Quantitative Analysis of Alcohol, Sugar, and Tartaric Acid in Alcoholic Beverages Using Attenuated Total Reflectance Spectroscopy

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Mid-infrared (MIR) spectroscopy in attenuated total reflectance (ATR) mode was used for quantifying ethanol, sucrose, and tartaric acid in alcoholic beverages. One hundred synthetic samples were prepared with different ethanol, sucrose, and tartaric acid concentrations. Experiments were carried out on Bio-Rad 175 C FTS using an ATR accessory. Spectra were recorded in the wavelength region 600–4000 cm$^{-1}$. Calibration was performed using partial least squares (PLS) algorithm. Commercially available alcoholic beverages (gin, rum, vodka, etc.) were experimented and concentration of ethanol in these samples was predicted using the developed calibration model. Chemical analysis of these commercial samples was carried out in order to compare the results. The agreement between ATR results with those of chemical analysis revealed good reliability and repeatability of the technique used.

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1. INTRODUCTION

Brewing industry has ever-increasing demand for quality products like wine and distilled spirits. The products should satisfy quality control parameters before commercialization. To achieve this, industry needs an efficient technique to monitor the quality parameters that are effective in controlling the production process. Different methods exist for understanding and characterizing the quality of beverages or to quantify ethanol to meet customer’s demands [1–5]. Enzymatic determination of ethanol, for example, has undergone considerable development and a series of enzymatic methods based on different techniques is available in literature [6–8].

Most of the wet chemical methods are toxic and involve hazardous chemicals in addition to being time consuming. Hence, various laboratory techniques have been refined at times to achieve the goal. Infrared spectroscopy based methods for analysis of wine and determination of ethanol in other alcoholic beverages are recently emerging because of their versatility, efficiency, being cost effective, fast and non-invasiveness [9–14].

A number of researchers tried to exploit mid-infrared transmission spectroscopic studies in brewing industry [15–17]. They observed strong absorption of water in mid-infra-red region that posed problems in spectral analysis. As a result, high prediction errors were reported. The use of attenuated total reflectance technique is shown to be a far better choice for analyzing biological samples. The focus of this article is to develop a calibration method using an ATR technique for quantification of ethanol, sugar, and tartaric acid in alcoholic beverages. The calibration model was built on synthetic samples using partial least squares algorithm.

2. EXPERIMENTAL DETAILS

Absolute ethyl alcohol (99.8 percent), sucrose (99.7 percent), and tartaric acid (99.7 percent) were procured from market for sample preparation. Considering the concentration of constituents in commercially available alcoholic beverages, hundred synthetic samples of different concentrations were prepared for calibration in the lab with the specifications: ethanol (10–50 percent), sucrose (0.8–8.8 percent), and tartaric acid (0.5–8.2 percent). The samples were prepared just prior to collecting the spectra. Six commercially available alcoholic beverages branded Brihans Brandy, Red Riband Vodka, Blue Riband Ginn, Old-Monk Rum, McDowell’s Whisky, and Royal Visa Red wine were procured from market for predicting ethanol concentration.
3. RESULTS AND DISCUSSION

Figure 1 depicts the representative spectrum of the synthetic sample. The C–H stretch absorption bands at 2986 cm\(^{-1}\) and 2901 cm\(^{-1}\) may be attributed to ethanol and tartaric acid present in the sample. An intense sharp band vivid at 1640 cm\(^{-1}\) may be due to C=O stretch because of tartaric acid. The three C–O stretch absorption bands at 1385 cm\(^{-1}\), 1044 cm\(^{-1}\), and 1082 cm\(^{-1}\) may be due to tartaric acid, ethyl alcohol, and sucrose, respectively. All the peaks described above are already defined in the literature [19]. Wavelength selection is important for any calibration to be accurate. Hence, the spectral region 900–1500 cm\(^{-1}\) was chosen for calibration to carry out this analysis since the region houses the major absorption bands of all the three components undertaken for this study.

The overlaid spectra of synthetic and real samples are shown in Figure 2. The position matching of absorption peaks of real samples with synthetic samples in Figure 2 ensures the substitution of synthetic samples for the real ones, to develop the calibration model. All the hundred synthetic samples were used for calibration. All the spectra accounted for calibration were mean centered. 312 data points were used for calibration.

Cross validation, a method that removes one sample at a time from the calibration set and uses it for prediction, was performed. The optimum number of the principal components (PC), which minimized the sum of the residuals, is given below,

\[
\text{PRESS} = \sum_{i=1}^{m} (Y_{MIR} - Y_{REF})^2, \quad (1)
\]

where \(m\) is the number of samples used to constitute the calibration model, \(Y_{REF}\) is the reference value, and \(Y_{MIR}\) is the value provided by the model. Table 1 summarizes the statistical data for all the components analyzed.

The score plot of PC1 versus PC2 for ethanol, sucrose, and tartaric acid is shown in Figures 3(a), 3(b), and 3(c). The samples are projected onto new variables, called principal components. Since these projections have been normalized, values in the plot reflect how similar each sample is to a given principal component. The PLS performed on the absorbance spectra discriminates the excipients well. Figure 3(a) reveals the PC plot of ethanol. Even though all the excipients are distributed well, most of them lie closer and form a group, which shows higher similarity in the samples used for calibration. Out of the first three principal components (PC), contribution of the first PC is the highest (69 percent), 17 percent and 13 percent are the contributions of the second and third PCs, respectively. The contributions of the first 3 PCs for the determination of sucrose concentration are 74 percent, 11 percent, and 13 percent, respectively. Figure 3(c) shows few groups of excipients that are randomly distributed in the score plot of tartaric acid, which denotes the resemblance of samples used for calibration. Variance values for
Table 1: Calibration statistics for ethanol, sucrose, and tartaric acid.

| Analyte       | PLS factors | Wavelength (cm$^{-1}$) | No. of variables | Calibration R$^2$ | RMSEC | Validation R$^2$ | RMSEV |
|---------------|-------------|------------------------|------------------|-----------------|-------|-----------------|-------|
| Ethanol       | 5           | 900–1500               | 312              | 0.9910          | 0.2043| 0.9896          | 0.2193|
| Sucrose       | 5           | 900–1500               | 312              | 0.9962          | 0.1062| 0.9956          | 0.1142|
| Tartaric acid | 4           | 900–1500               | 312              | 0.9898          | 0.1656| 0.9888          | 0.1699|

(a) Score plot of PC1 versus PC2 for ethanol in 900–1500 cm$^{-1}$ region

(b) Score plot of PC1 versus PC2 for sucrose in 900–1500 cm$^{-1}$ region

(c) Score plot of PC1 versus PC2 for tartaric acid in 900–1500 cm$^{-1}$ region

Figure 3

The concentration residual is a parameter that gives an idea about the range of variation that the predicted values of concentration exhibit for the samples involved in calibration procedure. Figure 4(a) represents the concentration residual of ethanol for factor 5. It may be observed that the residual ranges from +0.7 to −0.4, where most of the samples lie in the region ± 0.4. Moreover, majority of the residual values lie close to zero. Figures 4(b) and 4(c) show the residual values...
of sucrose and tartaric acid for their optimum factors 5 and 4, respectively. It is vivid that most of the concentration residual values for both the components lie within the range of ± 0.2.

Six commercially available alcoholic beverages branded Brihans Brandy, Red Riband Vodka, Blue Riband Gin, Old Monk Rum, McDowell’s Whisky, and Royal Visa Red wine were procured from market for analysis. The calibration model obtained for ethanol was then used to quantify ethanol in real samples. The comparison of results pertaining ethanol concentration between reference analysis and the ATR technique gave satisfactory results. Figure 5 shows the accuracy of results predicted through ATR. It is vivid from Figure 5 that the explored ATR technique may very well be an effective alternate to chemical analysis.

4. CONCLUSION

In this paper, ATR technique was investigated as an alternate to chemical analysis. A successful calibration model with cross validation was developed with one hundred synthetic samples that yielded good correlation coefficient (R²) values 0.9910, 0.9962, 0.9898 and lower RMSEC values 0.2043, 0.1062, 0.1656 for ethanol, sucrose, and tartaric acid, respectively. Comparison between ATR results to that of chemical analysis of ethanol gave promising agreement in case of real samples. Hence it may be concluded that the ATR technique may be good and effective alternate for chemical analysis for quantifying constituents in alcoholic beverages. ATR method is fast, accurate, and easy to handle for determination of the constituents in alcoholic beverages.
Figure 5: Predicted values versus reference values of ethanol (in percentage).

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