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# The use of linear and nonlinear methods for adsorption isotherm optimization of basic green 4-dye onto sawdust-based activated carbon

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#### Abstract

The abundance and low cost of sawdust as an agricultural by-product make it a potential candidate as precursor of activated carbon preparation. In this study, the sawdust samples were chemically treated with acid (20% H<sub>3</sub>PO<sub>4</sub>,170°C, 2h), base (20%KOH, ambient, 24h) and salt (1mol/L (NH<sub>4</sub>)<sub>2</sub>S<sub>2</sub>O<sub>8</sub>, ambient, 24h) separately followed by pyrolysis at 600 C for 3h, resulting respectively in SD-1, SD-2, SD-3 sawdust-based activated carbons respectively used as adsorbents for basic green 4 dye removal from aqueous media and the results were compared to the commercially available Merck activated carbon. Batch adsorption tests were performed and the experimental data analyzed using the Langmuir and Freundlich models in their linear and non linear forms, in order to estimate the equilibrium parameters. For optimum adsorption isotherm selection, a comparison of linear and nonlinear regression methods was applied. Three errors functions: Chi-square statistic ( $\chi^2$ ), root mean square error (RMSE) and Average percentage error (APE) were used for isotherm optimization prediction. While the coefficient of determination ( $R^2$ ) was used for the best-fit linear theoretical isotherm selection

# 1. Introduction

Effluents from industrial, agricultural and domestic origin are often charged with pollutants. Their impact on the environment is very harmful; they are sometimes recycled or quite simply rejected into nature which causes a capital problem and a major concern for local public authorities and international organizations [1-3]. Thus, all these encouraged the improvement of the existing techniques of depollution and the development of new processes, in accordance with the increasingly restrictive international standards. Textile and tannery are among major industries that use great amount of water, generating then an important waste water pollution charged with all kind of pollutant such as dyes, salts etc...[4-6]. Indeed, many conventional methods have been used for wastewater treatment such as precipitation, oxidation, flotation-coagulation and electro-coagulation. Even they appear effective, but they are limited to a variety of pollutants for technical reasons and high cost of exploitation or may not be capable of treating large volumes of effluent [7-9]

Among the mentioned methods for wastewater treatments that has drawn attention to many researchers in the last decades, adsorption using activated carbon, a phase transfer process has been widely used in practice to remove contaminants in all their forms (organic and inorganic) from fluid phases [10-13] because of the low investment in initial cost and design simplicity. Activated carbons for wastewater treatment are usually obtained from materials locally available such as natural materials, agricultural wastes and industrial wastes [14-16].

Adsorption isotherms which are usually used to determine the quantity of a given pollutant adsorbed are also used to describe adsorbent-adsorbate equilibrium relationship. A comparative study between linear (which is frequently used to determine the best-fitting isotherm) and nonlinear (which is used to avoid errors affecting  $R^2$  during linearity) isotherm analysis for the adsorption of basic green 4-dye by Sawdust-based activated carbon was the aim of this study [17-18].

# 1. Experimental details

## 2.1. Preparation of stock solutions

All reagents used in this study were of analytical grade from Merck, (Germany). A  $10^{-2}$  M of stock solutions was prepared according to standard procedure by dissolving the required amount of basic green 4-dye in distilled water. Successive dilutions were used to obtain working solutions of the desired concentrations. Table 1 shows Chemical structure and characteristics of basic green 4-dye.

Table1: Chemical structure and characteristics of basic green 4-dyc.					
	Characteristics	Chemical structure			
CI	42,000				
Molecular formula	$C_{52}H_{54}N_4O_{12}$				
Molecular weight	927.03g/mol				
Max. wave length $(\lambda_{max})^*$	615nm				
$LD_{50}$ (mice)	80 mg/kg	cí			
		н.с. С.			
		CH <sub>3</sub> CH <sub>3</sub>			

**Table1:** Chemical structure and characteristics of basic green 4-dye.

\* Experimentally obtained value

#### 2.2 Adsorbents preparation

In order to eliminate impurities, dust and to reduce moisture; sawdust samples were washed several times with distilled water then dried in an oven overnight at 378 K. The obtained samples were then grinded, sieved through 0.071 mm for intimate chemical agent–particle contact using a Vierzen Crosshop grinder before activation. Depending on the chemical agent chosen, three sawdust-based activated carbons were prepared in a two-step process involving a chemical impregnation using phosphoric acid, hydroxide potassium and sodium thiosulfate separately followed by pyrolysis at 600°C for 3 hours for each case as shown in Scheme 1.



Scheme 1: A two step-process carried out on activated process on sawdust (case: SD-1 preparation)

In general, activation using chemical agents or physical using heat is a way to enhance the adoptive properties of raw materials. In our case, we used an acid ( $H_3PO_4$ ), a base (KOH) and a salt ( $(NH_4)_2S_2O_8$ ) which are less haemfull to the environment. Phosphoric acid is the most preferred because of the environmental and economic concerns.Potassium hydroxide and salt develops large microporos. In all cases chemical agents and heat at 600°C increase the porosity of the precursor by eliminating most the volatile organic compounds and impureties which leave more space for pollutants adsorption, increasing then their uptake as described by Douara et al [27].

#### 2.3 Adsorption experiments

The effect of some important parameters such as adsorbent dosage, contact time range, pH and temperature on the adsorptive removal of basic green 4-dye onto the prepared activated carbons and the commercial one were performed in batch sorption experiments in a series of capped 250 mL Erlenmeyer flasks at a room temperature

 $25\pm1C$ . For such studies, a 25 mL volume of a dye solution of known initial concentration was mixed with a known amount of adsorbent. The resulting suspension was then agitated magnetically at a constant speed of 200 rpm until equilibrium had been reached. After the adsorption process had occurred, the resulting solution in each flask was centrifuged at 4000 rpm and the supernatant analyzed using a UV–Visible 2121 Optizen spectrophotometer at wavelength value of 615nm. The concentrations retained in the adsorbent phase ( $q_e$ ) were determined according to the mass balance using the following relationship:

$$q_{e} = \frac{\left(C_{0} - C_{eq}\right)}{m * 1000} * V \tag{1}$$

And the distribution coefficient is:

$$K_{d} = \frac{\left(C_{0} - C_{eq}\right)}{C_{eq} * m * 1000} * V$$
<sup>(2)</sup>

 $C_0$  and  $C_{eq}$  are the initial and the equilibrium dye concentration (mg/L) respectively, (mg/L), V is the volume of the liquid phase (mL), m is the mass of the activated carbon sample (mg) and  $K_d$  is the distribution coefficient (L/g). The performance of the adsorption is evaluated by using the removal efficiency ( $R_E$ ) defined as:

$$R_E(\%) = \frac{(C_0 - C_t)}{C_0} * 100$$
(3)

#### 2.4. Linear and nonlinear Langmuir and Freundlich isotherm

The isotherms data were analysed using two of the most commonly used equilibrium models namely the Langmuir isotherm which is based on monolayer coverage prediction of the adsorbate, this model also suggest that there is no lateral interaction between the sorbed molecules [19] and the Freundlich isotherm which is based on multilayer adsorption on heterogeneous surface [20]

Depending on both the nature of the adsorbent and the interaction type, various kinds of isotherms can be distinguished. In order to find a suitable model that can be used for design purposes; two well known and widely adopted adsorption isotherms, namely: Langmuir, which assumes that the adsorption takes place at specific homogeneous sites within the adsorbent with no lateral interaction between the sorbed molecules [19] and Freundlich [20] which describes adsorption as taking place on a heterogeneous adsorbent surface, were fitted to experimental data to describe the adsorption of basic green 4 dye at the solid-liquid interface. Table 2 summarizes the linear and non linear forms of the isotherms models used in this study. Where  $C_{eq}$  is the equilibrium concentration of the adsorbate (mg/L) and  $q_e$  is the amount of adsorbate adsorbed per unit mass of activated carbon (mg/g). Both Langmuir maximum uptake of dye per unit mass of adsorbent *b* (mg/g) and Langmuir constant related to the rate of adsorption  $K_L$  (L/mg) are obtained from ( $C_e/q_e$ ) vs  $C_e$  plots. While Freundlich constants  $K_f$  and *n* can be obtained from the intercept and the slope of  $ln q_e vs ln C_{eq}$  plots, with *n* indicating the favourableness (n>1) of the adsorption process and  $K_F$  the adsorption capacity of the adsorbent.

Table 2. Effect and non mean forms of Earginan and Fredhanen isotherms						
Isotherm	Nonlinear	Linear	Plot			
Freundlich	$q_e = K_F C_{eq}^{\frac{1}{n}}$	$\ln(q_e) = \ln K_F + (\frac{1}{n}) * \ln(C_e)$	$\ln(q_e) \text{ VS } \ln(C_e)$			
Langmuir-I		$\frac{C_e}{q_e} = (\frac{C_e}{q_m}) + (\frac{1}{b*q_m})$	$C_e/q_e$ vs $C_e$			
Langmuir-II	$q_e = \frac{q_m b C_e}{1 + b C}$	$\frac{1}{q_e} = (\frac{1}{b*q_m})*(\frac{1}{C_e})+(\frac{1}{q_m})$	$\frac{1}{q_e}$ vs $\frac{1}{C_e}$			
Langmuir-III	$1 + DC_e$	$q_e = q_m - (\frac{1}{b}) * (\frac{q_e}{C_e})$	$q_e vs q_e / C_e$			
Langmuir-IV		$q_e/C = b * q_m - b * q_e$	$q_e/c$ vs $q_e$			

Table 2: Linear and non linear forms of Langmuir and Freundlich isotherms

Another characteristic of Langmuir isotherm is the equilibrium parameter  $R_L$  given by equation (4) defining the nature of adsorption process that can be either unfavourable ( $R_L > 1$ ), linear ( $R_L = 1$ ), favourable ( $0 < R_L < 1$ ) or irreversible  $R_L = 0$ .

$$R_{L} = \frac{1}{1+b \ C_{0}} \tag{4}$$

#### 2.5. Error functions analysis

Distribution functions such as Chi-square ( $\chi^2$ ), Root mean square error (*RMSE*) and the Average percentage error (APE) are used in order to evaluate models [21], if data from the model are similar to the experimental

data,  $\chi^2$  will be a small number, if they are different,  $\chi^2$  will be a large number and the small the *RMSE* value, the better the curve fitting. In general plus values of errors function are low it means more there is an agreement between the experimental and calculate data and more the model converges and becomes favourable, their functions are listed below:

$$Chi - Square \ Statistic = x^{2} = \sum \frac{(q_{e,exp} - q_{e,cal})^{2}}{q_{e,cal}}$$
(5)

$$RMSE = \sqrt{(1/N - 2) * \sum_{1}^{N} (q_{e,exp} - q_{e,cal})^2}$$
(6)

$$APE = \left(\sum_{1}^{N} \left( \left| q_{e, \exp} - q_{e, cal} \right| / q_{e, \exp} \right) / N \right) * 100$$
<sup>(7)</sup>

With N is the number of observations in the experimental data.

## 2. Results and Discussion

The use of an iterative numerical optimization method to minimize the least-squares function is required by Non-Linear Regression (NLR), whereas an analytical solution is admitted by the linear one.

#### 3.1. linear fitting of the isotherm models

For the best fitting and finding the parameters of isotherms, linear regression analysis and the least squares method are frequently used. However, Langmuir and Freundlich isotherms in their linear forms mentioned in Table 1 were used not only for this purpose but also to describe the relationship between the amount of dye adsorbed  $q_e$  and its equilibrium concentration  $C_{eq}$  [21]. Figure 1 and Figure 2 show linear plots of both isotherm models for chosen dye by all considered adsorbents. Each of the four linear forms of Langmuir isotherm will result in different parameter estimates as shown in Table 3 [21-23]. The adsorption of basic 4- dye onto SD-1 and SD-2 adsorbents is well represented by Langmuir type I model with highest coefficient of determinations,  $R^2$  values as shown in Figure 1, indicate that there is strong evidence that the sorption of the chosen dye onto the prepared samples follows the Langmuir isotherm, from which a maximum capacity of 583.52 mg/g has been obtained compared to the other types of the same model according to  $R^2$  values and related Chi-square ( $\chi^2$ ), and Root mean square error (*RMSE*). While, the Average percentage error (*APE*) are higher for most linear forms except for the case of SD-2. This best fitting is due to the minimal deviations from the fitted equation resulting in the best error distribution. As shown in Figure 2, the Freundlich isotherm model also is well fitted for the adsorption of basic 4- dye onto using SD-1 and SD-3 with higher  $R^2$  and lower standard errors values suggesting that a satisfactory fit to the experimental data can be generated.

Table 4 summarizes the corresponding Freundlich isotherm parameters, their correlation coefficients ( $R^2$ ) and related standard errors for each parameter.

#### 3.2. Non linear fitting of the isotherm models

The non linear analysis is used in order to avoid errors raised by different estimates resulting from simple linear regression of the linearized forms of Langmuir equation presented in Table 2 which can affect  $R^2$  values significantly. To avoid such errors, we use the non-linear analysis as an adequate method [24]. It is an interesting way for describing adsorption isotherms used for many applications such as wastewater treatment from textile industry.

Langmuir and Freundlich adsorption isotherms of basic 4-dye by all four considered adsorbents using nonlinear analysis are shown in Figure 3; and their corresponding isotherm parameters, coefficients of determination ( $R^2$ ) and related standard errors (( $\chi^2$ , *RMSE*, *APE*) for each parameter are summarized in Table 3 and Table 4.

Higher values of  $R^2$  obtained in this study are derived by fitting experimental data into the Langmuir isotherm model for both SD-2 and Merck-AC samples. The Freundlich isotherm model best fits the data for SD-1 and SD-3 samples as shown in Table 1 and Figure 2. In addition, lower values of Chi-square ( $\chi^2$ ), and Root mean square error (*RMSE*) for each parameter obtained in both isotherm models with higher R<sup>2</sup> can generate a satisfactory fit to the experimental data. As shown in Table 1, maximum adsorption capacity values obtained using non linear Langmuir model, were 515.2, 435.9, 276.0 and 196.6 mg/g for SD-1, SD-2, SD-3 and Merch-AC respectively compared to inactivated sawdust capacity of 70 mg/g. These values are closer to the adsorbed quantities corresponding to saturation plateau obtained using equation (1) and shown in Figure 3, indicating the acceptability of Langmuir model.

The saturation amounts in Freundlich model obtained using non-linear regression are also closer to the experimental ones with lower errors function as shown in table 1 and Figure 4. which means that this model is

valid and can describe the equilibrium data. Figure 5 represents data obtained by both models in their linear and non linear forms compared to the experimental one for SD-2 sample taken as an example in this study. **Table 3:** Parameters of the Linear and non linear forms of Langmuir isotherms

	Туре	$R^2$	b (L/mg)	$q_m$	$\chi^{2}$	RMSE	APE		
SD-1 (H <sub>3</sub> PO <sub>4</sub> activation) Adsorption isotherm Conditions: $dose \ 8 \ \sigma/L: \ pH=8: time=3 \ h$									
$\frac{C_e}{q_e} = 0.0038 + 0.0017 * C_e$	Ι	0.998	0.452	583.52	0.0069	0.0026	61.733		
$\frac{1}{q_e} = 0.002 + 0.0002 * \left(\frac{1}{C_e}\right)$	II	0.529	11.556	501.06	0.0004	0.00024	9.161		
$q_e = 508.8 - 0.0945 * (\frac{q_e}{C_e})$	III	0.436	10.578	508.8	92.999	56.832	9.753		
$\frac{q_e}{C_e} = 2456.1 - 4.61 * q_e$	IV	0.436	4.61	532.7	2417.92	396.9	936		
$q_e = \frac{3554.511 * C_e}{1 + 6.902 * C_e}$	Non-linear	0.506	6.962	515.2	88.33	53.163	9.429		
A dama		SD-2 (KO	H activation	1) ~////(					
	ption isother		ions: aose 8	g/L; pH=0	; time=2 n				
$C_e/q_e = 0.0072 + 0.0022 * C_e$	Ι	0.997	0.309	448.788	0.003	0.0028	11.154		
$\frac{1}{q_e} = 0.0026 + 0.0043 * \left(\frac{1}{C_e}\right)$	II	0.871	0.61	377.809	0.0018	0.00088	11.608		
$q_e = 408.9 - 1.9454 * (\frac{q_e}{C_e})$	III	0.816	0.514	408.903	204.589	50.921	18.008		
$\frac{q_e}{C_e} = 180.91 - 0.419 * q_e$	IV	0.816	0.419	430.976	98.1148	23.653	44.704		
$q_e = \frac{163.268 * C_e}{1 + 0.374 * C_e}$	Non-linear	0.9647	0.3746	435.9	34.37	22.352	8.647		
Adsor	SD- ntion isother	-3 ((NH <sub>4</sub> ))	${}_{2}S_{2}O_{8}$ activations: dosa 8	tion) a/L : nH=6	time-? h				
	ption isother		ions: aose o	g/L;рп—о	; <i>ume–2</i> n				
$C_e / q_e = 0.3506 + 0.0037 * C_e$	Ι	0.976	0.011	268.419	0.66	0.186	91.409		
$\frac{1}{q_e} = 0.0059 + 0.0264 * \left(\frac{1}{C_e}\right)$	II	0.795	0.222	170.434	0.008	0.0021	25.246		
$q_e = 191.01 - 5.8438 * (\frac{q_e}{C_e})$	III	0.461	0.171	191.009	204.041	50.791	31.382		
$\frac{q_e}{C_e} = 16.873 - 0.079 * q_e$	IV	0.461	0.079	213.969	26.41	5.9	455.55		
$q_e = \frac{2.25 * C_e}{1 + 8.154 \text{E} - 03 * C_e}$	Non-linear	0.8672	0.0081	276	767.37	25.209	18.21		
Merck activated carbon Adsorption isotherm Conditions: dose 8 g/I : nH=6: time=2 h									
$C_{e/a} = 0.0168 + 0.005 * C_{e/a}$	I	0.996	0.296	201.26	0.0625	0.1539	100.76		
/ <i>Y<sub>e</sub></i> e									
$\frac{1}{q_e} = 0.0054 + 0.0019 * \left(\frac{1}{C_e}\right)$	II	0.911	2.811	184.33	0.00049	0.00075	7.898		
$q_e = 189.97 - 0.397 * (\frac{q_e}{C_e})$	III	0.853	2.478	190.128	15.35	18.45	10.27		
$\frac{q_e}{C_e} = 417.803 - 2.144 * q_e$	IV	0.853	2.119	194.965	-6.53	42.28	3163		

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$q_e = \frac{356.1 * C_e}{1 + 1.811 * C_e}$	Non-linear	0.9375	1.812	196.6	8.514	12.126	6.848
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Table 4: Parameters of the Linear and non linear forms of Freundlich isotherms

	$R^2$	п	$K_F$ (mg/g)	$\chi^{2}$	RMSE	APE				
SD-1 ( $H_3PO_4$ activation)										
Adsorption isotherm Conditions: <i>dose 8 g/L; <math>pH=8</math>; time=3 h</i>										
$q_e = 385.583 * C_{eq}^{8.753E-02}$	0.987	11.420	385.600	2.530	8.621	1.359				
$\ln(q_e) = 5.962 + 0.084 * \ln(C_e)$	0.984	11.779	388.442	0.001	0.021	0.225				
	SD-2 (KOH activation)									
Adsorption isotherm Conditions: dose 8 g/L; pH=6; time=2 h										
$q_e = 175.96 * C_{eq}^{0.227}$	0.894	4.389	176.00	78.300	38.610	14.575				
$\ln(q_e) = 5.003 + 0.2854 * \ln(C_e)$	0.875	3.501	148.803	0.0683	0.175	2.495				
SD-3 (NH <sub>4</sub> ) <sub>2</sub> S <sub>2</sub> O <sub>8</sub> activation										
Adsorptio	on isothern	n Conditions:	dose 8 g/L; pH	I=6; time=2	h					
$q_e = 34.803 * C_{eq}^{0.292}$	0.963	3.415	34.80	14.459	13.301	7.481				
$\ln(q_e) = 3.695 + 0.266 * \ln(C_e)$	0.969	3.752	40.243	0.019	0.089	1.363				
Merck activated carbon										
Adsorption isotherm Conditions: <i>dose 8 g/L; pH=6; time=2 h</i>										
$q_e = 128.423 * C_{eq}^{7.064 \text{E}-02}$	0.766	14.160	128.400	22.909	23.469	12.001				
$\ln(q_e) = 4.801 + 0.083 * \ln(C_e)$	0.767	12.017	121.650	0.033	0.164	2.447				



Figure 1: Linear fitting of the Langmuir (Case of SD-1 and SD-2 samples)



Figure 2: Linear fitting of the Freundlich isotherms models for all samples



Figure 3: Non Linear fitting of the Langmuir isotherms models for SD-1 and SD-2 samples



Figure 4: Non Linear fitting of the Freundlich isotherms models for SD-1 and SD-2 samples



Figure 5: Non Linear fitting of the Langmuir and Freundlich isotherms models for SD-2 sample

Separation factor  $R_L$ , a dimensionless constant defined by equation 4 is used to express the essential features of the Langmuir isotherm. [25]

All  $R_L$  values do exist between  $0 < R_L < 1$  indicating then, that the prepared samples are favorable for basic green dye adsorption under conditions used in this study as shown in figure 6 (case of SD-2 taken as an example). Also, the adsorption intensity for all adsorbents used in this study found from the linear Freundlich isotherm model are higher (n> 1) implying a good adsorption process with lower errors functions and higher R<sup>2</sup> values except for the one of commercial activated carbon (R<sup>2</sup> = 0.875). We noticed that the Langmuir model in its linear form (type-I) and the nonlinear one gave a high values of the equilibrium parameter compared to the other models.



Figure 6: Separation factor-linear and nonlinear Langmuir isotherm model (SD-2 case)

## Conclusions

The adsorption of basic green 4-dye onto a chemically modified sawdust using Phosphoric acid, potassium hydroxide and ammonium persulfate as oxidizing agent and Merck activated carbon (used as a reference), from aqueous solution was investigated on the basis of equilibrium studies. The isotherm models were carried out at different initial dye concentrations and constants characterising each system (such as pH; dose and equilibrium time) previously identified and recapitulated in above tables. Linear and nonlinear Langmuir and Freundlich isotherm and error functions were used to model the experimental data. Higher determination coefficients

provided by linear forms of Langmuir: type I (all samples) best fit the data with higher adsorption capacities compared to the other types and non linear isotherm models. Lower error function analysis for non linear regression of Freundlich isotherm meaning the validity of this model for describing the equilibrium data. The validity of this model can also be proven by the obtained R<sup>2</sup> values except for the Merck-AC and SD-2 samples. Higher uptakes ranging from 170.43 to 583.52 mg/g (depending of the oxidizing agent) corresponding to 99.9 %, 99.6% and 70.1% for SD-1, SD-2 and SD-3 respectively compared to 59.9% obtained by Merck-AC, indicate that the surface modified sawdust can be used as alternative adsorbents in textile wastewater treatment.

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#### References

- 1. M. Ahmad, A.S. Bajahlan, W.S. Hammad, Environ. Monitor. Asses. 147 (2008) 297.
- 2. O. Botalova, J. Schwarzbauer, Water, Air, & Soil Pollution. 221 (2011) 77.
- 3. Y. Akria, R.M. David, E. Philip, Proceedings of 21<sup>st</sup> Mid-Atlantic Industrial Waste Conference, Harrisburg, Technomic, Lancaster, PA, USA, (1989).
- 4. A.K. Verma, R. Dash, P. Bhunia, J. Environ. Manag. 93 (2012) 154.
- 5. I. Bisschops, H. Spanjers, Environ. Tech. 24 (2003) 1399.
- 6. S. Sirianuntapiboon, P. Srisornsak, Bioresour. Tech. 98 (2007) 1057.
- 7. T. Robinson, G. McMullan, R. Marchant, P. Nigam, Bioresour. Tech. 77 (2001) 247.
- 8. G. Mesohegyi, P. Frank, F.T. Van der Zee, J. Font, A. Fortuny, A. Fabregat, J. Environ. Manage. 102 (2012) 148.
- 9. J. Panswed, S. Wongchaisuwan, Water Sci. Tech. 18 (1986) 139.
- 10. C. Theivarasu, S. Mylsamy, Int. J. Eng. Sci. Tech. 2 (2010) 6284.
- 11. B. Emrah, O. Mahmut, S. Ayhan, Micro. Meso. Mater. 115 (2008) 234.
- 12. W. Eckhard, Adsorption Technology in Water Treatment, Fundamentals, Processes, and Modeling. Walter de Gruyter GmbH & Co. KG, Berlin/Boston (2012).
- 13. K.S. Bharathi, S.T. Ramesh, Appl. Water Sci. 3 (2013) 773
- 14. H.X. Li, R.J. Zhang, L.Tang, J.H. Zhang, Z.G. Mao, Bioresour. 9 (2014) 6698.
- 15. M. Urík, P. Littera, J. Ševc, M. Kolenčík, S. Čerňanský, Int. J. Environ. Sci. Tech. 6 (2009) 451.
- 16. R. Rajeshkannan, M. Rajasimman, N. Rajamohan, Chem. Ind. & Chem. Eng. Quarterly. 17 (2011) 67.
- 17. R. Rajeshkannan, M. Rajasimman, N. Rajamohan, Int. J. Civ. Environ. Eng. 2 (2010) 222.
- 18. Y.S. Ho, Polish J. Environ. Studies. 15 (2006) 81.
- 19. K.V. Kumar, S. Sivanesan, V. Ramamurthi, Process Biochem. 40 (2005) 2865.
- 20. I. Langmuir, J. Am. Chem. Soc. 40 (1918) 1361.
- 21. H. Freundlich, J. Phys. Chem. 57 (1906) 385.
- 22. S. Mohammadi-Kalalagh, H. Babazadeh, Int. J. Environ. Sci. Tech. 11 (2014), 111/
- 23. F. Yang, X.J. Song, W.M. Wei, S. Yao, C.K. Xia, Z.Z. Yang, Asian J. Chemistry. 26 (2014), 1521.
- 24. Z. L. Yaneva, B. K. Koumanova, N. V. Georgieva, J. Chemistry. Volume 2013 (2013) 10 pages.
- 25. C. Xunjun, Information 6 (2015) 14.
- 26. A.T. Adeolu, O.A. Okareh, A.O. Dada, Amer. J. Environ. Protect. 4 (2016) 7.
- 27. N. Douara, B. Bestani, N. Benderdouch, L. Duclaux, Desalin. Water treat. 57 (2015), 5529.

# (2018); <u>http://www.jmaterenvironsci.com</u>