

Spectrophotometric determination profile of bisphenol a in the photodegradation process with singlet molecular oxygen

Determinación espectrofotométrica del perfil del bisfenol a en el proceso de fotodegradación con oxígeno molecular singlete

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Abstract

In this study, a strategy using Partial Least Squares (PLS) as a chemometric tool to establish the robustness of the photosensitized degradation process of *Bisphenol A* in simulated effluents was assessed. In the first stage of the research, the validation of the methodology that allows the quantification of *Bisphenol A* (BPA) mixed with the photosensitizer *Eosin Y* (EOS) in aqueous solution was carried out. Through the use of spectrophotometry as monitoring technique, it was possible to establish the optimal number of factors and the wavelength range, and to determine that was possible to ensure the quantification of BPA by means of PLS, by demonstrating complex formation between BPA and EOS. The RMSE results were 0.004237 at 284 nm and 0.009167 at 524 nm, demonstrating good robustness in the analysis. Next, preliminary tests were carried out on the effect of direct light using a mercury lamp as a source of irradiation. In the case of BPA, it was determined that 100% degradation is achieved at 100 minutes.

Keywords: Chemometric, PLS, robustness, quantification, spectrophotometry, photosensitizer, photosensitization.

Resumen

En este estudio se propuso evaluar una estrategia utilizando Partial Least Squares (PLS) como herramienta quimiométrica para establecer la robustez del proceso de degradación fotosensibilizada de Bisfenol A (BPA) en efluentes simulados. En esta la etapa del trabajo de investigación, se llevó a cabo la validación de la metodología que permitió la cuantificación de BPA mezclado con el fotosensibilizador Eosin Y (EOS) en solución acuosa. Utilizando la espectrofotometría como técnica de seguimiento, fue posible establecer el número óptimo de factores y el rango de longitud de onda, y determinar que es posible asegurar la cuantificación de BPA mediante PLS al demostrar la formación de complejos entre BPA y EOS. Los resultados de RMSE fueron 0,004237 a 284 nm y 0,009167 a 524 nm, lo que demuestra una buena solidez en el análisis. A continuación, se realizaron pruebas preliminares sobre el efecto de la luz directa utilizando una lámpara de mercurio como fuente de irradiación. En el caso del BPA se determinó que se alcanza el 100% de degradación a los 100 minutos. El fotosensibilizador se fotodegradó a los 50 minutos de exposición

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a la luz directa. La mezcla de los dos compuestos en el análisis preliminar podría demostrar un efecto de pantalla por parte del fotosensibilizador.

Palabras clave: quimiometría, PLS, robustez, cuantificación, espectrofotometría, fotosensibilizador, fotosensibilización.

1. INTRODUCTION

The BPA is an ingredient of plastic products and has been used for over fifty years. Chemically, it is an organic compound consisting of two molecules of phenol attached to a chain of three saturated carbons (Figure 1). It is in two chemical configurations and has other isomers depending on several positions of substituent or united by a nucleophile. These compounds are normally manufactured from phenol condensate with acetone under low pH and high temperature in the presence of a catalyst. A second is similar but uses different catalysts and a different purification technology that generates a larger amount of pollutant residues[1].

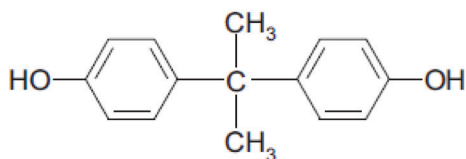


Figure 1. Chemical structure of BPA

The BPA is an ingredient of plastic products that has been used for over fifty years. The BPA is a

compound that persists in the environment and is an endocrine disruptor even in low concentrations[2] [3]. It can alter the natural hormonal balance in ecosystems and, consequently, in humans[4]. For this reason, much attention has been given to the effect of BPA in water. The US Food and Drug Administration (FDA) has regulated its level in food, since it is used to make polycarbonate, a hard transparent plastic used in many consumer products [5]. The Environmental Protection Agency [6][7] and the European Union, through the European Food Safety Agency (EFSA), set a permitted level of 4 µg/Kg [8]. Several methods have been proposed for the elimination of this compound in the environment. Among these, advanced oxidation processes (AOPs) are considered as a clean degradation alternative due to their use of oxidation reactions with reactive oxygen species, and can even lead to the mineralization of contaminants [9][10] [11]. The molecular singlet oxygen ($^1\text{O}_2$ ($^1\Delta_g$)) is generated through a process called photosensitization, which begins with the absorption of electromagnetic radiation by a substrate called “photosensitizer”, which is able to transfer its energy to other molecules in order to access the electrically excited triplet state. This process from molecular oxygen to the triplet of basal state is known as Type II mechanism and is shown in Figure 2 [12].

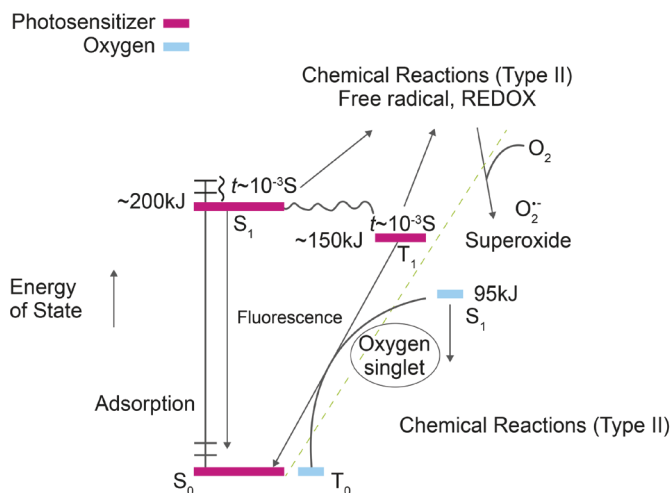


Figure 2. Generation of excited photosensitizer states and reactive dioxygen species. (Reproduced with permission from [13] Copyright (1995) Royal Chemical Society)

The success of the methodology lies in the use of a photosensitizer that has efficient inter-system crossing to the triplet state; thus, it can be established that it has a good quantum yield in the generation of singlet molecular oxygen. The mechanics of generating this excited state with oxygen require an excess of energy (approximately 94.2 KJ) for the electronic transition to take place. Colorants such as Bengal rose, methylene blue and *EOS* are some of the dyes that are considered good photosensitizers, and the last compound has a great quantum yield in ethanol around 0.57 [12][14], thanks to this characteristic, the *EOS* was used as standard photosensitizer for the photodegradation of *BPA*.

The high sensitivity chromatographic techniques have been implemented to allow the monitoring of the efficiency of the oxidation process in terms of degradation. However, these methodologies require expensive equipment and relatively large amounts of organic solvents, leading to high contaminant discharges. The application of chemometric techniques to spectrophotometric data has allowed the rapid quantification of components in complex mixtures without the need for previous separation or complex pre-treatments that involve higher costs and excessive work [15].

For this reason, chemometric methods have become very important since they make it possible to analyze mixtures and statistically ensure the quantification of these. The main advantage of PLS regression is that it allows the treatment of large volumes of data and is a modeling solution with strongly correlated variables, taking both the spectra and the concentrations, thereby reducing the uncertainty of the analysis [16]. PLS has been very popular in areas like chemical engineering, where predictive variables often consist of many different measurements in an experiment and the relationships between these variables are ill-understood[17]. PLS is widely used by chemists and chemical engineers for spectrophotometric calibration purposes. Therefore, this research work proposes the use of chemometric tools for chemistry analysis and, in this case, to improve the robustness of singlet molecular oxygen during photodegradation of *BPA* in a homogeneous system with *EOS* as photosensitizer. The PLS method was used to quantify the monitoring of *BPA* in admixture with the photosensitizer. This measurement was performed in the UV-visible spectrophotometer. The application of this

chemometric tool is important because this type of system can generate overlapping spectra that can affect quantification when only univariate calibration methods are used. It's important highlight that these methods is not conclusive, because need more data to give confidentiality at the measurements, but this research to look another alternative for the check the change in order in the photodegradation profile between *BPA* and *EOS* and use it like tool to support in the quantification of this kind of the pollutants.

2. MATERIALS AND METHODS

2.1 Reagents

99% *BPA* was purchased from Sigma Aldrich. For the standard photosensitizer, 99% purity *EOS* dye (2', 4', 5', 7' - tetrabromofluorescein) from Sigma Aldrich was used. MilliQ water was used for the preparation of the solutions.

2.2 Technique of determination

To determine the photodegradation percentage, each mixture was evaluated based on the changes in absorbance at λ_{\max} , using a single beam UV-visible spectrophotometer (Optizen Rose Scientific Ltd., Canada) with bandwidth: 2nm and cuvettes of 20mm quartz.

2.2.1 Univariate calibration

The concentration of *BPA* and *EOS* (photosensitizer) was measured individually to establish the wavelength range providing the most information about the compounds. The linear range of *BPA* and *EOS* is between 2 and 10 ppm. Each calibration curve was made in triplicate to ensure the accuracy of the analysis. The spectral runs were in the range 200-800nm, with a speed of 240 nm/min and with sweeps every 4 nm.

2.2.2 Multivariable Calibration

For the PLS analysis it is assumed that every absorption taken represents a factor (explanatory or independent variable) that affects the response variable Y (explained or dependent), which in this case is the concentration of the compounds of the mixture if the absorbances are the product of a

summation between the absorbance of *EOS* and *BPA* in the wavelength range where it absorbs *EOS*. The Beer-Lambert Law model was used for calibration standards containing *I* chemical compounds with the n-absorbance spectra [18]. For the analysis of the mixture, a spectral sweep was initially made using a 3^k factorial design with a central point, with k = number of factors. This was made in order to carry out the calibration and the validation of the chemometric model [19].

The chemometric model was valued through Leave One Out Cross Validation (LOO-CV). This is made with the purpose of identifying the components that represent the calibration data, as well as other aspects associated with the predictive capacity of the chemometric model [20]. The mathematical procedure to determine the mean square root deviation (RMSE) of the LOO-CV to select the optimal number of PLS factors is:

$$RMSE = \left[\frac{1}{(p)} \sum_{i=1}^p \left(y_i - \hat{y}_i \right)^2 \right]^{\frac{1}{2}} \quad (1)$$

Where p is the number of samples used for the compound, y_i is the known concentration, and \hat{y}_i is the predicted concentration. We used other statistical tools to analyze the samples, namely: Mean Square Error Prediction (RMSEP), that reflects all the variability involved in the model [16]; and Principal Component Analysis (PCA), defined as a procedure that converts a large number of correlated variables into new low dimension uncorrelated (even orthogonal) variables named Principal Components, which describe the greater variance as well as the scores of the data obtained [18][21].

The PLS regression transforms the matrix of predictors X of order $n \times p$, with the help of the vector of responses Y of order $n \times 1$, into a matrix of principal components [22]. This is so that, by including the effect of the response variable, it has better diagnostic capacity and is a better model for the output prediction [23]. The predictor variables are converted into orthogonal components, which represent the solution to the problem of multicollinearity and allows a reduction of the dimensionality of the space of these variables [22]. Moreover, this allows the treatment of large

volumes of data and is a modeling solution for strongly correlated variables, taking both the spectra and the concentrations and reducing the uncertainty of the analysis [16].

For the development of the mathematical model, the free software **Rstudio** was used. This has specialized packages that allow the correct analysis of this data volume. Other Rstudio packages were used such as **prcom**, for the analysis of the main components; **ggfortify**, for the analysis of the Scores and Loading graphs; **pls**, for the PLS analysis, in which **pcr** was applied for the calibration and **RMSEP** for the validation.

2.3 Preliminary tests for photodegradation

The proposed system for this reaction is a reactor of approximately 120mL (Figure 3). The reactor has an inlet at the top, specifically in the lid, for the lamp, another for the pH electrode and a third hole for taking samples with a septum. For the photodegradation of the *BPA*, a PENRAY UVP mercury lamp and the Neon Lamp were used for the initial photodegradation tests, and the direct photooxidation with *BPA* and *EOS* was analyzed with the substances individually and then mixed in a homogeneous solution. pH analysis was performed on the solution of *BPA* and evidence obtained that the pH of the solution throughout the experiment was maintained at 6.89, which is very consistent with miliQ water.

For the photodegradation analysis of *BPA*, starting from a standard solution of 40 ppm, an aqueous solution of 10 ppm of *BPA* was prepared. The sample was taken every 10 minutes, then the measurement was performed in the spectrophotometer. This test was done in triplicate to ensure the accuracy of the analysis. The wavelength chosen for the monitoring of *BPA* was at λ_{\max} : 284nm. In the case of *EOS*, a solution of 7 ppm was prepared, and the sample taken every 10 minutes. The wavelength chosen for the monitoring of *BPA* was at λ_{\max} : 524nm. For the preparation of the mixture, the same concentrations were taken and subjected to irradiation with the same lamp and monitored every 10 minutes for 100 minutes of exposure.

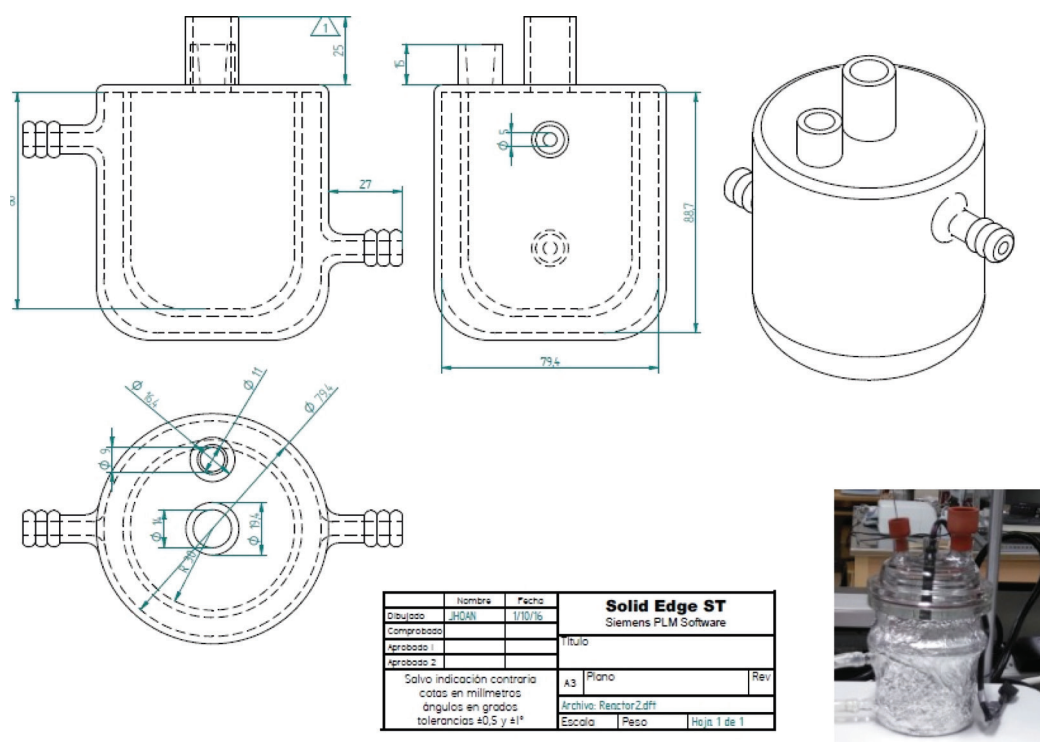


Figure 3. Reactor design used in the preliminary experiments

3. RESULTS AND DISCUSSION

3.1 Multivariate calibration.

Initially, the calibration curves for *BPA* and *EOS* were performed in a range from 3 ppm to 10ppm, which is similar to what has been reported in the literature on the study of this type of contaminants in aqueous media [24]. For the univariate data it was possible to establish a linear range with an average coefficient of determination for *BPA* of $R^2_{adj}=0.9995$ and for *EOS* of $R^2_{adj}=0.99999$. Therefore, it is possible to conclude that, with the established methodology using spectrophotometry, it is possible to quantify the compounds at low concentrations.

However, when trying to study the system in blends with the photosensitizer, it was evident that overlaps occur in some portions of the spectrum. Therefore, it became necessary to use the statistical tool PLS to determine the number of optimal factors that explain the greatest amount of information about what occurs in the reaction. This means that a spectral window is selected which will give the necessary information about photodegradation without being affected by the potential overlap.

In the Figure 4 shows that there was an overlapping portion between the *BPA* and *EOS* bands, which may give erroneous information about the photodegradation of the contaminant. Several research works exist that suggest that the best way to address these circumstances where there are mixes is the **Benesi-Hildebrand equation**. This proposes using a constant concentration of a metal and varying the concentration of the ligand ($[L]$), which results in a graph that relates to $b/\Delta A$ (ΔA is the difference between absorbance with or without ligand) against the inverse of the concentration of the ligand $1/[L]$ [25]. The value of K_f can be obtained with this linear graph by calculating the slope of the line, and with this value, the concentration of the other component of the mixture can be obtained. However, the main limitation of this type of method is that it is assumed that only one complex is formed, and it is possible that other complexes are formed in the same system that are absorbing at the same wavelength, resulting in erroneous reports. Therefore, in this case, multivariate analysis was carried out for *BPA* compared to the analysis of the mixture.

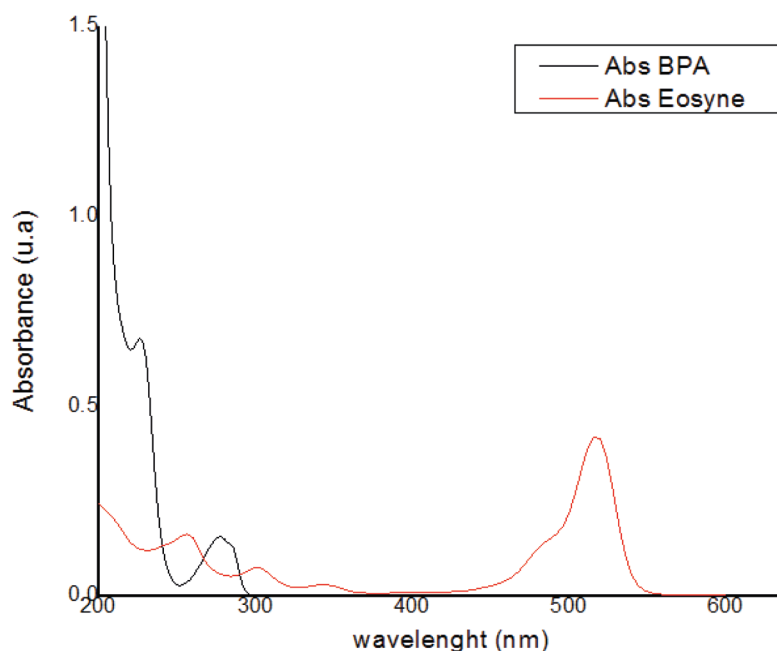


Figure 4. Absorption spectrum for the BPA and EOS

For the mathematical model, initially, it was intended to characterize *BPA* individually, to establish the window with the most information about this. Therefore, a spectral scan was performed from 200 to 800 nm, taking measurements every 4 cm^{-1} . This yielded a total of 151 items of data, which were taken to a new coordinate system defined by the main components to be defined in an array X , which would give information about the covariance of the total data. At this stage, it can be concluded that the first three principal components represent

the greatest variability of the data (approximately 97.83%), so the model is sufficiently robust to explain the behavior of *BPA*. Moreover, the most significant relation between them is concentration, which is directly related to the limit of detection. It is therefore ensured that the technique is appropriate for low levels of concentration. In addition to the PCA analysis it was established that the range with the most information about *BPA* is between wavelengths 190nm and 350nm.

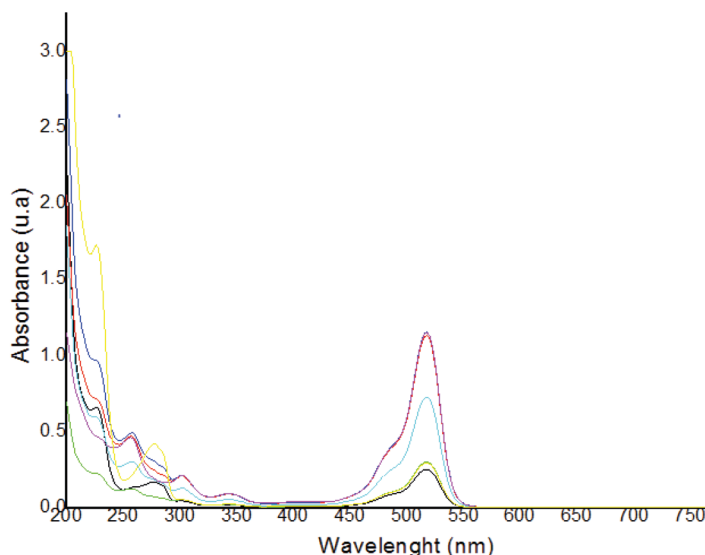


Figure 5. Absorption spectrum profile of BPA and EOS in homogeneous solution a several concentrations.

In the analysis of the mixture, 7 samples were taken with a 3^k model with a point to the center. These were taken in a scan between 200 and 800nm, taking measurements every 2 cm^{-1} resulting in a total of 301 items of data, as shown in the figure 5. In the adsorption spectra taken for the tests, *BPA* does not absorb in the portion where the wavelength of the *EOS*. This provided valuable information making it possible to establish the best range to monitor the concentrations of the mixture.

Initially, a main component analysis was performed on the samples obtained. This makes

it possible to classify and condense the different signals associated with the 301 items of data and thus obtain a resulting variable that is a linear combination associated with the original data. From the analysis of the most representative components of the experiment we can conclude that the principal components 1 and 2 represent the highest variability of the data, at 93.99%. Therefore, it is more than sufficient to use only two components for the model. Furthermore, for the model the samples with the highest correlation are those with a similar concentration of *EOS*, as shown in the Scores and Loadings diagram in Figure 6.

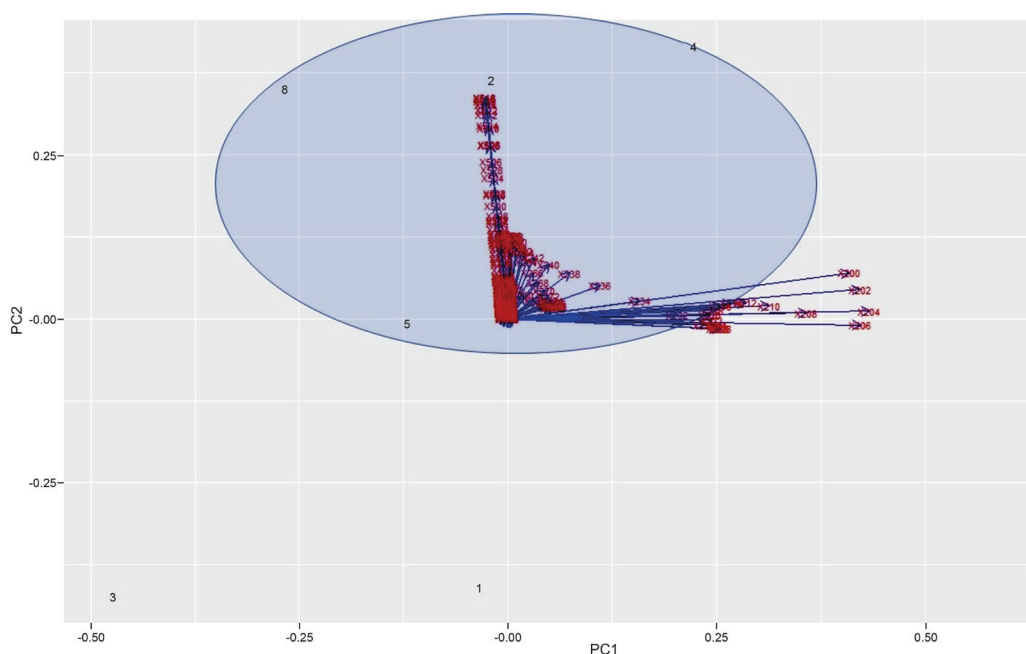


Figure 6. Plot of Scores and Loadings of the mixture (*BPA* + *EOS*) in a range of 200-800nm

In the table 1, the RMSE analysis of the calibration is shown, and its evidenced that component 2 represents the smallest variability in the experiment; in the figure 7 and the figure, shown the relationship between the measured and predicted values for both wavelengths of the compounds studied, from which a linear relationship can be intuited, although more data would be needed to confirm this.

Figure 8 shows that the number of components chosen for each response allows the generation of a good prediction model because both curves are observed to be similar. The curve analyzed at the

284 nm wavelength (associated with the maximum wavelength for *BPA*), shows that it is possible to explain the prediction model with the first two components. For the 524 nm wavelength associated with *EOS*, it shows information about the original variables in the first two components.

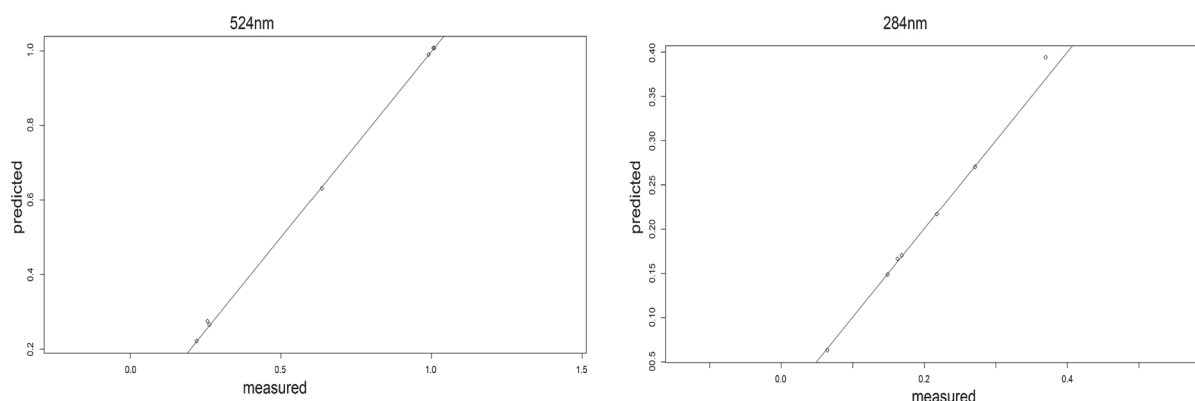


Figure 7. Graphical of the absorbances for calibration data at 284 and at 524 nm

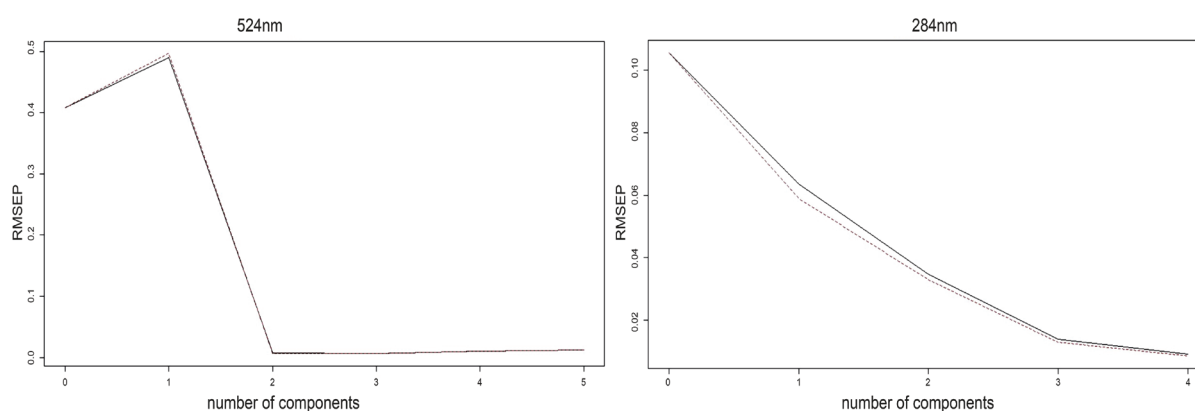


Figure 8. Graphical of the ratio with the number of components at 284 and at 524 nm

Table 1. Determination of mean square error of prediction (RMSEP)

VALIDATION: RMSEP						
Cross-validated using 5 leave-one-out segments.						
	(Intercept)	1 comps	2 comps	3 comps	4 comps	5 comps
CV	0.4084	0.4896	0.007137	0.006549	0.010773	0.01287
adjCV	0.4084	0.4964	0.006626	0.006110	0.009978	0.01192
TRAINING: % variance explained						
	1 comps	2 comps	3 comps	4 comps	5 comps	
X	63.430	99.78	100	100	100	
Abs	2.846	100.00	100	100	100	
VALIDATION: RMSEP						
Cross-validated using 5 leave-one-out segments.						
	(Intercept)	1 comps	2 comps	3 comps	4 comps	5 comps
CV	0.1057	0.07789	0.01078	0.002773	0.01235	0.01457
adjCV	0.1057	0.07210	0.01016	0.002653	0.01144	0.01349
TRAINING: % variance explained						
	1 comps	2 comps	3 comps	4 comps	5 comps	
X	63.43	99.78	100.00	100	100	
Abs	91.22	99.87	99.98	100	100	

The graphical scores in the Figure 9 were made to determine the degree of clustering and detect outlier points in the calibration data, in order to allow the construction of the model. This graph shows the percentage of variability of the main components. The data shows a wide variability. This coincides with what was obtained previously, since the values of all these possible combinations are given in the first two components, which represent approximately 100% of the proposed model, and makes it possible to be validated.

For the prediction of the model, the determination of the RMSE of the prediction for the measured and observed values in relation to the new data chosen for the validation was proposed. This process is carried out to corroborate that the prediction data of the model is adjusted to the actual occurrence in the experimental region. The absorbance is taken in the lengths 284 nm for the *BPA* and 524nm for the *EOS*.

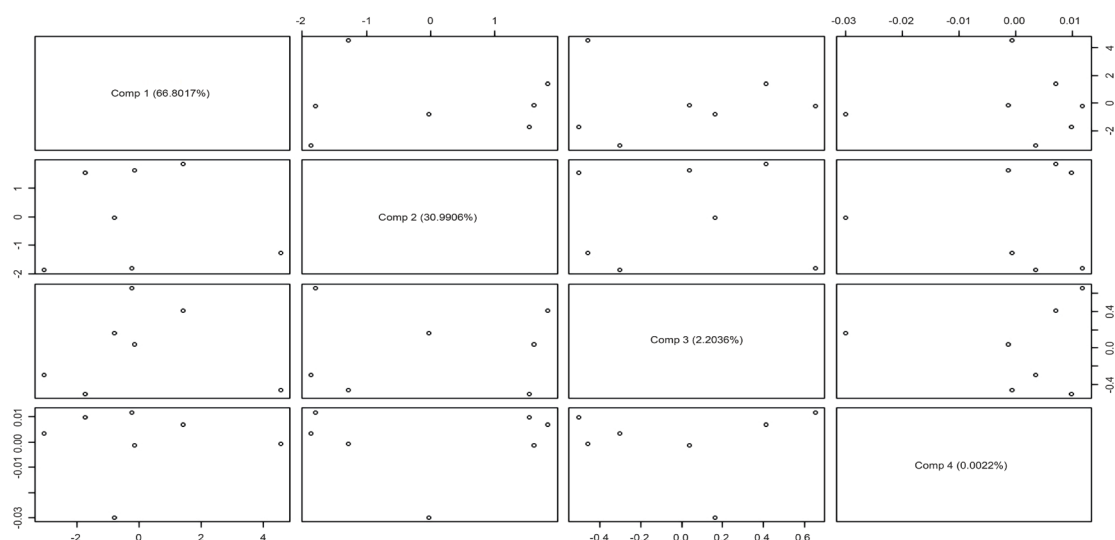


Figure 9. The Score graphical at 284 nm

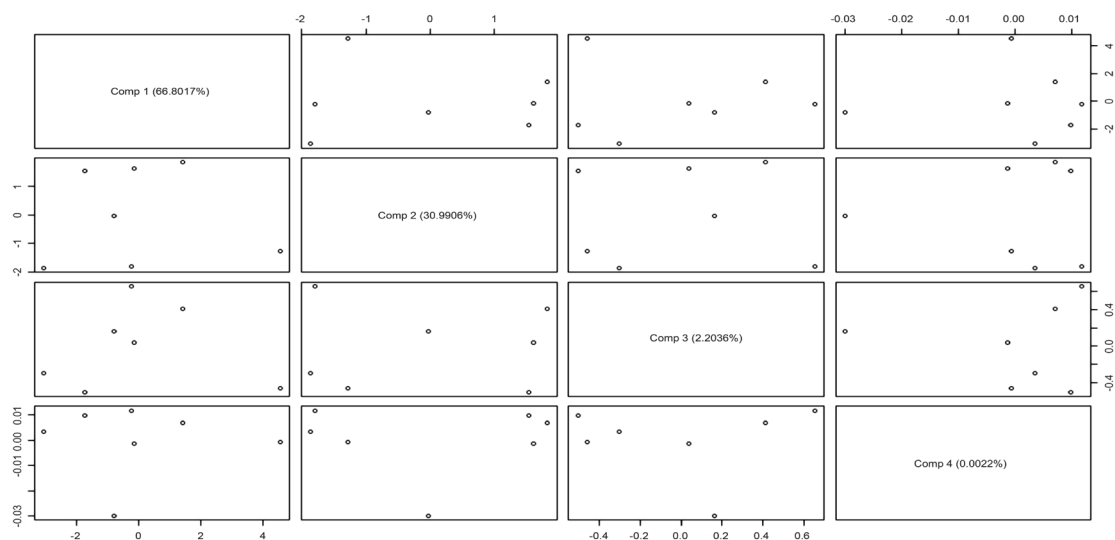


Figure 10. The Score graphical at 524 nm

Table 2. Values obtained with validation and predictions data

Experimental design 3 ^k		Measured vs predicted concentration					
Conc. BPA (ppm)	Conc. EOS (ppm)	Measured conc. (284 nm)	Predicted conc. (284 nm)	% relative error	Measured conc. (524 nm)	Predicted Conc. d (524 nm)	% relative error
10.0	2.0	11.32117	11.131127	0.1131	1.99364	2.004650	0.232
6.0	8.0	12.92701	12.940045	1.1567	7.70339	7.717685	-3.658
10.0	8.0	16.86861	16.709858	0.6710	7.93644	7.979326	-0.259
2.0	8.0	9.64234	9.602371	3.8012	7.89407	7.908689	-1.155
6.0	2.0	7.89051	7.799074	0.2998	1.99364	2.000912	0.046
6.0	5.0	10.59124	10.563029	0.7605	4.95975	4.996822	-0.064
2.0	2.0	4.31387	4.148324	1.0742	2.00424	2.015299	0.759

The values obtained for concentration are very close in the experimental and prediction measurements of the model shown in the Table 2. This is evidenced by the relative errors which do not exceed 2.2%, considering that the sample size is small. For this reason, it can be stated that either of the two wavelengths evaluated (284 nm and 524 nm) may be considered optimal for monitoring the degradation profile. However, to minimize effects associated with signal formation due to mixing, a reference wavelength of 524 nm is preferred because in this region only the photosensitizer EOS absorbs.

3.2 Preliminary photodegradation tests

For the degradation tests, a 10 ppm concentration of BPA was taken and incorporated into the reactor described in Figure 3, where it was subjected to a mercury lamp. 2 mL samples were taken every 10 minutes to evidence the effect of light on the analyte. In the first instance, it was analyzed in the dark, so the results at 110 minutes did not show any reaction. After this, direct photodegradation tests were performed for EOS and BPA individually, in the case of the BPA was observed an upper signal, this phenomenon is possible for the formation of the new sub-products of the photodegradation that can absorbed in the same region. For the other side, the photodegradation direct occurred with a time difference: In the case of EOS, complete

photodegradation was evident at 50 minutes while for BPA, changes were observed after 80 minutes of exposure totally and the BPA suffer a small photodegradation at the same time. All these results are shown in Figure 10, which shows the ratio of the concentration over time in relation to the initial concentration, versus time of exposure. It should be noted that these results are quite consistent, since the lamp used in these tests has an emission spectrum which can cover the entire absorption zone for both compounds.

After the individual analyses, the same experiment was performed with the mixture of the compounds (photosensitizer and pollutant), in order to try to explain accurately the reaction, which results were shown in Figure 11. The BPA was transformed by only 10% from 80 minutes to 120 minutes, except when the reaction was direct photo-oxidation, In the case of the EOS, the reaction was immediately and of course in the dark noting happened. Then in the figure 12, the reaction with the mercury lamp with the mixture is shown, which occurred in 10 minutes; however, in this case, the photosensitizer suffered photobleaching and the BPA change the signals, is possible for the formation the other compounds suggest transformation and consequentially photodegradation for $^1\text{O}_2$. All these experiments evidence that the hypothesis made is correct and the Mercury lamp is efficient for reaction of the photodegradation [24][26].

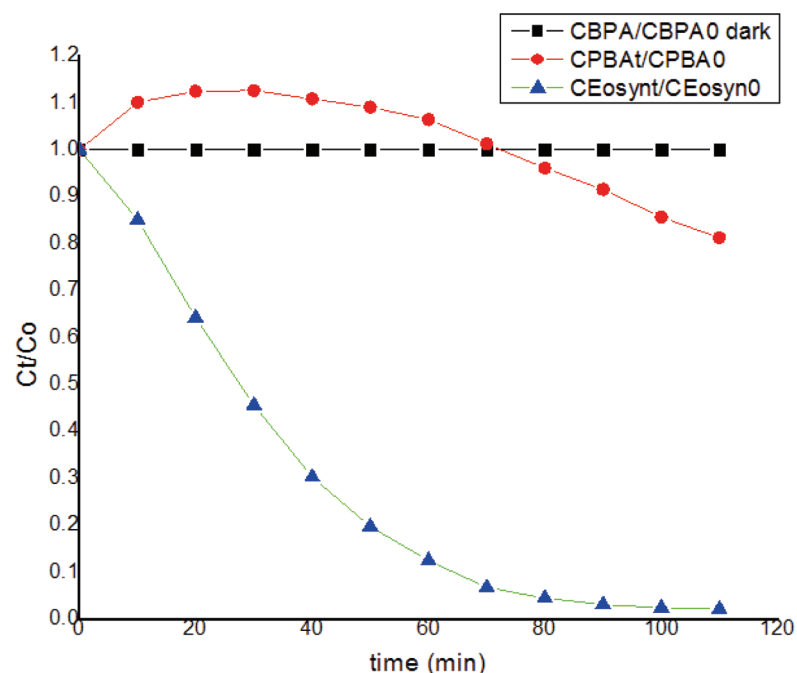


Figure 11. Photolysis of *BPA* and *EOS* using the Mercury lamp.

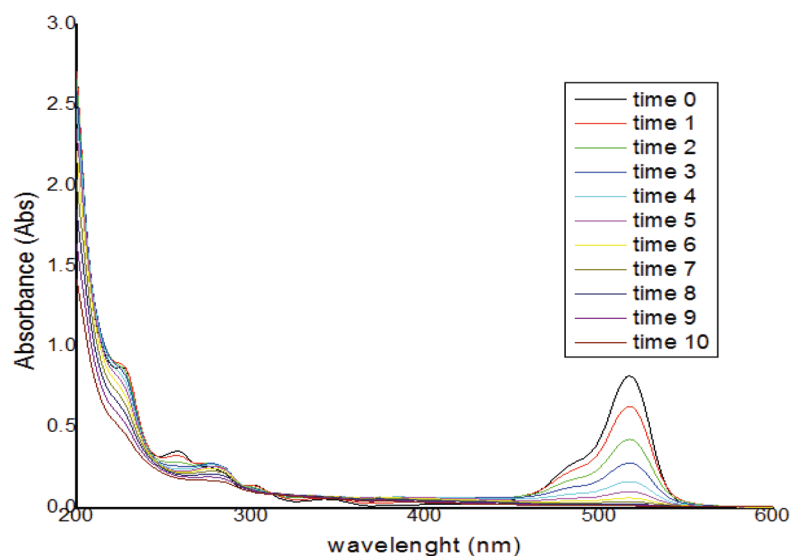


Figure 12. Photodegradation of the *BPA* + *EOS* for photosensibilization reactions

For the possible contribution of the mercury lamp in the reaction for photosensibilization, the tests were carried out with the neon lamp. The first reaction was between *BPA* and light, shown in the figure 13, there was not reaction, confirmed the hypothesis, not exist interaction directly between the pollutant and the light, for that suggest that the reaction is for other ways. The second experiment was the *EOS* with the light shown in the figure

14, there was not reaction, and the mixes between *BPA* and *EOS* exposed at the light shown in the figure 15, neither the reaction, the possible reason is that the reaction require other conditions for that happen, so, in this case it's necessary an experimental design.

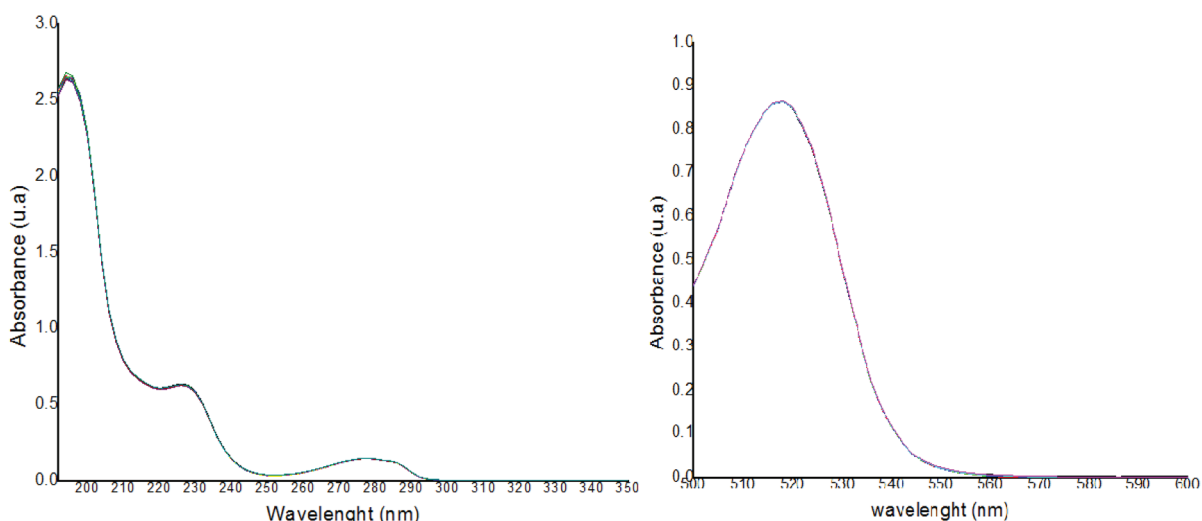


Figure 13. Photodegradation of *BPA* and *EOS* with the neon lamp monitored for 130 minutes

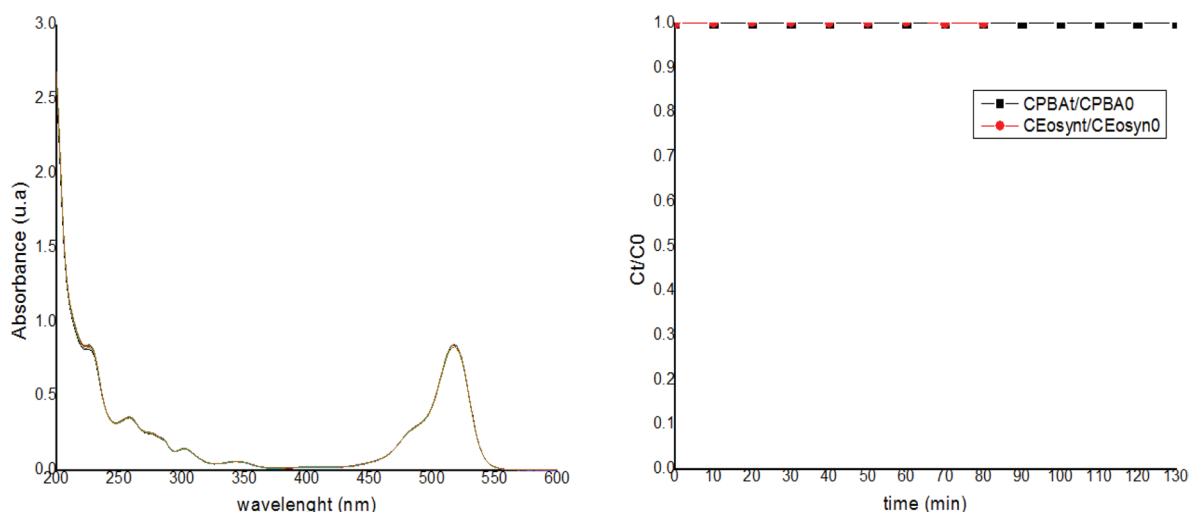


Figure 14. Photodegradation of *BPA* and *EOS* with Neon lamp

Then, for the other experiments using the neon lamp, it was possible to demonstrate that the photosensibilization reaction, in the preliminary result, was possible to be established at pH 1, concentration of *BPA* of 4ppm, of *EOS* of 8 ppm at 240 minutes, and with a photodegradation of 100%. In the figure 15 the results with water milliQ and tap water are shown and demonstrate the same results. Finally, for the analyses of the possible way of the reaction, analyze with furfuryl alcohol as scavenger of $^1\text{O}_2$, Ethanol as scavenger

of OH^\cdot , chloroform as scavenger of $\text{O}_2^{\cdot-}$, and $\text{CO}_3^{2-}/\text{HCO}_3^{-1}$ were used for the tap water at neutral and acidic pH, using as control the optimal condition. The results shown with the scavenger for the $^1\text{O}_2$ do not produce a reaction; for this reason, it was concluded that the $^1\text{O}_2$ species had a significant participation in the photodegradation.

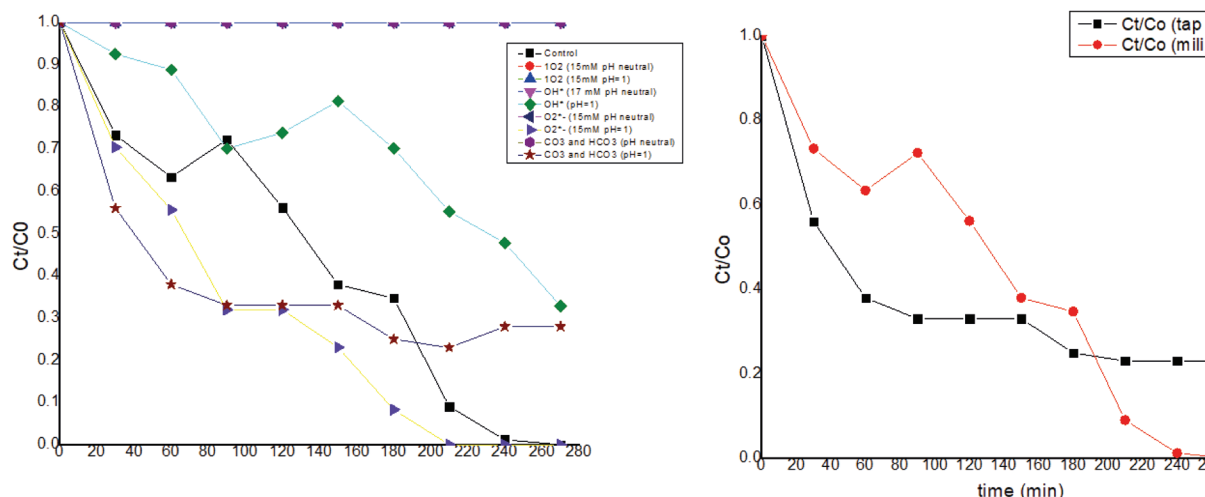


Figure 15. Kinetics of the Photodegradation for the mixed EOS and BPA (right). The experiments with the influence the species in the photodegradation with the scavenger used (left).

4. CONCLUSIONS

Considering that the results of the PCA and PLS applications were significant, the model made predictions with great statistical accuracy, this is evidenced in its RMSEP. In general terms, it is a versatile methodology aiming to reduce systematic errors, such as those caused by the traditional methodology, so that the application of chemometric tools such as PLS and PCA allow us to detect the regions with the most information and to make use of instrumental work techniques that give sufficient robustness to the results.

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