L}$). Solid line corresponds to the HSDM model.

For conventional adsorbents, such as powders or grains of activated carbon, the average particle size is commonly determined by physical analysis (optical granulometry or sieving). However, these approaches are not relevant for ACFC, so the particle size was determined using image analysis by scanning electron microscopy. The images obtained show that ACFC KIP-1200 is constituted of 430- μ m yarns, which are themselves made of fibers (10- μ m average diameter). As established by Mathews & Weber (1977), the HSDM was not appropriate to take into account this dual geometry. Both scales, yarn and fiber, were considered for the adjustment of the mass transfer coefficients (Kf and Ds). Using the fiber diameter as the elementary particle size resulted in large differences between the modeled and experimental values. On the contrary, when considering the yarn diameter, the adjustment was relevant (Figure 2). Therefore, the external resistance to the mass transfer occurred in the film around the yarn while the flow between the fibers was neglected. Consequently, the adjusted Ds coefficients combined the diffusions inside the porosity and through the liquid film inside the yarn, between the fibers.

		C ₀ = 10 mg.L ⁻¹		C ₀ = 10 μg.L ⁻¹				
	Ds (m².s⁻¹)	Kf (m.s⁻¹)	Biot number	Ds (m².s⁻¹)	Kf (m.s⁻¹)	Biot number		
Diclofenac	6.1×10 ⁻¹⁴	2.0×10 ⁻⁴	21.8	1.3×10 ⁻¹³	2.8×10 ⁻³	53.3		
Acetaminophen	2.1×10 ⁻¹³	2.3×10 ⁻⁵	2.5	1.2×10 ⁻¹³	4.1×10 ⁻⁴	1.5		
Carbamazepine	3.5×10 ⁻¹⁴	1.3×10 ⁻⁵	6.1	1.4×10 ⁻¹³	7.4×10 ⁻⁴	6.6		

Table 5. Mass transfer coefficients at different intital concentrations (C₀) of micropollutants

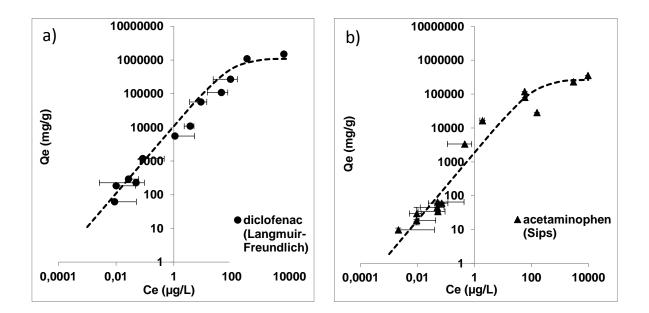
In Table 5, the Biot numbers (ranging from 1.5 to 53.3) highlight the preponderance of both surface diffusion and external mass transfer as the limiting steps for mass transfer. The coefficients of external transfer, Kf, described the diffusion molecules through the boundary layer around the yarn. Furthermore, these external coefficients were shown to be solely influenced by the range of concentration, and independent of the type of adsorbed compound. At the highest concentration, Kf ranged from 1.3×10^{-5} to 2.0×10^{-4} m.s⁻¹. At trace concentrations, this coefficient increased to between 4.1×10^{-4} and 2.8×10^{-3} m.s⁻¹.

Ds values, obtained at the high concentration, varied by three orders of magnitude, $3.1 \times 10^{-15} \text{ m}^2 \text{.s}^{-1}$, for diclofenac. In comparison, at the low concentration, the surface diffusivities were similar for all compounds and only varied by one order of magnitude, from 4.4×10^{-14} to $1.4 \times 10^{-13} \text{ m}^2 \text{.s}^{-1}$. For granular activated carbon, (Baup *et al.*, 2000) have shown Ds coefficients equal to $10^{-14} \text{ m}^2 \text{.s}^{-1}$, for an initial concentration of 0.5 µg.L⁻¹. (Al Mardini and Legube, 2009) have reported some coefficients of $10^{-16} \text{ m}^2 \text{.s}^{-1}$ for powdered activated carbon, with an initial concentration of 1 µg.L⁻¹.

Two types of behavior were thus observed. On one hand, for carbamazepine, a significant decrease in Ds coefficients was seen as the bulk concentration decreased. On the other hand, no significant changes were observed for acetaminophen and diclofenac with the initial concentration. The steric effect could not explain these different behaviors. For instance, the molecular weight of diclofenac (M = 318.1 g.mol⁻¹) is half that of acetaminophen (M = 151.2 g.mol⁻¹) but these compounds exhibited similar surface diffusivities. Dipole moment, molecular volume or polarizability could explain these different behaviors.

3.2 Modeling adsorption isotherms

Isotherm curves of adsorption were obtained for each compound independently in ultra-pure water, without competition with organic matter. Every isotherm was modeled using the adsorption models mentioned in Table 3. The isotherm parameters were adjusted by minimizing the error functions (Table 3). Figure 3 presents the best "isotherm/error function" couple determined over the entire range of concentrations for each compound.



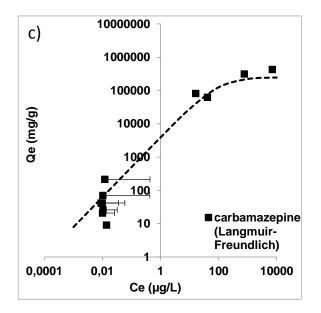


Figure 3. Adsorption isotherms of diclofenac, acetaminophen and carbamazepine on ACFC KIP-1200

In Figure 3, at the highest concentrations (above 10 ng.L⁻¹), the adsorption capacities were similar for all compounds (between 10 and 1000 µg.g⁻¹). On the contrary, when the concentration decreased, the adsorption capacities were significantly influenced by the nature of the adsorbed molecule. Likewise, different shapes of the adsorption isotherm curves were observed. The organic compounds had an "L-shape" isotherm. Each type of isotherm was related to different adsorption mechanisms. The concave "L-shape" isotherms were the most common. These experimental data were described by the Freundlich or the Langmuir model (Limousin *et al.*, 2007).

For the highest concentrations, experimental errors were quite low. For instance, uncertainties were less than 5 % for diclofenac. However, at the lowest concentrations, values were more dependent on the compound. For diclofenac, some experimental errors of 72 % were observed, whereas for acetaminophen, they were only 20 %. An explication could come from the dispersion of points. For acetaminophen, values are relatively close to each other. On the contrary, for diclofenac, the dispersion is quite important.

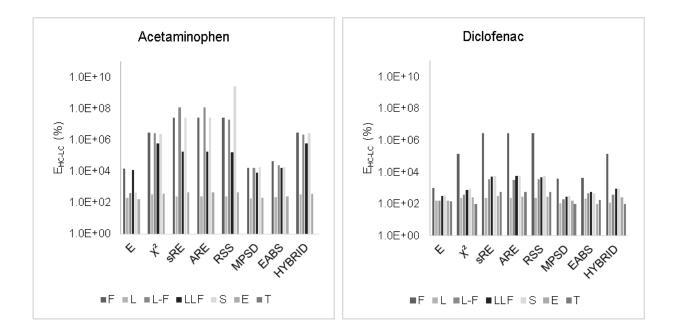
The best descriptors for each adsorption isotherm are presented in Table 6.

Table 6. Best descriptors	for each adsorption	isotherm equation

	Full range of con	centration	High range of concentration			
Compound	Isotherm model	Average E (%)	Isotherm model	E (%)		
Diclofenac	Langmuir-Freundlich/E	55.5	Toth/E	13.9		
Carbamazepine	Langmuir-Freundlich/E	42.4	Elovich/MPSD	17.3		
Acetaminophen	Sips/E	40.9	Elovich/E	66.6		

3.3 Extrapolation of adsorption capacities

The influences of the isotherm model, error function, and nature of the organic compound were significant with average relative errors E_{HC-LC} ranging from 60 to 10^{11} %. For extrapolation (using the procedure described in Figure 1) from high concentration parameters, relative errors were mostly over 100 % for the three compounds (Figure 4).



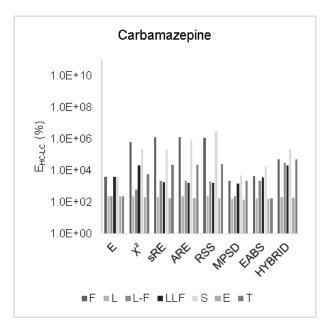


Figure 4. Extrapolation errors E_{HC-LC} from high concentration parameters (F: Freundlich; L: Langmuir; L-F: Langmuir-Freundlich; LLF: Linearized Langmuir-Freundlich; S: Sips; E: Elovich; T: Toth)

For diclofenac, the best extrapolations were obtained using the E error. Langmuir-Freundlich and Sips models led to the best extrapolation from high to low concentrations with E_{HC+LC} errors from 67 to 134 %, respectively. In a recent study, Nam *et al.* (2014) demonstrated that the Freundlich model was the most appropriate to describe the adsorption of diclofenac at low concentrations. For this study, the authors carried out diclofenac adsorption from 20 to 500 ng.L⁻¹, with powdered activated carbon (1 mg.L⁻¹). They showed that adsorption parameters were well adjusted by the Freundlich isotherm for hydrophobic compounds. However, in this study, Freundlich models, with all error functions, led to significant deviations in terms of the extrapolation. Errors were widely influenced by the high range of concentrations. The Freundlich isotherm can describe neither the linear range at very low concentrations nor the saturation effect at high concentrations (Worch, 2012).

Moreover, the linearized Langmuir-Freundlich model did not provide a better extrapolation than the non-linear Langmuir-Freundlich model. (Choi *et al.*, 2006) have shown that linear models have a poor fit for hydrophobic compounds. This conclusion was thus confirmed.

In the case of carbamazepine, Langmuir and Elovich models provided an acceptable extrapolation with an error of about 100 %. Moreover, the error function did not play a significant role in the accuracy of the extrapolation. On the contrary, the selected error function greatly influenced the accuracy of extrapolation when the Toth or Langmuir-Freundlich models were used. With these models, acceptable results were obtained with the E error (about 100 %) but revealed a large deviation (over 1000 %) when the hybrid error function was considered.

According to the results obtained for carbamazepine, the Freundlich model appeared to be the worst choice to predict adsorption capacities at trace concentration from values obtained at high concentrations. However, this model is not so inaccurate for fitting experimental data at high concentrations or for the full range of concentrations. Like for diclofenac, linearization of the Langmuir-Freundlich model did not improve its extrapolation ability.

Similar conclusions were obtained with acetaminophen. Langmuir and Elovich models led to the most accurate predictions irrespective of the error function. However, if all models are considered, E, MPSD and EABS errors gave the best extrapolations. Except for Langmuir and Elovich, RSS and ARE had errors of extrapolation higher than 10⁵ %. Unlike diclofenac and carbamazepine, the linearized Langmuir-Freundlich model gave the best extrapolation. This result is in agreement with Kumar *et al.* (2008) and Nam *et al.* (2014) who observed that acetaminophen could be fitted with precision by a linear isotherm.

For all organic compounds, for two-parameter isotherm models, E and MPSD functions were found to be a good option to minimize the error distribution between the experimental and predicted adsorption capacities. Kumar *et al.* (2008) obtained similar results for their predicted isotherms at high concentrations. Three-parameter models were not better than two-parameter models during extrapolation. This observation is not true for fitting; the addition of parameters decreased the error (Table 6). For three-parameter equations, E and EABS were found to be the best functions in minimizing the error distribution. More generally, for all studied molecules and isotherm models, the

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error function E was highlighted as the most relevant, although significant deviations were computed in some cases. If all molecules are considered, the Langmuir/MPSD, Elovich/E, Langmuir/E, Elovich/MPSD and Langmuir-Freundlich/E couples could be used to extrapolate the models to the lowest residual concentrations. These relationships were the most efficient. For instance, Elovich/E exhibited relative errors of 66 %, 113 % and 96 % for acetaminophen, diclofenac and carbamazepine, respectively.

In order to extend the discussion, the errors of extrapolations from the high range of concentrations $(E_{HC-LC}, previously shown in Figure 4)$ were compared to those determined for the full range of concentrations (E_{FR-LC}) . The results are presented in Figure 2. Thus, if the ratio E_{HC-LC}/E_{FR-LC} approaches 1, it means that there is no additional error caused by the extrapolation procedure and the model could be perfectly fitted at the highest concentrations and then extrapolated to the lowest concentrations. In these cases, the determination of isotherm parameters based only on the high concentration is relevant to predict equilibrium at trace concentration.

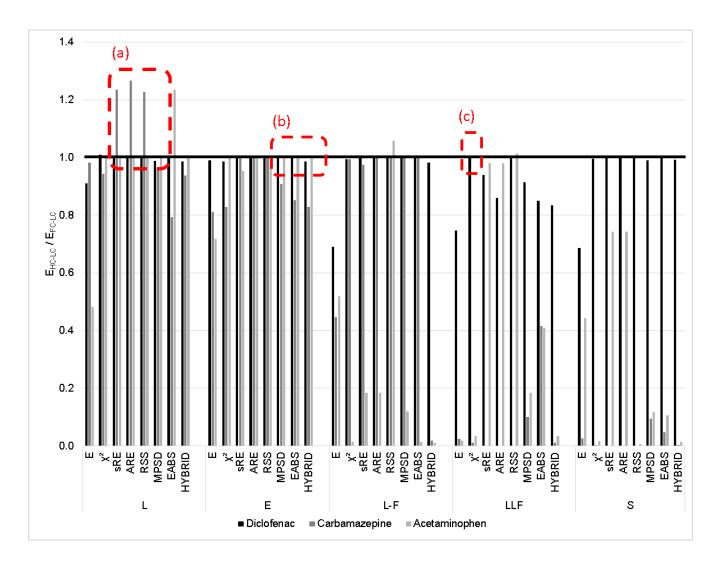


Figure 5. Evolution of the error function calculated from extrapolation of adsorption curves at high and full range of concentrations

In Figure 5, three zones are highlighted (a, b and c), which show the best "model/error function" combinations for all organic molecules. For these couples, the ratio of errors was close to 1, meaning that no additional error was added by the extrapolation from high concentrations. Namely, the Langmuir equation could be associated with the S_{RE} , ARE, RSS and MPSD error functions, the Elovich equation with S_{RE} , ARE, RSS and MPSD and the Langmuir-Freundlich model with RSS.

It should be noted that, from Figure 4, the Langmuir/MPSD, Elovich/E, Langmuir/E, Elovich/MPSD and Langmuir-Freundlich/E couples were highlighted for their absolute accuracy of extrapolation.

Therefore, considering the results presented in Figure 5, it was concluded that, for all organic compounds, the Langmuir/MPSD couple was the best choice for extrapolation. In fact, the extrapolation error was among the smallest and extrapolation from the high concentrations did not add significant errors compared to the adjustment over the entire range of concentrations.

Compound	lsotherm model	Full range of concentration		High range of concentration			Low range of concentration			
Diclofenac	Langmuir- Freundlich/E	Qm	b	n	Q _m	b	n	Qm	b	n
Carbamazepine	Langmuir-	1525.8	1.0	1.3	1611	0.7	1.1	1670.4	16.8	2.1
Curbumazepine	Freundlich/E	q _m 521.1	b 0.1	n 0.7	q _m 568.1	b 2.3	n 0.9	q _m 1954.9	b 0.6	n 0.9
Acetaminophen	Sips/E	q _m	b	n	q _m	b	n	q _m	b	n
		4149.1	0.1	0.8	2893.9	0.1	0.9	165.0	0.1	0.5

Table 7. Isotherm parameters for the best model

Units are based on liquid-phase concentration expressed in µg.L⁻¹ and amounts adsorbed in µg.g⁻¹

As shown in Table 7, isotherm parameters have different behaviors according to the compound and range of concentration. The "n" parameter is little impacted by the extrapolation of adsorption capacities from high to low concentrations. Except for diclofenac, the change in concentration has no influence on the "b" parameter.

For diclofenac, which presents the lowest extrapolation errors, the q_m coefficients are not impacted by extrapolation. For the other compounds, this parameter is influenced by the range of concentration.

4. Conclusion

In this study, we investigated the sorption of three pharmaceutical compounds onto activated carbon fiber cloth. Adsorption kinetic coefficients were little impacted by the initial concentration for diclofenac and acetaminophen. For a residual concentration of 0.1 μ g.L⁻¹, adsorption capacities were 116, 134 and 248 μ g.g⁻¹ for diclofenac, acetaminophen and carbamazepine, respectively.

The ability of several isotherm models associated with different error functions to extrapolate adsorption capacities from high to low concentrations was evaluated.

Among the 104 combinations, for diclofenac, carbamazepine and acetaminophen, which present the usual isotherm shape, the Langmuir-Freundlich model adjusted with the MPSD error enabled the best extrapolation from $mg.L^{-1}$ to $\mu g.L^{-1}$.

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