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# **RESEARCH ARTICLE**



# Infrared spectroscopy chemometric model for determination of phenolic content of plant leaf powder

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### Keywords

Infrared spectroscopy Chemometrics Phenolics Plant leaf powder

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# Abstract

**Introduction:** The leaf is the part of the plant often used in traditional medicine because it is rich in one class of secondary metabolite compounds, namely phenolic compounds. **Objectives:** This study aims to establish a chemometric model for determining the phenolic content of plant leaf powders using the infrared spectroscopy (FTIR and NIR) method with a combination of chemometrics. **Methods:** The dried and powdered plant leaves were scanned using FTIR and NIR spectroscopy. Spectra were used to form calibration models. The calibration models were Partial Least Square (PLS), Principal Component Regression (PCR), and Support Vector Regression (SVR). The selected calibration model was validated using LOOCV and external cross-validation. The best calibration model was PCR, with R<sup>2</sup> and RMSEC values of FTIR and NIR of 0.9918885; 0.9752648 and 0.8675906; 1.5150245, respectively. **Results:** The results of the Paired-Sample T-test analysis of actual samples determined by the selected calibration model compared to the comparison method showed no significant difference.

# Introduction

The leaf is the part of the plant often used in traditional medicine. It is abundantly available in nature and is easy to use and retrieve (Jadid *et al.*, 2020). One of the secondary metabolites found in leaves is phenolic compounds. Phenolic compounds have one or more hydroxyl groups attached to an aromatic ring and are formed from the shikimic acid metabolic pathway. Phenolic compounds are secondary metabolites primarily distributed in plants and essential for growth (Bhatla & Lal, 2018). Additionally, phenolic compounds have antioxidant, antiinflammatory, antiproliferative, antimutagenic, antimicrobial, anticarcinogenic, and cardioprotective properties (Akhtar *et al.*, 2019).

# Methods

In this study, the determination of the total phenolic content in leaf plants was done with UV-Vis spectroscopy and Folin-Ciocalteu's reagent. Gallic acid

used as the standard served to compare the total phenolic content (Blainski *et al.*, 2013). This study also used the Fourier Transform InfraRed (FTIR) and Near InfraRed (NIR) spectroscopy coupled with chemometric analysis. Infrared spectroscopy is an effective analytical technique because it is non-destructive, uses simple preparation, does not require chemicals, and analyses quickly (Haas & Mizaikoff, 2016). However, the spectra generated from FTIR and NIR spectrophotometers are very complicated and overlapping.

Chemometrics were used to assist in analysing the spectral data generated from these two instruments. The authors had previously developed a method to determine the total phenol in plant leaf extracts using IR spectroscopy and chemometrics (Wulandari *et al.,* 2020). This method will be applied in this study on plant leaf powder.

# Materials

Twenty-five (25) leaf samples of medicinal plants with

varieties of phenolic content were collected from the residential and plantation areas in Jember city, East Java, Indonesia (Table I). Reagents used were ethanol 96%, gallic acid (Sigma-Aldrich), Folin-Ciocalteu's reagent (Merck), Na<sub>2</sub>CO<sub>3</sub> (Merck), filter paper, and distilled aqua dest. Five commercial samples (capsule

preparations) were purchased from a pharmacy department store in Jember city. The instruments used were Moisture Analyser type PMB 53, NIR spectroscopy (Brimrose Corporation Luminar 3070), FTIR spectroscopy (Bruker Alpha), and UV-Vis spectroscopy (Hitachi U 1800).

Table I: Samples of leaves used and to	otal phenolic content of the samples
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No	Training set sample		mg GAE/g powder ± RSD (%)	
	Code	Name		
1	А	Carica papaya	14.43 ± 3.520	
2	В	Moringa oleifera	28.33 ± 1.547	
3	С	Averrhoa Carambola	29.60 ± 3.056	
4	D	Ocimum basilicum	31.83 ± 1.248	
5	E	Gnetum gnemon	33.31 ± 1.201	
6	G	Pandanus amaryllifolius	34.50 ± 0.785	
7	н	Leucaena leucocephala	34.59 ± 0.994	
8	I	Artocarpus heterophyllous	35.19 ± 0.754	
9	J	Morinda citrifolia	35.57 ± 1.817	
10	L	Diplazium esculentum	37.08 ± 2.410	
11	Ν	Syzgium aqueum	39.06 ± 1.643	
12	0	Annona muricata	39.42 ± 2.630	
13	Q	Psidium guajava	41.56 ± 1.770	
14	R	Sauropus androgynous	45.19 ± 1.741	
15	Т	Anredera cordifilia	45.98 ± 0.833	
16	U	Syzygium polyanthum	47.19 ± 0.873	
17	V	Cosmos caudatus	47.31 ± 0.333	
18	W	Nephelium lappaceum	50.28 ± 1.294	
19	Y	Persea Americana	60.27 ± 0.527	
20	F	Piper betle	34.12 ± 1.752	
21	К	Coffea canephora	36.89 ± 1.598	
22	Μ	Chrysophyllum cainito	38.20 ± 2.418	
23	Р	Pluchea indica	40.47 ± 3.556	
24	S	Pometia pinnata	45.27 ± 1.608	
25	х	Mangifera indica	55.34 ± 1.879	

# Sample preparation

Leaf samples were powdered by a blender and divided into two groups, the training set and the test set. All samples were dried to fulfil the requirement of the moisture content of herbal powder that is below 10%.

# Determination of the total phenolic content by UV-Vis spectroscopy as a comparing method

For the standard solution, 25 mg and 50 mg of gallic acid were weighed and dissolved in ethanol 96%, then diluted to concentrations of 10 ppm, 20 ppm, 40 ppm, 60 ppm, 80 ppm, 100 ppm, 120 ppm, 140 ppm, and 200 ppm. Then, 100.0  $\mu$ l of each standard and sample solution was pipetted and added with 500.00  $\mu$ l of Folin-Ciocalteu's reagent, left for six minutes, added with 400.0  $\mu$ l of Na<sub>2</sub>CO<sub>3</sub> at 7.5 %, then allowed to stand for 80 minutes. The mixed solution was measured at a wavelength of 743 nm.

# Determination of FTIR and NIR Spectra Data

All samples and gallic acid standards were scanned five replications by placing difference powder in the sample compartment of NIR and FTIR spectrometers. Each scan had five shots (Rahmawati *et al.*, 2015; Wulandari *et al.*, 2016). The NIR and FTIR spectra data were obtained through Acquire Brimrose (NIR) and OPUS (FTIR) software and each spectra data were code-named.

# Determination of model calibration and validation of the model

Spectral data from FTIR at 4000–650 cm<sup>-1</sup> and NIR at 850–2000 nm were analysed by chemometrics using the Unscrambler X software version 10.2. Partial Least Square (PLS), Principal Component Regression (PCR), and Support Vector Regression (SVR) analysed FTIR and NIR spectral data to form a calibration model, and then the best model was selected based on  $R^2$  value close to

one and the smallest value of RMSEC (Root Mean Square of Calibration) and RMSECV (Root Mean Square Error of Cross-Validation). The chosen model was validated using leave one out cross-validation (LOOCV) by taking out one of the training set sample data. The remaining data were used to reform the model. For external cross-validation, the test set consisting of six independent samples was analysed by the model, and the accuracy data results were determined. After validation, the model could be applied to determine the phenolic content of actual samples on the market. The result of the actual sample analysis was compared with the reference method (UV-vis spectrophotometry) with paired sample T-test (Kumar *et al.*, 2015; Nicenboim & Vasishth, 2016).

# Results

Each sample was provided with an identity code. The moisture content of all samples was less than 10% to eliminate the interference of water spectra and prevent microbial growth. The water content of the training set and the sample test set ranged from 3-6%. FTIR spectra of sample and gallic acid showed similar profiles with different intensities of reflectance (Figure 1). The results of total phenolic levels from the training set and the sample test set can be seen in Table I. The results obtained by the comparison method are in the range of 14.43-60.27 mg GAE/g.



<sup>3824,144 3645,618 3477,01 3322,57 3163,88 3005,19 2846,5 2692,061 2524,87 2366,18 2207,49 2048,8 1892,944 1715,835 1545,81 1387,12 1228,43 1069,74 912,467 746,6926</sup> cm -1



60 876 902 928 954 980 1010 1044 1078 1112 1146 1180 1212 1246 1280 1314 1348 1382 1414 1448 1482 1516 1550 1584 1616 1650 1684 1718 1752 1786 1818 1852 1886 1920 1954 1986 🖬

Figure 1: FTIR and NIR spectra of samples and gallic acid

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The formation of the calibration model carried out on the training set data was obtained from FTIR and NIR spectra, processed, and developed for a calibration model with chemometrics. Table II shows that the best calibration model of FTIR spectra was PCR, with R<sup>2</sup>, RMSE (Root Mean Square Error), and RMSECV values of 0.992, 0.868, and 0.980, respectively. The best calibration model of NIR spectra was also PCR, with R<sup>2</sup>, RMSE, and RMSECV values of 0.975, 1.515, and 1.577, respectively. The results of the cross-validation showed that the selected model (PCR) was related to infrared spectra and phenolic content based on the R<sup>2</sup> result of LOOCV, higher than 0.91, with RMSE (Root Mean Square Error) and RMSECV having already small values (Lengkey *et al.*, 2013). The result of the external cross-validation that was carried out using independent samples from a test set of known groups (Jung & Hu, 2015) showed valid results with an  $R^2$  value of 0.989 and an RMSE value of 0.748, indicating that the selected model has good reliability. The selected and validated model was applied to determine the phenolic content in five actual samples as described in Table III. Table III shows that the total actual phenolic content measured by the FTIR-NIR chemometric model and the reference method (UV-vis spectroscopy) had no significant difference (p>0.05) (Santoso, 2014).

Spectra	Model	RMSE	R-Square
FTIR	PLS		
	Calibration	1.079	0.987
	Validation	1.192	0.985
	PCR		
	Calibration	0.868	0.992
	Validation	0.980	0.990
	SVR		
	Calibration	1.837	0.967
	Validation	1.837	0.967
Spectra	Model	RMSE	R-Square
NIR	PLS		
	Calibration	1.638	0.971
	Validation	1.702	0.969
	PCR		
	Calibration	1.515	0.975
	Validation	1.577	0.973
	SVR		
	Calibration	2.334	0.943
	Validation	2.438	0.937

### Table II: Calibration results of the chemometric model

# Table III: Results of total phenolic levels in actual samples

Real Sample	mg GAE/ g powder ± RSD (%)		
	FTIR	NIR	UV-Vis
SN 1	25.06 ± 0.028	24.94 ± 0.064	25.15 ± 1.569
SN2	38.06 ± 0.015	39.13 ± 0.360	39.26 ± 2.927
SN3	21.55 ± 0.039	21.77 ± 0.160	21.74 ± 3.049
SN4	40.65 ± 0.020	40.46 ± 0.200	40.49 ± 1.903
SN5	$15.84 \pm 0.018$	$16.02 \pm 0.890$	16.03 ± 1.338

# Discussion

A total of 25 samples were used, divided into 19 training sets and six sample test sets. The standard used was gallic acid because of its three hydroxyl groups (the more the hydroxyl groups, the higher the antioxidant activity) and because it is a simple phenolic derivative (Fernandes & Salgado, 2016). In this study, the Folin-

Ciocalteu's reagent was used to measure phenolic compounds in the test sample by colourimetric oxidation and reduction reaction. This reagent can oxidise phenolic compounds and reduce heteropoly acid to a molybdenum-tungsten complex (Hudz *et al.,* 2019). The addition of 7.5% Na<sub>2</sub>CO<sub>3</sub> made the medium alkaline because phenolic compounds can only react

with the Folin-Ciocalteu's reagent under alkaline conditions. The dark blue colour indicated the higher concentrations of phenolic compounds. More phenolic ions reduce heteropoly acids to detect them later by UV-Vis spectroscopy (Asrin *et al.*, 2018). The calibration model showed good results with R<sup>2</sup> closer to one and a small RMSE value. RMSEC and RMSECV were based on the smallest value. R<sup>2</sup> was close to one, indicating a linear correlation between the response variable and the predictor variable. The smaller the RMSE, the smaller the error of the model in predicting the response (Georgieva *et al.*, 2013).

# Conclusion

The infrared spectroscopy chemometric model can be used to determine the phenolic content of plant leaf powders. This method is simple, accurate, and environmentally friendly.

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