

PARTIAL LEAST SQUARES–NEAR INFRARED SPECTROMETRIC DETERMINATION OF ETHANOL IN DISTILLED ALCOHOLIC BEVERAGES

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ABSTRACT. In this study, using near infrared (NIR) spectrometry–partial least squares (PLS) calibration model was used for the determination of ethanol content in distilled alcoholic beverages. In NIR, 1660–1720 nm, the PLS modeling and analysis made use of 24 standards which contain 2–15% (w/w) ethanol with 0.1–1% (w/w) methanol. One-half was used for calibration and the other half for validation. Derivative, mean centering and subtracting minimum value were used as data treatment techniques for noise reduction and baseline correction. Mean centering has given the best partial least squares model with $R^2 = 0.999$ and root mean square error of prediction (RMSEP) = 0.06 for ethanol and $R^2 = 0.929$ and RMSEP = 0.08 for methanol, respectively. The percentage recovery obtained for ethanol and methanol ranges from 97–98% (w/w). The developed method produced good agreement with reference method, gas chromatography.

KEY WORDS: Distilled alcoholic beverages, Ethanol, Methanol, Near infrared spectrophotometry, Partial least squares, Chemometry, Gas Chromatography

INTRODUCTION

Ethanol is the major constituent of alcoholic beverages especially for distilled alcoholic beverages. The analytical control of ethanol is required for health and economic reasons, owing to the special taxes imposed on alcohol [1, 2]. With methanol, the high toxicity following its ingestion, inhalation or percutaneous absorption makes necessary its careful control in beverages and also in cosmetics [1].

The level of alcohol in alcoholic beverages has been determined using densitometry and pycnometry [3, 4], or by other analytical techniques such as refractive index [3, 4], oxidation of distillate [3, 4], dichromate oxidation [3, 4], enzymatic [3], biosensor [3, 5], potentiometry [3, 6], gas chromatography [3, 7], capillary electrophoresis [3], high performance liquid chromatography [3], beer analyzer [3, 7], colorimetric method [8] and flow injection analysis [9–11]. However, the techniques employed are tedious, expensive, time consuming and require sample preparation [12].

Infrared spectroscopy based methods are recently emerging because of their versatility, efficiency, being cost effective, fast and non-invasiveness [13, 14]. However, there are two major problems. First of all, the components under investigation are chemically very similar and therefore display similar IR absorption characteristics. The second problem is that the dominating absorption of ethanol, water and in some cases, sugars strongly influences the determination of one of these components in the presence other two components [15]. Conversely, since chemometrics provides powerful tools for extracting analytical information from a large amount of data produced [1] infrared spectroscopy combined with multivariate data analysis has been developed for the rapid quantitative analysis of the most important compounds in wines [16], spirit drinks and beer in official food control [17]. Near infrared (NIR) and mid infrared (MIR) are the two techniques which have been used, with NIR methods dominating [1, 18].

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NIR spectral region from 714 to 2500 nm [19] was employed for simultaneous determination of methanol and ethanol contents in gasoline and fermented beverages [17, 19]. But, the PLS regression models were developed using the spectral range from 1105 to 1682 nm [19], 2000 to 10,800 nm [16] and 2220 to 2354 nm [20] in different studies.

The band between 1120 and 1270 nm refers to the 2nd overtone of the C–H bond stretching from the diverse hydrocarbons and the alcohols. Another band, also wide, between 1350 and 1670 nm, results from the overlapping of two absorption bands. The first band, located between 1350 and 1550 nm, is related to first overtone of the combination band from C–H + C–H and C–H + C–C stretching and, the second band, between 1400 and 1670 nm, is related to the first overtone of O–H stretching from the alcohols (methanol and ethanol) [17, 19]. In the region below 1449 nm, methyl (–CH₃) and methylene (–CH₂–) groups absorption, coming from the different hydrocarbons and alcohols present in the beverages, are observed [17, 19]. It is also possible to observe, absorption above 1650 nm refers to aromatic hydrocarbons first overtone of C–H + C–H and C–H + C–C stretching, which is overlapped by absorptions of O–H stretching of alcohols [17, 19]. At 1790 nm the first C–H stretching overtone occurs. Another strong band characteristic of alcoholic beverages occurs at 1950 nm. This band is related to the combination of the moves of stretching and angular bending of the water OH. In the vicinity of 1160 nm, carbonilic and aromatic compounds usually exhibits its second overtone absorption [17].

PLS–FTIR has been employed to determine sucrose in fruit juices and glucose, total protein, cholesterol, urea and uric acid in blood, in pharmaceutical analysis; it allows the simultaneous determination of paracetamol, acetylsalicylic acid, caffeine, etc [1] and for simultaneous determination of methanol and ethanol contents in gasoline [19, 21] and beverages [21, 22].

Fourier transform mid-infrared spectrophotometry (FT-MIR), 1180–950 cm⁻¹, and near-infrared spectrophotometry (NIR), 1720–1660 nm, have been used complementarily for the direct determination of ethanol and methanol in distilled alcoholic beverages [23]. In mid-infrared, ethanol and methanol were identified separately; hence, it has been used to confirm the absence or presence of methanol. In the absence of methanol, both methods were used to determine the alcoholic strength independently; however, near-infrared was used without diluting the samples. The reported methods are simple, fast, precise, and accurate [23].

The direct determination of ethanol in traditional fermented alcoholic beverages using mid infrared spectroscopy with partial least squares regression (MIR-PLS) has also been reported [24]. The MIR-PLS predicted values were found to be in excellent agreement with gas chromatographic measurements. The matrix effects of the samples were found negligible. The method was found appropriate to quantify ethanol in fermented alcoholic beverages. The developed method was simple, fast, precise and accurate. Moreover, no sample preparation was required at all. However, it should be noted that the present procedure is probably not usable for regulatory purposes (e.g. controlling labelling) [24].

The literature survey revealed that FT-MIR and MIR-PLS have been used for the determination of ethanol in the beverages. However, there is no any report on the simultaneous determination of ethanol and methanol by the PLS-NIR method. Thus, the main aim of this work is to develop PLS calibration models for the simultaneous determination of the ethanol and methanol contents of the distilled beverages using NIR, and comparing the result obtained by the reference method, gas chromatography (GC).

EXPERIMENTAL

Preparation of standard solutions

Twenty-four standard samples within the range of 2–15% ethanol and 0.1–1.0% methanol, respectively, were prepared for calibration. The mixture standard solutions were 2, 6, 10 and 15% (w/w) ethanol with 0.1, 0.2, 0.4, 0.6, 0.8 and 1% (w/w) of methanol for their simultaneous

determination. These concentration ranges were selected on the basis that methanol concentration does not exceed more than 1% in real samples. While 1–50% ethanol standard solutions were prepared for the determination of ethanol in the absence of methanol. All the standard solutions were prepared using methanol (99.7%, Sigma-Aldrich, France) and ethanol (99.99%, Fisher scientific, UK) in distilled water. Horizontal type ATR accessory with zinc selenide (ZnSe) crystal was employed for spectral measurements. Spectra were collected in the spectra region between 1720 and 1660 nm with air and water as backgrounds. For the determination of ethanol in the real samples, 10 mL of the samples were diluted to 50 mL in distilled water.

For GC-FID standard solutions ranges from 1–50% (w/w) were prepared from 99.99% (v/v) ethanol in 5% n-propanol. For the determination of ethanol, 10 mL of the samples were diluted to 50 mL in distilled water and n-propanol was used as an internal standard. n-Propanol was added before dilution.

Instrumentation and data analysis procedure

UV-Vis spectrophotometer (Lambda 950, Perkin Elmer) and gas chromatography with flame ionization detector and ECTm-5 capillary column (GC 1000, Dani) were used to determine ethanol and methanol in the samples and standards. Density meter (DMA 4500M, Anton Paar) was used to measure the density of methanol and ethanol standard solutions to convert w/w % to v/v %. Quartz cuvette was used as sample holder for NIR spectrometry. Balance (Adventurer, OHAUS) was used to weight the samples and standard solutions.

The working conditions selected for GC-FID were 3 μ L injection volume, oven temperature initially at 75 °C for 2 min, and then increased to the final temperature of 80 °C in 5 min at rate of 1 °C/min, 210 °C injection port temperature, 0.5 bar pressure, carrier gas flow rate 1.4 mL/min and 300 °C detector temperature. Internal standard is commonly used for the quantification of ethanol using GC-FID. Thus, n-propanol was used as an internal standard since it was not found in the samples. It gave a well separated peak from that of ethanol. The calibration curve was constructed using peak area ratio (ethanol : n-propanol) vs concentration of ethanol, % (w/w). The ethanol concentration of the real samples were determined using the calibration curve, $y = 0.14183x + 0.00743$ having $R^2 = 0.9999$.

The calibration model was obtained by the MATLAB 7.8.0 software. In the ethanol determination, the PLS/NIR calibration made use of 12 standards while the other 12 standards were used for validation.

RESULTS AND DISCUSSION

Spectra of methanol and ethanol

The spectra of mixture of ethanol and methanol showed only one peak (Figure 1). The two components (methanol and ethanol) peaks showed an overlap. Here, a single effect on one component has been observed as an effect of the total. This is because both of them (methanol and ethanol) involve in the absorbance of O–H stretch (Figure 1) [17, 19]. Therefore, NIR requires multivariate technique to resolve these collinear data. Accordingly, the partial least squares method was developed and compared with reference method, gas chromatography.

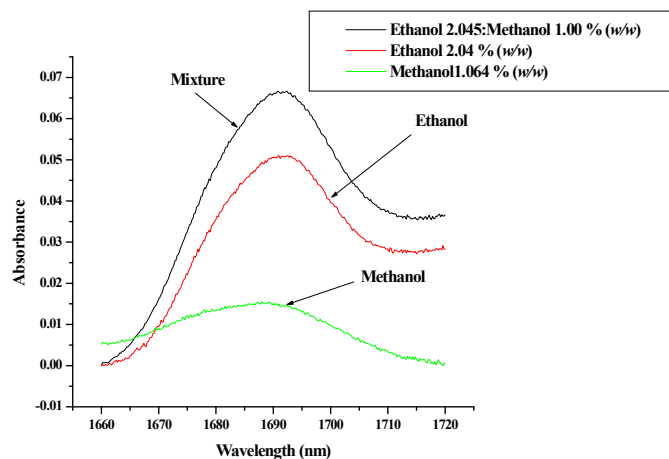


Figure 1. NIR spectra of 2.04% ethanol and 1.0% methanol mixture, 2.04% pure ethanol and 1.064% pure methanol.

Wavelength selection

Wavelength selection is important for any calibration to be accurate. Spectral regions were selected by the spectral distribution of the relative standard deviation in concentration (Sc/c) [21] and shorter spectral range [20]. In this study, the range 1660–1720 nm was selected due to the fact that the solvent (water) does not show any absorbance, the intensity of the source is sufficient; the obtained peaks are comparably sharp in selected region having short spectral range and no study has been reported in the selected region. In addition, the O–H stretching of alcohols may not overlap with aromatic hydrocarbons first overtone above 1650 nm, because the beverages are distilled [17, 19]. Therefore, the partial least squares model was selected to overcome the problem associated with overlap of peaks (Figure 1). The spectral range selected was 1660–1720 nm and the data acquisition interval was 0.2 nm. Accordingly 301 data points were collected.

The partial least squares model

Score plot and partial least squares components

The score plot of the standards in the region is shown in the Figure 2. In the score plot the standards are projected onto new variables, called principal components. Since these projections have been normalized, values in the plot reflect how similar each standard is to a given principal component. In the score plots there are no groups of standards, which show higher similarity in the standards used for calibration (Figure 2). This confirmed that the calibration set has a good distribution.

The contribution of the partial least square components vary depends upon the regions and types of components selected. Similarly, in NIR the first principal component contributed to almost all of the variation. The contribution of the second component is very limited (Figure 3). In this study both components of the mixture; ethanol and methanol have showed the same pattern of partial least components. But, as it is seen in Figure 3, the first principal component

has contributed to almost all of the variation (99.86%) while the other components have negligible contribution. Even though the other components have very little contribution, six components were chosen to construct PLS-NIR model to have the total contribution to 100% which has a great match with the original data.

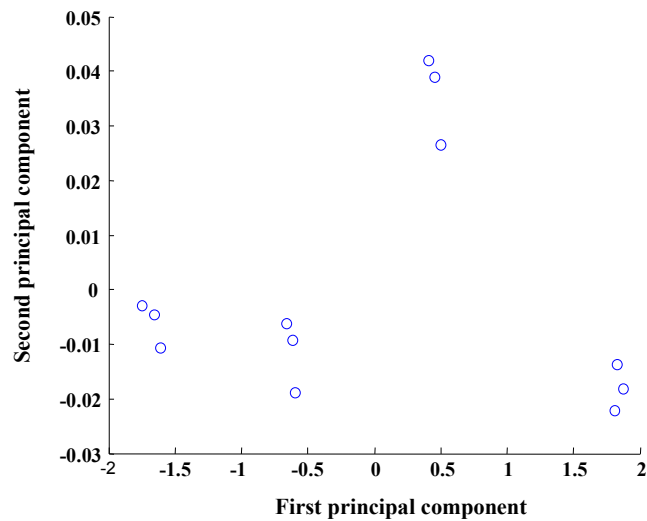


Figure 2. Principal component scatter plot of the NIR calibration set. The mixture standard solutions were 2, 6, 10 and 15% (w/w) ethanol with 0.1, 0.2, 0.4, 0.6, 0.8 and 1% (w/w) of methanol.

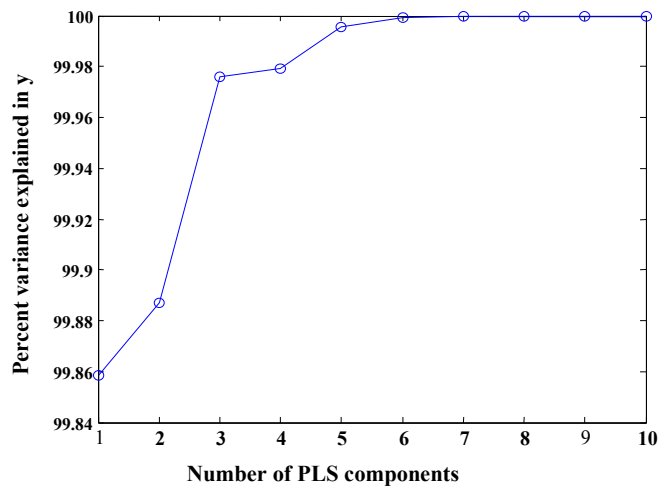


Figure 3. Partial least squares components for ethanol in NIR.

Pretreatment methods

Pretreatment methods are various in numbers. They have been applied for different purposes such as for noise reduction, base line correction, etc [25]. Based on this some data pretreatment methods have been applied, the results obtained are indicated in the Table 1. All of them are best for ethanol in both calibration and validation sets. In spite, they differ by the data they have for methanol. Among the methods, mean centering has more R squared value, $R^2 = 0.929$ and minimum prediction error, RMSEP = 0.08. Therefore, this one has taken to determine the concentration of ethanol and methanol in the real samples.

Table 1. Results of PLS-NIR calibration models for the determination of ethanol and methanol in distilled alcoholic beverages.

Data treatment	Component	Validation method (PC)	Calibration		Validation	
			R ²	RMSEE	R ²	RMSEP
Subtracting minimum	Ethanol	6	1.0000	0.0011	0.999	0.1
	Methanol	6	0.9986	0.0001	0.926	0.07
Mean centering	Ethanol	6	1.0000	0.0002	0.999	0.06
	Methanol	6	0.9997	0.0000	0.929	0.08
Derivative	Ethanol	6	1.0000	0.0000	0.999	0.17
	Methanol	6	0.9999	0.0000	0.067	0.28

PC: number of principal components; RMSEE: root mean square error of estimation, RMSEP: root mean square error of prediction.

Method validation

Validation of the developed model was done using a validation set that contains 12 synthetic samples. The amounts of the measured and predicted ethanol using the models are indicated in Table 2 and the R² and the RMSEP are shown in Table 1. The prediction errors of the model for ethanol and methanol contents are 0.06 and 0.08% (w/w). In addition, the predicted amounts were evaluated and compared with the measured values at 99% confidence level. The results obtained indicated that no significance difference between them. And also the average standard deviation for ethanol and methanol were 0.08 and 0.09% (w/w), respectively. Each sample was analyzed in triplicate and standard deviation was calculated. The average standard deviation was obtained as the mean of standard deviations of the 12 samples. Differences between the measured and predicted values (the error) were in the range (0.00–0.14%, w/w).

Table 2. Amount of ethanol and methanol in the validation set samples.

Sample	Amount measured (% w/w)		Amount obtained (predicted) using the model (% w/w)		Difference between the measured and predicted values (% w/w)	
	Ethanol	Methanol	Ethanol	Methanol	Ethanol	Methanol
1	2.02	0.14	1.99	0.11	0.03	0.03
2	2.04	0.42	2.04	0.42	0.00	0.00
3	2.06	0.85	2.12	0.73	0.06	0.12
4	6.04	0.15	6.06	0.25	0.02	0.10
5	6.05	0.43	6.01	0.43	0.04	0.00
6	6.03	0.83	5.99	0.97	0.04	0.14
7	10.02	0.12	9.88	0.15	0.14	0.03
8	10.03	0.44	10.15	0.34	0.12	0.10
9	10.02	0.81	10.06	0.77	0.04	0.04
10	15.08	0.14	15.07	0.16	0.01	0.02
11	15.02	0.50	14.95	0.59	0.07	0.09
12	15.09	0.88	15.08	0.81	0.01	0.07

Real sample analysis

Three commercial alcoholic beverages, Stolichnaya Vodka, Red Label and White Horse and three traditional beverages, Cereal, Yekosso and Yegibto Arekes were procured from the market for analysis. Ethanol in real samples was quantified using PLS-NIR calibration model and GC-FID. The conversion factor for converting % (w/w) into % (v/v) of ethanol described by Debebe *et al.* [23] was used to convert % (w/w) into % (v/v) of ethanol in the real samples. The % (v/v) of ethanol was obtained by multiplying the % (w/w) with conversion factor 1.2188 [23]. The results obtained by GC-FID and PLS-NIR were statistically analyzed with one-way ANOVA using Origin Lab-Origin 8 at 99% confidence level. The comparison of results pertaining ethanol concentration between reference analysis, GC-FID and the PLS-NIR model is illustrated in Table 3 and Figure 4.

Table 3. Ethanol content in real distilled alcoholic beverage samples as determined by PLS-NIR and compared to the reference method, GC-FID values and the reported method [23].

No.	Samples	Ethanol in % (w/w) found by this study		Ethanol in % (v/v) found by this study		Ethanol in % (v/v) reported in [23]
		PLS-NIR	GC-FID	PLS-NIR	GC-FID	
1	Yegibto Areke	40.12 ± 0.05	38.89 ± 0.05	48.90 ± 0.06	47.40 ± 0.06	48.7
2	Red label	32.79 ± 0.12	33.55 ± 0.68	39.96 ± 0.15	40.89 ± 0.83	39.1
3	S. Vodka	33.17 ± 0.09	34.33 ± 0.32	40.43 ± 0.11	41.84 ± 0.39	39.4
4	Cereal Areke	40.52 ± 0.03	39.62 ± 1.79	49.39 ± 0.04	48.29 ± 2.18	48.2
5	White Horse	30.51 ± 0.38	31.68 ± 0.74	37.19 ± 0.46	38.61 ± 0.90	--
6	Yekosso Areke	39.88 ± 0.11	39.36 ± 0.42	48.61 ± 0.13	47.97 ± 0.51	48.8

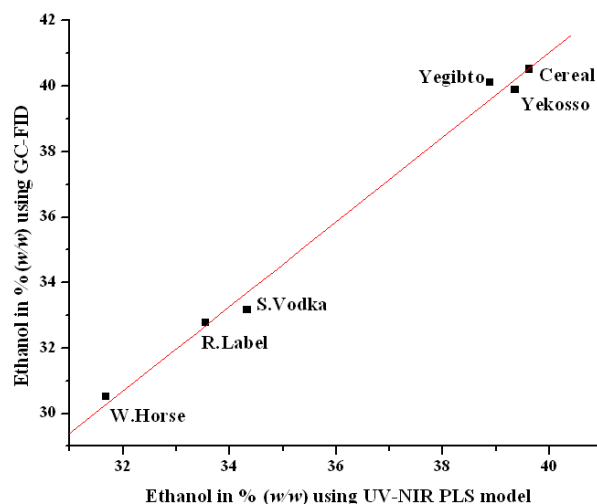


Figure 4. Predicted values versus reference values of ethanol.

Table 3 displays the amount of ethanol determined by the proposed and the reference methods. The obtained result found that the techniques are comparable and did not show any significant difference at 99% ($\alpha = 0.01$) confidence level. Furthermore, in order to check the accuracy of the developed model recovery test was conducted with spiking 2.5 and 5% (w/w)

ethanol in randomly selected three real samples. The percentage recovery obtained ranges from 97–98% (w/w). The amount of ethanol determined by the proposed method was also compared with the reported method [23]. The obtained results showed that the two methods are comparable and did not show any significant difference at 99% ($\alpha = 0.01$) confidence level.

Figure 4 shows the accuracy of the predicted results with the reference ones. Therefore, this confirms the model with few calibration set (12 samples) can be used in NIR for the quantification of ethanol in the distilled alcoholic beverages.

Attempt was also made to determine methanol, but it was not detected in the real samples by both the PLS-NIR and GC-FID methods.

CONCLUSION

In this study, partial least squares model of NIR was found as an alternative technique in the determination of ethanol in alcoholic beverages. The result obtained has a good agreement in case of synthetic and real samples with correlation coefficient (R^2) values of 0.999 and 0.06 RMSEP. Therefore, it can be concluded that the model is the promising alternative in the determination of ethanol in distilled alcoholic beverages.

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