Factorial design and statistical analysis of smectite clay treatment by hydrocyclone

(Planejamento fatorial e análise estatística do tratamento de argilas esmectíticas por hidrociclone)

A. J. A. Gama¹, J. M. R. Figueirêdo¹, A. L. F. Brito², M. A. Gama², G. A. Neves², H. C. Ferreira²

¹Programa de Pós-Graduação em Ciência e Engenharia de Materiais; ²CCT, UFCG, Av. Aprigio Veloso 882, Campina Grande, PB 58429-900
agama@reitoria.ufcg.edu.br, jullymrc@gmail.com, andre.fiquene@ufcg.edu.br,
mgama@reitoria.ufcg.edu.br, gelmires.neves@ufcg.edu.br, heber.ferreira@ufcg.edu.br

Abstract

Bentonite clays are materials composed by one or more smectite clay minerals and some accessory minerals, mainly quartz, cristobalite, mica, feldspars and other clay minerals such as kaolinite. These contaminants present in clays have a large distribution of particle sizes which severely restrict their industrial applications, with the use of hydrocyclone as a likely solution for their reduction. This study aims to analyze the treatment of smectite clays from the state of Paraíba using modeling, simulation and optimization of the variable average particle diameter in relation to various process variables related to the hydrocyclone. In this study, the average diameter of smectite clays was evaluated as a function of the factors: pressure, apex diameter and vortex diameter of the hydrocyclone. Complete factorial design and addition in the central points were used to model the hydrocycloning process. The results evidenced reduction in equivalent average particle size of approximately 19.2%. Regarding the simulations, the optimum point with the lowest value was found for the average diameter of 4.033 µm, with a pressure of 4.3 bar, apex opening of 5.3 mm, and vortex opening of 6.3 mm, all at a 95% confidence level.

Keywords: modeling, simulation, hydrocyclone, smectite.

INTRODUCTION

Bentonites are sedimentary clays formed by devitrification and subsequent chemical alteration of volcanic tuffs or ash, consisting essentially of the clay mineral montmorillonite belonging to the group of smectite. These clays are constituted by lamella formed by an octahedral sheet of alumina (Al₂O₃) between two tetrahedral silica sheets (SiO₂). Bentonite have wide industrial applications, being used as sand binders in foundry/casting molds, in drilling fluids, etc. [1-4]. Smectite clays, in addition to presenting montmorillonite clay minerals in their constitution, also present high amounts of accessory minerals such as quartz, carbonate, mica, feldspar and kaolinite. These high levels of contaminants severely restrict their industrial applications, with the use of hydrocyclone as a likely solution for their reduction.

The hydrocyclone is important equipment for the separation of metallic and non-metallic minerals. The principle of separation is centrifugal sedimentation, in which the suspended particles are subjected to a centrifugal field that causes their separation from the fluid [5, 6]. It consists of a conical section/base connected to a cylindrical part in which there is a tangential inlet for the feed slurry.
The inlet flow is divided into two flows: an upper flow, which leads to the overflow of the classifier and carries the lighter particles; and a lower flow, which leads to the underflow of the classifier, and carries heavier particles [7, 8]. As they do not have moving parts, hydrocyclones require low investment in installation and maintenance and are simple to operate, being widely used in the chemical, mineral, textile and petrochemical industries. Some studies using hydrocyclone for the treatment of minerals have been highlighted in recent years, among them the works [9-21]. These studies concluded that this type of equipment has high efficiency in the separation of fine and coarse particles, with easy operation requiring low investment for installation and maintenance, and having relevant importance in the treatment of non-metallic minerals.

An important tool in the treatment of clays by hydrocycloning is modeling and simulation, which enables optimization of products and processes, thus minimizing duration and treatment costs, and maximizing performance, productivity and the quality of products. Modeling is able to determine and even to quantify the influence of independent variables on treatment and on dependent variables, making it possible to obtain reliable results that can be statistically analyzed [22]. Few studies on modeling, simulation and optimization of response variables in relation to the average diameter of clay particles are found in the literature. Villasana et al. [23] studied the planning and modeling of hydrocyclone in relation to the pressure parameters and flow rate of an integrated hydrocyclone system. Schneider and Neege [24] examined the hydrocyclone simulation for the treatment of non-metallic minerals. Özgen et al. [25] performed modeling and simulation with bentonite in relation to the response variables: cation exchange capacity, free swelling volume and apparent viscosity. Menezes et al. [26] evaluated the influence of CMC (carboxymethyl cellulose) in rheological properties of smectites using simulation for use in oil well drilling fluids. Boylu [12] conducted studies with sodic bentonite for optimization of hydrocyclone, and concluded that the best results were obtained with 4 bars of pressure, vortex opening of 14.3 mm, and apex of 8.0 mm.

In this study, the average diameter of smectite clays was evaluated in terms of hydrocyclone pressure factors, apex diameter and vortex diameter. A complete factorial design was adopted with addition at the central points was used to model the hydrocycloning process, as optimizing factorial design with contour surface can be successfully applied in processing clays in order to evaluate the optimum factor points, simulate the best responses, and evaluate the effects of the main factors, in addition to evaluating the interactions between the adopted factors. This study aimed to investigate the treatment of smectite clays from the state of Paraíba, located on Campos Novos farm in the municipality of Cubati-PB, and designated as AM1 was used. The sample was enriched through disaggregation in a ball mill, then classified through a ABNT nº 200 sieve (0.074 mm). According to [21], the chemical composition of clay was composed of: 50.77% SiO₂, 21.78% Al₂O₃, 7.43% Fe₂O₃, 3.07% MgO, 2.47% CaO, 0.89% K₂O, 1.30% other oxides, and 12.29% (LOI).

A RWK 42L hydrocyclone manufactured by Netzsch AWK, Germany, was used for treating this clay. It was dimensioned for particle separation with an average diameter of 2 to 5 µm. A schematic of this equipment can be found elsewhere [27]. The process variables used in this study were: pressure (3.0 to 4.0 bar), opening of the vortex (5.0, 5.5 and 6.0 mm) and opening of the apex (3.0, 4.0 and 5.0 mm). The dispersions were prepared in the following manner: 1200 g of clay (dried in an oven at 60 °C for moisture removal) was mixed in a volume of 30 L of distilled water with sodium hexametaphosphate at a ratio of 45.7 g/L, obtaining a dispersion of clay in water at a concentration of 4% by mass kept under agitation at a speed of 2000 rpm for a period of 24 h at room temperature. After this period, the dispersion was pumped into the hydrocyclone for centrifugation with pressures ranging from 3.0 to 4.0 bar for each configuration of the apex and vortex.

Physical, chemical and mineralogical characterization of the sample in natura and after hydrocycloning were performed using: granulometric analysis by laser diffraction (Cilas, mod. 1064); X-ray diffraction (XRD, Shimadzu, XRD-6000), with Ka radiation of Cu (40 kV/30 mA) and goniometer speed of 2 °/min, step width of 0.02° and scan range from 5° to 40°. Experimental planning: a b² factorial design was adopted with the addition of 3 replicates at the center point, considering the center point as the arithmetic average of the levels in the factors. In the 2k planning, k represents the number of factors and b represents the number of levels. A 2² design was used with addition of three central points, considering the factors: apex diameter, vortex diameter and pressure. Table 1 shows the corresponding levels of the factors. A concern in using 2k planning is the linearity assumption on the effects of factors. 2k planning works well when the linearity assumption remains approximate. Therefore, the addition of CtPt is used to increase the number of levels in 2k, meaning, from 2 to 3 levels. The central points were added in triplicate, since they do not have repercussions on estimates of the effects on planning [28, 29].

Statistical analysis of data: analysis of variance (ANOVA) was carried out in order to verify whether there are significant differences between the results obtained for each factor. The data were submitted for ANOVA using the statistical package Statgraphics Centurion 18. A 5% level of significance was adopted for the F test. Table 1 shows the results of the analysis of variance for the factors analyzed. The data were subjected to the F test, and the results are shown in Table 1. Table 1 includes the F values, the p values and the level of significance (α = 0.05). The results of the analysis of variance show that the factors apex diameter and vortex diameter have a statistically significant effect on the response variables, as the p values are less than 0.05, indicating that these factors are significant factors in determining the response variables. The remaining factors have a non-significant effect on the response variables, as the p values are greater than 0.05. These results indicate that the factors apex diameter and vortex diameter have a significant impact on the response variables, and should be considered in future studies.

<table>
<thead>
<tr>
<th>Factor</th>
<th>Low level (-1)</th>
<th>Center point (0)</th>
<th>High level (+1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Apex diameter (mm)</td>
<td>3.0</td>
<td>4.0</td>
<td>5.0</td>
</tr>
<tr>
<td>Vortex diameter (mm)</td>
<td>5.0</td>
<td>5.5</td>
<td>6.0</td>
</tr>
<tr>
<td>Pressure (bar)</td>
<td>3.0</td>
<td>3.5</td>
<td>4.0</td>
</tr>
</tbody>
</table>

MATERIALS AND METHODS

A sample of bentonite clay from the new deposits of the state of Paraíba, located on Campos Novos farm in the municipality of Cubati-PB, and designated as AM1 was used. The sample was enriched through disaggregation in a ball mill, then classified through a ABNT nº 200 sieve (0.074 mm). According to [21], the chemical composition of clay was composed of: 50.77% SiO₂, 21.78% Al₂O₃, 7.43% Fe₂O₃, 3.07% MgO, 2.47% CaO, 0.89% K₂O, 1.30% other oxides, and 12.29% (LOI).

A RWK 42L hydrocyclone manufactured by Netzsch AWK, Germany, was used for treating this clay. It was dimensioned for particle separation with an average diameter of 2 to 5 µm. A schematic of this equipment can be found elsewhere [27]. The process variables used in this study were: pressure (3.0 to 4.0 bar), opening of the vortex (5.0, 5.5 and 6.0 mm) and opening of the apex (3.0, 4.0 and 5.0 mm). The dispersions were prepared in the following manner: 1200 g of clay (dried in an oven at 60 °C for moisture removal) was mixed in a volume of 30 L of distilled water with sodium hexametaphosphate at a ratio of 45.7 g/L, obtaining a dispersion of clay in water at a concentration of 4% by mass kept under agitation at a speed of 2000 rpm for a period of 24 h at room temperature. After this period, the dispersion was pumped into the hydrocyclone for centrifugation with pressures ranging from 3.0 to 4.0 bar for each configuration of the apex and vortex.

Physical, chemical and mineralogical characterization of the sample in natura and after hydrocycloning were performed using: granulometric analysis by laser diffraction (Cilas, mod. 1064); X-ray diffraction (XRD, Shimadzu, XRD-6000), with Ka radiation of Cu (40 kV/30 mA) and goniometer speed of 2 °/min, step width of 0.02° and scan range from 5° to 40°.

Experimental planning: a b² factorial design was adopted with the addition of 3 replicates at the center point, considering the center point as the arithmetic average of the levels in the factors. In the 2k planning, k represents the number of factors and b represents the number of levels. A 2² design was used with addition of three central points, considering the factors: apex diameter, vortex diameter and pressure. Table 1 shows the corresponding levels of the factors. A concern in using 2k planning is the linearity assumption on the effects of factors. 2k planning works well when the linearity assumption remains approximate. Therefore, the addition of CtPt is used to increase the number of levels in 2k, meaning, from 2 to 3 levels. The central points were added in triplicate, since they do not have repercussions on estimates of the effects on planning [28, 29].

Statistical analysis of data: analysis of variance (ANOVA) was carried out in order to verify whether there are significant differences between the results obtained for each factor. The data were submitted for ANOVA using the statistical package Statgraphics Centurion 18. A 5% level of significance was adopted for the F test. Table 1 shows the results of the analysis of variance for the factors analyzed. The data were subjected to the F test, and the results are shown in Table 1. Table 1 includes the F values, the p values and the level of significance (α = 0.05). The results of the analysis of variance show that the factors apex diameter and vortex diameter have a statistically significant effect on the response variables, as the p values are less than 0.05, indicating that these factors are significant factors in determining the response variables. The remaining factors have a non-significant effect on the response variables, as the p values are greater than 0.05. These results indicate that the factors apex diameter and vortex diameter have a significant impact on the response variables, and should be considered in future studies.

<table>
<thead>
<tr>
<th>Factor</th>
<th>Low level (-1)</th>
<th>Center point (0)</th>
<th>High level (+1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Apex diameter (mm)</td>
<td>3.0</td>
<td>4.0</td>
<td>5.0</td>
</tr>
<tr>
<td>Vortex diameter (mm)</td>
<td>5.0</td>
<td>5.5</td>
<td>6.0</td>
</tr>
<tr>
<td>Pressure (bar)</td>
<td>3.0</td>
<td>3.5</td>
<td>4.0</td>
</tr>
</tbody>
</table>
effects between the mean responses of the treatments. The procedure was used to infer whether such effects really exist, to determine the level of confidence [10]. In this case, the idea was to perform a breakdown of the variation into the variation attributed to differences between the experimental units (mean square of the residue - MSR) and the variation attributed to differences between the experimental units and attributed to the differences caused by treatments or factors (mean square of treatments - MS treat). The test is based on two hypotheses: $H_0$: there is no main effect of factors, interaction or curvature (means are equal); $H_1$: there is at least one major effect of factors, interaction or curvature (means are different). The F-test was used to compare means. If the value of $F_{Cal}$ is $>F_{Tab}$, the mean of the factors or interaction are different, meaning that there was effect of the factor and/or interaction. In this case, the controllable variables influenced the response variable. If the value of $F_{Cal}$ is $≤F_{Tab}$ the means of factors or interaction are equal. This means that there is no effect of the factor and/or interaction, and that the controllable variables do not influence the response variable. Table II shows the configuration for Anova (analysis of variance). The coefficient of determination ($R^2$) was calculated using Eq. A. It represents the explained variation of the model.

$$R^2 = \frac{\text{SQ Main - effect}}{\text{SQ Total}}$$  \hspace{1cm} (A)

The correlation coefficient ($R$) was calculated using Eq. B which represents the correlation between the variables. $R$ values range between -1 and 1, and the closer to 1 means better results. Value -1 indicates a negative linear correlation, and value 1 indicates a positive linear correlation.

$$R = \sqrt{R^2}$$  \hspace{1cm} (B)

The maximum percentage of explainable variation was calculated using the Eq. C, which represents the maximum variation that the model explains. This value was compared with the value explained by the model ($R^2$max).

$$R^2 \text{ max} = \frac{\text{SQ Total}}{\text{SQ Pure error}}$$  \hspace{1cm} (C)

In the factorial design $2^3$ with addition of 3 CtPt (central points), the adopted model is presented in Eqs. D and E, assuming that the response surfaces are linear and/or quadratic function of factors:

$$y = \beta_0 + \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 AB + \beta_5 BC + \beta_6 ABC + \varepsilon$$  \hspace{1cm} (D)

$$y = \beta_0 + \beta_1 A + \beta_2 B + \beta_3 C + \beta_4 AB + \beta_5 BC + \beta_6 ABC + \beta_7 A^2 + \beta_8 B^2 + \beta_9 C^2 + \varepsilon$$  \hspace{1cm} (E)

where $\beta_0, \beta_1, \beta_2, \beta_3, \beta_4, \beta_5, \beta_6, \beta_7$ and $\beta_8$ are the parameter estimates of the model, while $A$, $B$ and $C$ represent the factors adopted for apex, vortex and pressure, respectively.

**RESULTS AND DISCUSSION**

Fig. 1 shows the XRD patterns of the AM1 sample in natura, with and without ethylene glycol. It was observed that the sample presented the following mineralogical compositions: smectite (JCPDS 29-1497), characterized by basal interplanar distances of 15.66, 3.59, 4.43 and 2.55 Å, which was confirmed by the use of ethylene glycol which increased the interlayer distance of 15.66 for 16.66 Å; presence of the kaolinite clay mineral (JCPDS 14-0164), characterized by a basal interplanar distance of 7.13 Å; presence of accessory minerals such as quartz (JCPDS 46-1045), characterized by basal interplanar distances of 4.22, 3.34, 1.82 and 1.66 Å; feldspar (JCPDS 13-0294), characterized by a basal interplanar distance of 3.22 Å; and calcite (JCPDS 05-0586), characterized by a basal interplanar distance of 3.01 Å. These results showed that the clay used in this study had elevated levels of accessory minerals such as kaolinite, mica, quartz, feldspar and calcite. These results are similar to preliminary results found in [19].

Fig. 2 illustrates the particle size distribution of the sample in natura. It was observed that the clay in natura presented bimodal behavior with wide distribution range varying from 0.1 to 11 µm. The clay fraction corresponded to 20.89% (D<2 µm), D50 of 4.45 µm and an average particle diameter of 5.21 µm. These values were similar to those found in the literature on bentonite clays originating in the state of Paraíba [30].

**Table II - Analysis of variance (ANOVA) for 3 factors.**

<table>
<thead>
<tr>
<th>Source of variation</th>
<th>D.F.</th>
<th>Square sum (SS)</th>
<th>Adjusted mean square (MS)</th>
<th>$F$ value $^1$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Main effects</td>
<td>2</td>
<td>SS main effects</td>
<td>MS main effects</td>
<td>-</td>
</tr>
<tr>
<td>Interaction</td>
<td>1</td>
<td>SS interaction</td>
<td>MS interaction</td>
<td>-</td>
</tr>
<tr>
<td>Curvature</td>
<td>1</td>
<td>SS curvature</td>
<td>MS curvature</td>
<td>-</td>
</tr>
<tr>
<td>Error</td>
<td>2</td>
<td>SS error residual</td>
<td>MS error</td>
<td>-</td>
</tr>
<tr>
<td>Total</td>
<td>6</td>
<td>SS total</td>
<td></td>
<td>-</td>
</tr>
<tr>
<td>$R^2$</td>
<td></td>
<td>$R^2$ max</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: $^1$ - significant effect ($F_{Cal} > F_{Tab}$); non-significant effect ($F_{Cal} ≤ F_{Tab}$); D.F. - degrees of freedom; $R^2$ - coefficient of determination; $R^2$ max - maximum explainable variation.
Table III contains the results of the average diameter of particle sizes. Regarding the hydrocyclone process variables, from in natura clay particle size distribution (Fig. 2), a reduction in the average equivalent diameter of the particles was found to be approximately 19.2% for the configuration apex 5 mm, vortex 6 mm and pressure 4 bar. In general, it can be concluded that the increased pressure and apex/vortex openings have very important influence on the hydrocycloning process in relation to the reduction in particle size, a fact that can greatly contribute to the use of these clays for new industrial applications. In studies [21, 30] on the use of hydrocycloning for low pressures below 3 bar, it was found that the clay fraction concentrate values were lower than those

Table III - Average diameter of particle sizes after hydrocycloning.

<table>
<thead>
<tr>
<th>Essay</th>
<th>Apex, A (mm)</th>
<th>Vortex, B (mm)</th>
<th>Pressure, C (bar)</th>
<th>Average diameter (µm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>3</td>
<td>5</td>
<td>3</td>
<td>4.61</td>
</tr>
<tr>
<td>2</td>
<td>5</td>
<td>5</td>
<td>3</td>
<td>4.85</td>
</tr>
<tr>
<td>3</td>
<td>3</td>
<td>6</td>
<td>3</td>
<td>4.96</td>
</tr>
<tr>
<td>4</td>
<td>5</td>
<td>6</td>
<td>3</td>
<td>4.31</td>
</tr>
<tr>
<td>5</td>
<td>3</td>
<td>5</td>
<td>4</td>
<td>4.58</td>
</tr>
<tr>
<td>6</td>
<td>5</td>
<td>5</td>
<td>4</td>
<td>4.65</td>
</tr>
<tr>
<td>7</td>
<td>3</td>
<td>6</td>
<td>4</td>
<td>4.68</td>
</tr>
<tr>
<td>8</td>
<td>5</td>
<td>6</td>
<td>4</td>
<td>4.24</td>
</tr>
<tr>
<td>9</td>
<td>5</td>
<td>5.5</td>
<td>3.5</td>
<td>4.81</td>
</tr>
<tr>
<td>10</td>
<td>4</td>
<td>5.5</td>
<td>3.5</td>
<td>4.82</td>
</tr>
<tr>
<td>11</td>
<td>4</td>
<td>5.5</td>
<td>3.5</td>
<td>4.83</td>
</tr>
</tbody>
</table>

**Notes:** 1 - significant effect ($F_{cal} \geq F_{tab}$); 2 - non-significant effect ($F_{cal} < F_{tab}$); D.F. - degrees of freedom; $R^2$ - coefficient of determination; $R^2_{max}$ - maximum explainable variation.
obtained in this study.

Table IV shows the ANOVA (analysis of variance) results of the average diameter. It was observed that all the factors were significant at a 95% confidence level, except for the interaction between the apex factor and pressure. By evaluating only the apex factor and the factor pressure, we can assert with a probability of 5% error that these two factors interact without influencing the response. On the other hand, the apex, vortex and pressure factors influenced the response when analyzed individually. Fig. 3 shows the values of the main factor effects for the average diameter. The higher response values for the average diameters of the apex, vortex and pressure occurred when the level was at the lowest value: 3 mm for apex, 5 mm for vortex, and 3.0 bar for pressure.

In Table IV, the analysis of variance showed that a second-order model can possibly be adopted, as there is curvature in the evaluated region. However, since the obtained model (Eq. A) can explain 99.96% of the data and main effects were significant, Eq. F can be adopted as being predictive at a 95% confidence level. In Eq. F the principal factors apex, vortex and pressure were significant and appear in the equation with high coefficients, being respectively: +5.450, +4.145 and +4.325. The positive sign that precedes the factors indicates the importance of the factor in hydrocyclone. Regarding the interaction, the significant factors influencing the response were: -1.015 for the apex and vortex interaction, -0.820 for the vortex and pressure interaction, and +0.190 for the apex, vortex and pressure. The model explains 99.96% of the data ($R^2$), indicating that the experimental error presented a small contribution. Fig. 4 confirms the explanation of the model in which the adjusted values as a function of the residues were in the range between +0.001 and -0.010, indicating a small margin of error.

$$Y(\mu m) = -17.15 + 5.450 \text{Apex} + 4.145 \text{Vortex} + 4.325 \text{Pressure} - 1.015 \text{Apex.Vortex} - 0.820 \text{Vortex.Pressure} + 0.190 \text{Apex.Vortex.Pressure} + 0.210 \text{Cl.Pt}$$

Regarding optimization, it was possible to obtain the optimal or stationary point of the adopted factors. In this case, the optimum point was equal to 4.24 $\mu$m (average diameter) which was obtained when the apex factor was equal to 5 $\mu$m, the vortex factor was equal to 6 $\mu$m, and pressure factor to 4 bar (Fig. 5). This value represents the optimal response of initially adopted factors in the experimental design. Fig. 6 illustrates the surface contour and optimized response. The interactions of the vortex and apex factors shown in Fig. 6a and the pressure and vortex interactions shown in Fig. 6b show that the best responses for the mean particle diameter are at the top of the graphs and indicate the region of the best response. From the result obtained for the optimum point (4.24 $\mu$m) which is the optimized response, several simulations in the optimum region can be performed.

Table V contains the simulated values in the optimum point region of the significant factors. 20 simulations were performed to test the best response of the average diameter. First, by varying the pressure and by fixing the apex and vortex (simulations 1 to 10); then, by varying all factors in the region near the optimum point (simulations 11 to 20). The lowest values for the average diameter was achieved when pressure increased. This value was obtained in the simulations 1 to 10, wherein the diameter of the apex was set at 4 mm, the vortex at 5 mm (optimal points), and the pressure in the optimum region varied (between 8 to 10.5 bar). The best result was obtained with a pressure equal to 10.5 bar, obtaining average size equal to 3.867 $\mu$m. Next, simulations
11 to 20 were performed by varying all the factors in the regions close to the optimum point; the lowest value for the average diameter was obtained in simulation 16, wherein the value for the average diameter was equal to 4.033 µm and pressure, apex and vortex values were higher at 4.3 bar, 5.3 mm and 6.3 mm, respectively.

CONCLUSIONS

After studying hydrocyclone process variables for treating smectite clays using modeling, simulation and optimization, it can be concluded that: i) there was a reduction in the average equivalent particle size of approximately 19.2% for the hydrocyclone configuration of apex (5 mm), vortex (6 mm) and pressure (3.5 bar); ii) all factors were significant at a 95% confidence level, meaning that apex, vortex and pressure factors influence the average diameter of the clay when analyzed individually; iii) the optimum point or minimized response was equal to 4.24 µm (average diameter) which was obtained when the apex factor was equal to 5 mm, vortex factor was equal to 6 mm and pressure factor was 4 bar; however, by performing simulations at the optimum point, the lowest value for the average diameter equal to 4.033 µm was found with far superior pressure, apex and vortex of 4.3 bar, 5.3 mm and 6.3 mm, respectively; iv) the process variables have a significant influence in the hydrocycloning process in relation to each type of accessory minerals present in clay; a fact that can greatly contribute to the use of these clays in industrial applications.

REFERENCES


(Rec. 02/12/2016, Rev. 10/02/2017, 27/04/2017, Ac. 27/04/2017)