SIMULTANEOUS DETERMINATION OF PROTOCATECHUIC ALDEHYDE AND PROTOCATECHUIC ACID USING THE LOCALIZED SURFACE PLASMON RESONANCE PEAK OF SILVER NANOPARTICLES AND CHEMOMETRIC METHODS

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A simple and sensitive spectrophotometric method is proposed for the simultaneous determination of protocatechuic acid and protocatechuic aldehyde. The method is based on the difference in the kinetic rates of the reactions of analytes with $[Ag(NH_3)_2]^*$ in the presence of polyvinylpyrrolidone to produce silver nanoparticles. The data obtained were processed by chemometric methods using principal component analysis artificial neural network and partial least squares. Excellent linearity was obtained in the concentration ranges of 1.23-58.56 µg mL⁻¹ and 0.08-30.39 µg mL⁻¹ for PAC and PAH, respectively. The limits of detection for PAC and PAH were 0.039 and 0.025 µg mL⁻¹, respectively.

Keywords: protocatechuic aldehyde; protocatechuic acid; chemometric methods.

INTRODUCTION

Protocatechuic acid (PAC) and protocatechuic aldehyde (PAH) are the most effective active ingredients in some of the Chinese conventional herbal drug such as Salivia miltorrhrza,¹ Fufang-Pugongying-Mixture,² Shuangdan granule,³ and Zaoren-an-shen capsules.⁴ These phenolic compounds are known as the strong antioxidants and free radical scavengers.^{5,6} They can improve blood circulation,⁷ decrease the area of cerebral infarct,⁸ and inhibit the rennin angiotensin system.⁹ Therefore, in order to study its clinical effects, it is significant to develop an easy, sensitive and fast method for the simultaneous determination of PAC and PAH in pharmaceutical preparations. So far, the commonly used techniques for simultaneous determination of PAH and PAC in pharmaceutical preparations are chromatography (TLC and RP-HPLC)^{2,10} and capillary electrophoresis with electrochemical, UV-visible absorbance and chemiluminescence detectors.¹¹⁻¹³

Silver nanoparticles (Ag-NPs) are among the most widely studied nanomaterials due to their attractive physicochemical properties. One of the most interesting characteristics of these nanoparticles is due to their ability to produce localized surface plasmon resonance (LSPR).^{14,15} This phenomenon occurs when the incident oscillating electromagnetic field of the light interacts with the conduction electrons of the metal nanoparticles. This interaction causes a collective coherent oscillation of the nanoparticles free electrons in respect to the positive metallic lattice. This process is at resonant at a specific frequency of the light and is termed LSPR.16 Therefore, the Ag-NPs display a strong absorption band around 450 nm with an extremely large molar extinction coefficient ($\sim 3 \times 10^{11}$ mol L⁻¹ cm⁻¹). Recently, the gold and silver nanoparticles have been used as a colorimetric sensor for the detection of tryptophan¹⁷ and glucose¹⁸ by the naked eye. Colorimetric nanoprobes using plasmon absorbance of noble metal NPs have also been developed for the sensitive determination of neurotransmitter,19 uric acid,20 phenolic compounds,21 ascorbic acid,22 thiol-containing amino acids,23 DNA,24 and pesticides.25

Kinetic methods based on the difference in the reaction rates are often useful for the simultaneous determination of the analytes in multicomponent mixtures without any physical separation. Generally, in these methods, similar analytes react with a common reagent, the difference in the reaction rates is monitored and with the help of the chemometric methods the components are simultaneously determined.^{26,27}

The proportional equations, curve resolution,^{28,29} and Kalman filter³⁰ are among the chemometric methods applied to resolve the multicomponent kinetic systems. Linear and nonlinear chemometric methods, including principal component regression (PCR), partial least squares (PLS), and artificial neural network (ANN), have also been widely used for the simultaneous determination of analytes in the mixtures by means of kinetic-spectrophotometric procedures.³¹⁻³⁴

The ANN that stimulates the biological neural network has layered structure. It is consists of several neuron layers including input, output, and at least one hidden layer. Synapses connect the layers together form the input to the output through hidden layer. Among various types of neural network structures, the feed forward artificial neural network (FFANN) is the most popular and applicable types of ANNs in analytical chemistry and was used in this work. Basic theory and application of this method to chemical problems have been explained in the literature.35-37 However, with large number of input data, training process is complex and very large amounts of calibration sample is needed. So, for the spectrum and kinetic data, a method for reduction of input data is necessary. Principal component analysis (PCA), a powerful multivariate technique, rotates the data and reduces the dimensionality of them while retained the maximum information, thus, principal components (PCs) can be useful inputs for ANN.38 The use of the PCA data in the design of the ANN model have the advantages of decrease the effect of adjustments of weights variation, increase the numerical stability of the model construction process and reduce the amount of colinearity between variables.39

PLS and ANN as two known linear and nonlinear chemometric techniques have the ability to relate the input and output variables without the need for having any preknowledge on the system.⁴⁰ PLS modeling is one of the powerful multivariate techniques in which

the building of regression models is based on the latent variable decomposition that takes into account the information of responses and concentrations simultaneously and is widely used for the simultaneous determination of species which have spectrum overlap. Its basic concept was originally described by Wold *et al.*⁴¹ The theory and application of PLS methods have been discussed by several workers.⁴² The chemometric resolution of the multicomponent kinetic systems has the advantages that the details of kinetic of reactions such as the order of the reaction and the rate constants are not needed. Furthermore, the analyte-analyte interaction, the synergistic effect and the multistep process is not important comparing to the conventional techniques.⁴³

In this work, a simple and sensitive kinetic spectrophotometric method has been reported for the simultaneous determination of PAC and PAH. The method is based on the difference of the kinetic rates of the analytes with silver nitrate in the presence of polyvinylpyrrolidone (PVP) to produce Ag-NPs. Calibration models have been built from the kinetic data derived from the analyte mixtures with the aid of principal component artificial neural network (PC-ANN) and PLS. Then, these models have been verified and compared. Finally, PC-ANN has been selected and applied for the simultaneous determination of analytes in real samples.

EXPERIMENTAL

Chemicals and reagents

All chemicals used were of the analytical reagent grade from Merck (Darmstadt, Germany) in the highest purity available, except for polyvinylpyrrolidone (PVP) with average molecular weight of 10,000 that was purchased from Acros (Geel, Belgium). The aqueous solutions were prepared using distilled deionized water. The experiments were performed at the ambient temperature $(25 \pm 2 \,^{\circ}\text{C})$. A stock solution of AgNO₃ $(1.0 \times 10^{-3} \text{ mol L}^{-1})$ was prepared by dissolving an appropriate amount of AgNO₃ in distilled water. A solution of polyvinylpyrrolidone (PVP) (0.4 g L⁻¹) was prepared daily by dissolving 0.01 g of PVP in distilled water and diluting to 25 mL. Fresh standard solutions of analytes were prepared daily by dissolving an appropriate amount of reagents in deionized water.

Apparatus

The UV-Vis absorbance spectra were recorded on a HACH model DR 3900 (Loveland, Colorado, USA) spectrophotometer with a 1.0 cm glass cell. The measurements of the pH were made with an EcoMet model P25 pH meter (Guro-gu, Seoul, Korea) equipped with a combined glass calomel electrode. The sizes of the Ag-NPs were characterized by the transmission electron microscopy (TEM) using a Zeiss transmission electron microscope (Jena, Germany) operated at an accelerating voltage of 80 kV. A PLS program for the calibration prediction and the experimental design was written in MATLAB according to the algorithm described by Martens and Næs⁴⁴ and the PLS routine of PLS Toolbox (Eigenvector, version 2.1). The principal component analysis (PCA) process was performed also by the PLS Toolbox, and the scores of the PCA running were fed into ANN program as inputs. The ANN calculations were performed using WinNN32 software.

General procedure

In a 5.0 mL volumetric flask the solutions were added in the following order: 3.0 mL of 1.0×10^{-3} mol L⁻¹ AgNO₃, 0.1 mL of 0.25 mol L⁻¹ NH₃, 0.7 mL of 0.4 g L⁻¹ PVP and 0.1 mL of 0.1 mol L⁻¹ NaOH and was mixed thoroughly. Then, an appropriate volume of PAC or PAH standard or sample solution was added. The mixture was diluted to 5.0 mL with distilled water and was mixed slowly. It was transferred into a 1 cm spectrophotometric cell and the absorbance was recorded against time for 30 min in the 0.1 min interval. The absorbance was measured against a reagent blank at 450 nm, the λ_{max} of silver nanoparticles localized surface plasmon resonance peak.

RESULT AND DISSCUSION

Kinetic study of reaction between PAC and PAH and $[Ag(NH_3)_2]^+$

The preliminary studies indicated that PAC and PAH can reduce $[Ag(NH_3)_2]^+$ complex to the Ag-NPs (Scheme 1) in the presence of polyvinylpyrrolidone (PVP) as the stabilizer, at an alkaline medium.

In the absence of analytes, no absorption peak was observed in the UV-Vis region. But, after addition of analytes, the $[Ag(NH_3)_2]^+$ was reduced to the Ag-NPs and the localized surface plasmon resonance peak of Ag-NPs was observed (Figure 1). TEM image showed that (Figure 1S) the synthesized Ag-NPs with an average diameter about 10-20 nm were dispersed in aqueous solution. The plot of the absorbance versus time for analytes (Figure 2) revealed that the reaction of silver ions with PAC is faster than that with PAH. Therefore, a kinetic spectrophotometric method based on the difference between



Figure 1. LSPR spectra of Ag-NPs formed by PAC. Conditions: $AgNO_3$ 6.0 × 10⁻⁴ mol L⁻¹, NaOH 2.0 × 10⁻³ mol L⁻¹, NH₃ 5.0 × 10⁻³ mol L⁻¹, PVP 6.0 × 10⁻² g L⁻¹, PAC 46.25 µg mL⁻¹



Scheme 1. Reduction process of $[Ag(NH_3)_2]^+$ by the PAC and PAH



Figure 2. Kinetic curves corresponding to the absorbance changes (at $\lambda = 450 \text{ nm}$) of the analytes stimulated growth of Ag-NPs. Conditions: AgNO₃ 6.0 × 10⁴ mol L⁻¹, NaOH 2.0 × 10³ mol L⁻¹, NH₃ 5.0 × 10³ mol L⁻¹, PVP 6.0 × 10⁻² g L⁻¹, PAC 46.25 µg mL⁻¹, PAH 30.38 µg mL⁻¹

the reaction rates of PAC and PAP with $[Ag(NH_3)_2]^+$ was designed for the simultaneous determination of these analytes.

It should be noted that in the presence of an excess amount of silver nitrate, there is no remarkable red or blue shift of the LSPR peak with different concentrations of the analytes.

Optimization of experimental factors

Type and concentration of the stabilizer, concentrations of silver nitrate, NaOH and NH_3 are the main factors that must be optimized for improvement of the sensitivity of the method. So, the effect of these parameters on the maximization of the LSPR peak at 450 nm were carefully studied and optimized, individually.

The choice of an appropriate stabilizing agent is very important in the syntheses of nanoparticles as the stabilizing agent prevents the nanoparticle agglomeration. Nanoparticles are often stabilized according to the two basic modes: electrostatic and steric stabilization.²⁰ Electrostatic stabilization is the result of the production of charges on the surface of nanoparticles whereas the steric stabilization is achieved when organic molecules or polymers act as the protective shields on the metallic surface (e.g., PVP). In this study PVP, CTAB, Triton X-100 and sodium citrate were examined as the stabilizing agent to prevent the Ag-NPs agglomeration. Figure 3 shows that with the PVP as the stabilizer, the absorbance and consequently the sensitivity of method is higher than with the other stabilizing agents examined. The higher sensitivity of the method in presence of PVP as the stabilizer can be related to its macromolecular structure with N-heteroatom, forming a thick adsorption layer on the surface of the generated Ag-NPs. Thus, it significantly decreases the availability of Ag metal surface, where the catalysis of further silver ions from the solution proceeds. In other words, strong interaction of PVP with the growing Ag-NPs, revealed the most significant deceleration of the studied reaction. The results show (Figure 2S) that the extinction of Ag-NPs resulted from both analytes increased with an increase in the PVP concentration and reaches its maximum at the PVP concentration of 6.0×10^{-2} g L⁻¹ PVP. Similar results were observed in our previous study.⁴⁵ Therefore, the concentration of 6.0×10^{-2} g L⁻¹ was chosen as the optimum concentration of PVP for the subsequent studies.

The effect of AgNO₃ concentration on the intensity of LSPR peak of the Ag-NPs was investigated by varying the concentration in the range of $1.0-8.0 \times 10^{-4}$ mol L⁻¹. It was found (Figure 3S) that the absorbance intensity increases by addition of AgNO₃ concentration up to 6.0×10^{-4} mol L⁻¹. In higher concentrations of AgNO₃ some black precipitates was formed in the solution, which were probably





Stabilizer agent

Figure 3. Effect of stabilizer type on the extinction of Ag-NPs. Conditions: AgNO₃ 6.0 × 10⁻⁴ mol L⁻¹, NaOH 2.0 × 10⁻³ mol L⁻¹, NH₃ 5.0 × 10⁻³ mol L⁻¹, PAC 26.21 µg mL⁻¹, PAH 27.62 µg mL⁻¹

micro-scale silver particles,²⁰ and caused a decrease in LSPR peak intensity. So, a concentration of 6.0×10^{-4} mol L⁻¹ was selected as the optimum silver nitrate concentration for the subsequent studies.

As was stated before, the proton is produced during the reaction of analytes with silver ions and the redox potential of PAC and PAH decreases with an increase in the pH; thus the reducing power of analytes is enhanced in the alkaline medium. Therefore, the effect of the alkalinity of the solution on the reactions was studied by varying the concentration of NaOH within the range of 0.5×10^{-3} - 3.0×10^{-3} mol L⁻¹. It was observed that (Figure 4S) the absorbance increased by an increase in the NaOH concentration up to a 2.0×10^{-3} mol L⁻¹ and was decreased at higher concentration. The decrease of absorbance at higher concentration of 2.0×10^{-3} mol L⁻¹ of NaOH was selected as the optimum for further studies.

The effect of ammonia concentration on the absorbance of Ag-NPs was also investigated. The results demonstrated that (Figure 5S) the absorbance increased with an increase in NH₂ concentration and reaches its maximum at 5.0×10^{-3} mol L⁻¹. The decrease in the signal at concentration higher than 5.0×10^{-3} mol L⁻¹ may be related to the lower yield of the redox reaction between $[Ag(NH_3)_2]^+$ complex ion and the analytes at higher concentrations of ammonia. This decrease can be related to the lower yield of the redox reaction between [Ag(NH₃)₂]⁺ complex ion and analytes at higher concentrations of NH₃. As can be understood from Nernst equation for electrode potential of $[Ag(NH_3)_2]^+/Ag$, with increasing the concentration of NH₃ the potential will decrease. So, the difference between the redox potentials of [Ag(NH₃)₂]⁺/Ag and analytes and the equilibrium constant for the redox reaction would also decrease.20 Thus, a concentration of 5.0×10^{-3} mol L⁻¹ of NH₃ was selected as the optimum NH₃ concentration.

Analytical performance

The LSPR spectra of the Ag-NPs formed upon treatment with different concentrations of the PAC and PAH are shown in Figures 4a and 4b, respectively. The figures of merit of the method under the optimum experimental conditions (Table 2S) were as follows: The calibration curve for PAC was linear in the range of 1.23-58.56 μ g mL⁻¹ with R² = 0.9976 and for PAH was linear in the range of 0.08-30.39 μ g mL⁻¹ with R² = 0.9972. The limit of detection (LOD) of PAC and PAH defined as 3S_b/m³⁵ (where S_b and m are the standard deviation of the blank and the slope of the calibration curve) were 0.039 μ g mL⁻¹ and 0.025 μ g mL⁻¹, respectively. The relative standard deviation (RSD)



Figure 4. LSPR spectra of Ag-NPs in the presence of different concentration of (A) PAC (1.24-58.56 μ g mL⁻¹), (B) PAH (0.08-30.38 μ g mL⁻¹) (The inset is a calibration curve of PAC and PAH obtained by measuring the absorbance at 450 nm). Conditions: AgNO₃ 6.0 × 10⁻⁴ mol L⁻¹, NaOH 2.0 × 10⁻³ mol L⁻¹, NH₃ 5.0 × 10⁻³ mol L⁻¹, PVP 6.0 × 10⁻² g L⁻¹

for five replicate determinations at 14.50 μ g mL⁻¹ of PAC and PAC were 3.1% and 1.8%, respectively.

Chemometric methods

In order to obtain the optimum conditions from the minimum experimental trials, 45 samples of binary mixtures were designed. The concentration levels for the analytes were as: PAC (1.24, 6.16, 15.42, 26.20, 35.45, 46.24 and 58.56 μ g mL⁻¹); and PAH (0.08, 0.72, 2.76, 8.28, 13.82, 20.72 and 30.38 μ g mL⁻¹). The binary mixtures were analyzed according to the given procedure and the kinetic data were recorded. Then, PLS and PC-ANN were tested for the construction of calibration model. In this network, the data move only in the forward direction from the input nodes through the hidden one to the output nodes without any loops or cycles and a three-layer network was applied.

For doing the PC-ANN analysis, the 45 calibration samples were randomly divided into two sets; training set (containing 35 trials (Table 1S)) and validation set (containing 10 trials) manually. The neural network was constructed using WinNN32 software and effective factors such as number of principal components, number of layers, number of hidden layers, and number of nodes (neurons) in hidden layers and learning function and learning parameters were optimized for each for each analyte individually. The results of optimization of PC-ANN parameters and the obtained statistical parameters for test set are summarized in Table 1. Another eight synthetic mixtures of the two analyte compounds were used as prediction or test set and the constructed PC-ANN model was applied to its analysis, the results of this study are shown in Table 2.

All 45 calibration samples were mean centered and used for the construction of PLS-1 model. One of the most important parameters in PLS is the number of significant factors. To ascertain the number of significant factors (latent variables) in the PLS algorithm without over fitting, leave-one-out cross-validation method was used.⁴⁶ One reasonable choice for the optimum number of latent

 Table 1. Optimized parameters used for construction of PC-ANN in simultaneous determination of PAC and PAH

Network parameters	PAC	PAH
Number of layers	3	3
Input nodes (number of PCs)	4	5
Hidden nodes	6	6
Hidden layer transfer function	sigmoid	sigmoid
Output nodes	1	1
Output layer transfer function	sigmoid	sigmoid
Momentum	0.04	0.02
Number of iteration	12.000	10.000

Table 2. Analysis of the eight synthesis samples and their relative error (RE), RMSEP and R² at the optimum conditions

РАН				PAC					
Added (µg mL ⁻¹) —		Found (µg mL ⁻¹)				Found (µg mL ⁻¹)			
	PLS	RE	PC-AANN	RE	Added (µg mL ·)	PLS	RE	PC-AANN	RE
1.077	1.627	-0.338	1.056	0.019	7.089	5.455	0.299	7.164	-0.011
1.077	1.624	-0.336	1.071	0.005	38.314	33.084	0.158	38.118	0.005
14.778	15.609	-0.053	14.237	0.038	7.089	5.204	0.362	7.114	-0.003
14.778	15.606	-0.053	14.785	0.002	38.314	32.831	0.167	38.156	0.004
22.099	22.749	-0.028	21.685	0.019	7.089	5.395	0.313	7.022	0.009
22.099	22.745	-0.028	22.052	0.002	38.314	33.024	0.160	38.456	-0.003
28.635	26.314	0.088	28.642	0.001	2.366	1.653	0.431	2.345	0.008
28.635	26.892	0.064	28.526	0.003	45.212	39.624	0.141	45.024	0.004
Statistical parameters									
RMSEP	0.125		0.00281			0.146		0.00627	
\mathbb{R}^2	0.852		0.997			0.824		0.996	

variables (LV) is a number that yields the minimum root mean square errors of cross-validation (RMSECV). As it was observed that (Figure 6S) 12 latent variables was selected for both analytes for further analysis. Then, the PLS models were used for analysis of the eight synthetic mixtures of prediction or test set. The results are also summarized in Table 2.

Performances of the methods were tested with the R² and root mean square errors of prediction (RMSEP) using the following equation:

$$R^{2} = \frac{\sum_{i=1}^{n} (y_{i} - \overline{y})^{2}}{\sum_{i=1}^{n} (y_{i} - \overline{y})^{2}}$$
$$RMSEP = \sqrt{\sum_{i=1}^{n} \frac{(y_{i} - y_{i})^{2}}{n}}$$

where \hat{y}_i , \overline{y} and y_i are the estimated output, average of output and the actual output sets, respectively; and *n* is the number of the samples in each set. The closeness of R² and RMSEP to one and zero respectively, indicate that the method is suitable for data analysis. In the PC-ANN model with the 45 calibration samples, R² for PAC and PAH were 0.996 and 0.997 and RMSEP were 0.00627 and 0.00281, respectively, whereas in the case of PLS model, R² for PAC and PAH were 0.824 and 0.852 and RMSEP were 0.146 and 0.125, respectively. Thus, the PC-ANN model is more suitable for analysis of these data. Furthermore, the results of Table 2 clearly indicate that the accuracy of PC-ANN in analysis of the eight synthetic mixtures is much higher than the PLS model. This is probably due to the large numbers of latent variable that is inevitable for both PLS models because of the interaction of the analytes. In other words, it seems that these interactions can be best described by the systems that have more nonlinearity. Consequently, PC-ANN was selected for further studies.

Study of interferences

The effect of some potentially interfering substances on the determination of a mixture of 26.21 μ g mL⁻¹ PAC and 27.62 μ g mL⁻¹ PAH by the proposed method was examined at the optimum conditions. The tolerance limit was defined as the maximum concentration of the interfering substances that give a relative error of less than ±5%. The results revealed that the glucose and lactose at mole ratio of 800; Na⁺, K⁺, Mg²⁺, SO₄²⁻, and NO₃⁻ at mole ratio of 400; L-cysteine at mole ratio of 300; Cu²⁺, Zn²⁺and CO₃²⁻ at mole ratio of 250; sucrose, Ca²⁺ and Al³⁺ at mole ratio of 200 and uric acid and Fe³⁺ at mole ratio of 100 had no significant interference in the determination of analytes.

Application

The feasibility of the technique for the determination of PAC and PAH in real sample was examined. In order to have a sample resembling the drug, the Ringer's injection solution a nonpyrogenic and sterile serum solution containing common electrolyte with a variety of clinical uses including fluid and electrolyte replenishment, was spiked with different concentration of PAC and PAH and the procedure with the PC-ANN model given above was followed. The results of eight spiked samples are given in Table 3. The recoveries of the added analytes are good (95.3 to 103.8%) indicating the suitability of the method for simultaneous determination of PAC and PAH in this type of sample.

CONCLUSIONS

In this research, a novel kinetic-spectrophotometric method combined with chemometrics interpretation of the results for the simultaneous determination of PAC and PAH based on LSPR peak of Ag-NPs was reported. The method is based on the different rates of the

 Table 3. Determination of PAC and PAH in Ringer's injection solution by the PC-ANN method

Sample —	PAC (µg mL ⁻¹)			PAH (µg mL ⁻¹)		
	Added	Found ^a	Recovery (%)	Added	Found ^a	Recovery (%)
1	1.82	1.81 ± 0.03	99.4	1.08	1.06 ± 0.03	98.1
2	16.95	$17.2\ 1\pm 0.05$	101.5	1.08	1.03 ± 0.01	95.3
3	1.82	1.80 ± 0.02	98.9	14.79	14.36 ± 0.06	97.1
4	16.95	16.72 ± 0.02	98.6	14.79	14.26 ± 0.04	96.4
5	1.82	1.89 ± 0.03	103.8	22.10	22.16 ± 0.01	100.3
6	16.95	17.32 ± 0.08	102.2	22.10	22.02 ± 0.02	99.6
7	40.21	40.46 ± 0.06	100.6	28.54	28.11 ± 0.05	98.5
8	52.31	51.98 ± 0.07	99.4	28.54	28.65 ± 0.02	100.4

^a The results are mean of five measurements ± standard deviation.

Table 4. Comparision of figures of merit of the present method with other reported methods for determination of PAC and PAH

Method	PAC (µg	; mL ⁻¹)	РАН (µg	Dof	
	LDR ^b	LOD ^c	LDR ^b	LOD ^c	- Kei
RP-HPLC	2-200	-	2-200	-	2
Electrochemical ^a	0.50-100.0	0.25	0.25-100.0	0.10	11
UV-Vis ^a	5-65	-	4-80	-	12
Chemiluminescence ^a	0.03-1.54	0.008	0.08-4.14	0.009	13
LSPR	1.23-58.56	0.039	0.08-30.39	0.025	This work

^a Electrophoresis technique with implicit detector. ^b Linear dynamic range. ^c Limit of detection.

reactions of PAC and PAH as the reducing agents with [Ag(NH₃)₂]⁺ complex ion, in alkaline medium and in the presence of PVP as the stabilizing agent. It was shown that the PC-ANN model is more reliable than PLS for this study. Some of the figures of merit of the proposed method is compared with some other reported methods for the simultaneous determination of PAC and PAH (Table 4). The LOD of the proposed method for the PAH is lower than the most of other methods^{2,13} and the linear dynamic range for the analytes is wider than the capillary electrophoresis with on-line chemiluminescence detection.¹³ However, in compare to the RP-HPLC,¹⁰ electrochemical¹¹ and UV-Vis,¹² the dynamic linear range of LSPR technique is narrower. This can be related to the shift in the LSPR peak due to the speed of formation and consequently the change in size and shape of the nanoparticles at high and low concentration of analytes. The main advantages of the proposed method are simplicity, sensitivity, low cost, and ease of operation.

SUPPLEMENTARY MATERIAL

Figures 1S to 6S and Tables 1S to 2S are freely available at http:// quimicanova.sbq.org.br/ in pdf format.

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