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Prediction of total antioxidant activity of *Prunella* L. species by automatic partial least square regression applied to 2-way liquid chromatographic UV spectral images

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ABSTRACT

Four different data representations were evaluated for the determination of the total antioxidant activities of four different Prunella L. species, which are Prunella vulgaris, Prunella grandiflora, Prunella laciniata, and Prunella orientalis Bornm. Three different antioxidant assays, 2,2'-azino-bis(3-ethylbenzothiazoline-6-sulfonic acid) (ABST), 2,2-diphenyl-1-picrylhydrazyl (DPPH), and Folin-Ciocalteu (FC) reagent measured the total antioxidant activity and phenolic content of the four Prunella L. species that were extracted with 12 different solvent systems. The data set of 48 Prunella L. extracts was collected by high-performance liquid chromatography (HPLC) with ultraviolet diode array detection. The prediction of total antioxidant activity of Prunella L. species by super partial least square (sPLS) regression was obtained using four different representations of the data; the entire two-way chromatographic-spectral images, the average UV spectra, the total absorbance chromatogram, the lambda max (λ_{max}) chromatogram. The coefficients of determination (R^2) for the entire two-way chromatographic-spectral images (the ABST (0.943 ± 0.008), the DPPH (0.91 ± 0.01), and the FC (0.963 ± 0.006)) indicated good accuracy for predicting antioxidant activities. The three different wet chemical assays are known to yield different values so it is advantageous to estimate the three separate values with a single LC measurement. The entire two-way chromatographic-spectral images have been used to the first time for calibration. Acidic hexane, as an extraction solvent, gave the least root mean square error of prediction (RMSEP) for the two-way chromatographic-spectral images, so it would be the best solvent for modeling antioxidant activities.

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1. Introduction

There has long been an interest in studying and evaluating different plant species for their potential health and medicinal benefits. More frequently, plants containing high concentrations of phytochemicals such as flavonoids, phenolic acids, and terpenes have received much attention due to their potential antioxidant properties that have been observed [1]. Antioxidants can be considered as the general term for these compounds which either inhibit the formation of free-radicals or consumes them [2]. Free radicals may initiate chain reactions that cause the degradation of biomolecular pathways, which subsequently result in cancers and heart diseases [3–6]. Because antioxidants are thought by some to scavenge free radicals, there is interest in characterizing the

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http://dx.doi.org/10.1016/j.talanta.2016.09.014 0039-9140/© 2016 Elsevier B.V. All rights reserved. antioxidant capacity of plants and foods. Such species include the berry family of fruits, artichoke heart, and walnuts [6], all of which contain phenolic compounds which have been evaluated and shown as functional forms of chemical protection from diseases, displaying anti-carcinogenic and anti-mutagenic traits, for example [2,7]. However, overdose consumption of antioxidant-related diets may have no benefit for prevention of disease for human health [8], and surprisingly may increase the probability of facing some health-issues for certain people [1]. There is no clearcut answer if antioxidant-related diets make some contributions to promoting human-health [9].

Here, from the family of *Lamiaceae* as medicinal herbs, *Prunella* L. species have been studied for their antioxidant potentials, and their extracts have been measured to construct models to predict antioxidant activity. The *Prunella* L. species used in this study are *P. vulgarus, grandiflora, laciniata,* and *orientalis Bornm*. One of these species, *Prunella vulgarus,* is considered as an inhibitor of the human immunodeficiency virus (HIV) in early and recently published







papers [5,10,11].

Many different standard procedures exist for evaluating antioxidant activity of botanicals [12,13]. Three antioxidant assays were applied to extracts from the *Prunella L*. species. The first one is 2,2-azino-bis-(3-ethylbenzothiazoline-sulfonic acid) (ABST), which creates a radical cation (ABST $^{\bullet+}$) with an oxidizing agent, such as met-myoglobin [14]. The second assay uses 2,2-diphenyl-1-picrylhydrazyl (DPPH), with the effect determined through a combination of its concentration, the solvent used, the pH value of the mixture, and the length of time [15,16]. The third antioxidant assav involves Folin-Ciocalteu (FC) reagent, which is referred to as a total phenolic assay and is a mixture of phosphomolybdate and phosphotungstate, which acts as a colorimetric assay for phenolic and polyphenolic antioxidants in foods and supplements [12]. Both the ABST and the DPPH are referred to as Trolox equivalent antioxidant capacity (TEAC) assays while the FC method is a gallic acid equivalent (GAE) assay, with all three being based on electron transfer [12]. Collectively, these are considered as non-specific analytical methods, and therefore it is not possible for their results to determine an accurate measure of the antioxidant capacity value of the target species as each method utilizes a different procedure to collect its data [17]. So it is difficult to compare the results from different research groups who are not using precisely the same method [12]. A study by Bolling *et al.* tried to report total antioxidant activity by a single dilution factor to introduce consistent results for different antioxidant assays of different fruit juices, but failed due to the inconsistent measures of total antioxidant activity values, and concluded that a standardized total antioxidant activity assay was immediately needed for an operative comparison [18]. However, our approach seeks to generate reliable measures for each assay from a single chromatographic run.

Because these three assays lack analytical specificity, they are generally not used with the aim of identifying relationships between specific antioxidant-related compounds and health-outcomes [17]. Consequently, if the study of causal relationships is a requirement, there is a need to use a specific analytical method to ascertain any ensuing health benefits associated with specific antioxidant-related compounds. In this case, the identification, isolation, and evaluation of individual antioxidant-related compounds in target food and plant samples may be accomplished by any one of several analytical methods, such as liquid chromatography and mass spectrometry (LC-MS) and high performance liquid chromatography (HPLC). HPLC with ultraviolet diode array detection (HPLC-UV-DAD), for example, was used for assaying the 48 Prunella L. extracts. The specific and comparable findings from such procedures may be crucial for reporting any health outcomes associated with antioxidant intake, and in developing future treatments or preventative steps for diseases [17].

Multivariate calibration is an alternative approach which can be used to predict the total antioxidant activity directly from spectra or chromatographic measurements, and it saves considerable time compared to the aforementioned procedures involving the ABST, the DPPH, and the FC wet-chemical assays. Partial least squares (PLS) regression is one of the most popular multivariate calibration methods [19–21], and this works by modeling the relationships between an independent block of measurements **X** and a dependent block of properties **Y** [21]. This chemometric technique is based on modeling *n* independent variables of **X** (predictor variables) and *l* dependent variables of **Y** (responses), and comprises a set of components that characterize the covariance between the blocks **X** and **Y** [19–21].

Pioneering work has demonstrated that using two-way chromatographic spectral images retains the greatest amount of information and is the best representation for classification [22–27]. This paper demonstrates that the two-way chromatographic spectral images may be used for calibration as opposed to other standard methods such as the total absorbance chromatogram. In the study, four different representations of the data were evaluated with respect to their efficacy for predicting the antioxidant activity: the two-way chromatographic-spectral images, the average UV spectrum, the total absorbance chromatogram and the lambda max (λ_{max}) chromatogram. This paper is the first time that the two-way chromatographic-spectral images have been used to construct a calibration model although other work has been published on peak clusters and selected regions of the chromatogram. For authentication of complex materials, it makes sense to use the entire chromatographic regions because in chemical profiling, the identity of the active compounds and their peaks may be unknown or variable.

2. Theory

2.1. Data preprocessing

The data were not normalized because normalization worsened the prediction rates. However, because the volume that was injected was carefully controlled by the LC sample loop normalization of the chromatograms was unnecessary. Furthermore, because the goal of the measurement is to determine the absolute antioxidant capacity from different extracting solvents, normalization would confound the variations that are to be measured.

Retention time alignment was performed to correct drift in the chromatographic peaks. Peak drift was ascertained by comparing the UV spectra among peaks in the same retention time window. After alignment, the square root transform of the absorbances in the two-way chromatographic-spectral images improved the prediction rates of the models by inflating the relative importance of the smaller features in the chromatographic data and was used consistently throughout the study.

The two-way chromatographic-spectral images were aligned using a two-step polynomial interpolation procedure. First, the total absorbance chromatograms for each solvent used for extraction were very different from each other. The species difference was a minor variation. The two-way chromatographic-spectral images for each of the 12 solvents were aligned by maximizing the correlation with the first occurring chromatogram image in the solvent group using a 5th order polynomial for interpolation.

The second step, used two peaks that were common to all the chromatograms located at 5.4 and 25.2 min to align among the 12 solvent groups by using a simple linear interpolation. For this step, the chromatogram images were aligned with respect to the first object in the data set. This approach worked quite well as can be seen in Fig. 1 and Fig. 2.

2.2. Partial least squares regression

Partial least squares (PLS) regression is a widely used chemometric technique that is usually applied to multivariate data [28,29]. PLS is based on determining the inner relationship or covariance between the column-spaces of independent variables **X** and dependent variables **Y**. The independent matrix comprises one of the four representations of the LC data and the dependent matrix comprises the measured antioxidant activities. All four representations comprised 48 samples. The four species of *Prunella* were extracted with 12 different solvents to furnish the 48 samples.

The PLS procedure was modified so that it would automatically determine the optimum number of latent variables or components [30]. The PLS model internally applied bootstrap Latin partitions [31] with 4 partitions and 20 bootstraps to only the calibration set

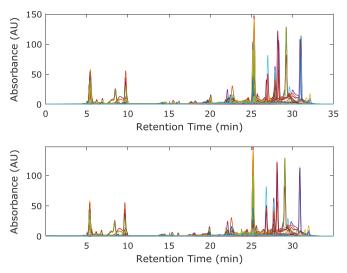


Fig. 1. The total absorbance chromatograms before (top) and after (bottom) retention time alignment.

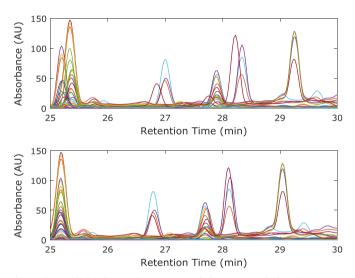


Fig. 2. The total absorbance chromatograms before (top) and after (bottom) retention time alignment zoomed in to 25–30 min.

of data. From the internal bootstrap of the calibration set, the number of latent variables that yielded the lowest average prediction was then used to build a model from the entire calibration set. This algorithm provides a parameter free version of PLS that is affectionately named *Superpls* (sPLS).

2.3. The bootstrapped Latin-partition

Separate from the internal bootstrap of sPLS, an external bootstrap with Latin partitions was also applied for validating the PLS model and comparing the four different data representations. The bootstrapped Latin partitions (BLPs), given in detail elsewhere [31], is a method for evaluation of calibration based on cross-validation that seeks to characterize the variances introduced to chemometric models by training and prediction set selection. This method is advantageous compared to other bootstrapping methods because the distribution of levels (i.e., activities) are similar between the model building and prediction sets, and each object is used only once for prediction so the results may be pooled, so that every object in the data set is used once for prediction for each bootstrap. The pooled results are then averaged across the bootstraps and provide precision measures in the form of 95%

confidence intervals. Because the results from this outer bootstrap procedure are never used to refine the method or estimate parameters, this outer bootstrap applies several hundred external validations (i.e., 4 partitions and 100 bootstraps). The external bootstrap procedure provides a general measure of the prediction error with bounds so that any repeat of the experiment will yield average errors that fall within the bounds.

3. Materials and methods

3.1. Plant materials and preparation of the extracts

Four different *Prunella* L. species (*Prunella vulgaris* L. in Keles, Bursa, *Prunella laciniata* L. in Cali-Inegazi, Bursa, *Prunella grandiflora* L. in Edremit, Balikesir, *Prunella orientalis Bornm.* in Kemer, Antalya) were collected from locations in Turkey in June-July of 2009. The detailed sample preparation is given in references [32,33].

3.2. Reagents

The reagents purchased for the ABST were from Fluka (Buchs, Switzerland), for the DPPH, the FC, and Trolox $[(\pm)-6-hydroxy-2,5,7,8-tetramethylchroman-2-carboxylic acid] from Sigma-Aldrich (St. Louis, MO, USA), and for analytical grade of hydrochloric acid, HPLC grade of methanol, butanol, ethyl acetate, acetonitrile, hexane, formic acid, ascorbic acid, and caffeic acid from Merck (Darmstadt, Germany).$

3.3. Data collections

High-performance liquid chromatography (HPLC) with UV detection was applied to measure 48 *Prunella* extracts [32,33]. The average values of two replicates of the antioxidant measurements were reported in the previous work and are used to construct the calibration models.

4. Experimental section

Calibration models to predict the total antioxidant activity of *Prunella* L. species were built using an automated form of PLS, super PLS (sPLS). For each set of calibration data, an internal bootstrap Latin partition (BLP) is used [31]. For this study, 20 internal bootstraps were used with 4 partitions that are similar to the external bootstrap procedure is described below.

The measured total antioxidant activity of *Prunella* L. species comprise matrix **Y**, and the representations of the chromatographic and UV data comprise matrix **X**. Bootstrapping is a resampling technique performed to generate a generalized measurement of prediction accuracy with confidence intervals. The data-set was divided into 4 equal groups because there were 48 samples collected from 4 different *Prunella* L. species. The partition ratio of the data-set was a 3:1, so that 75% of the data was used for calibration and 25% of the data is used for prediction. The average coefficients of determination R^2 were calculated from the 100 bootstraps.

The preprocessing of the data only used the square root transformation. Average root mean square errors of prediction (RMSEP) were calculated from the 100 bootstraps as well. Confidence intervals were calculated from the standard deviations of the RMSEP of each bootstrap.

The two-way chromatographic-spectral representations were unfolded into vectors so that they would form a data matrix for which each row corresponded to the sample and each column corresponded to an absorbance measurement made at a specific wavelength and retention time. This data matrix had 48 rows and 139,072 columns. The lambda max (λ_{max}) chromatogram is simply the maximum absorbance for each UV spectrum plotted with respect to retention time.

All evaluations were made using MATLAB R2016a (The Math-Works Inc., Natick, MA). The computer operated under Microsoft Windows 7 Enterprise Edition 64-bit Enterprise Edition (Red-mond, WA) and was implemented on a computer with an Intel CoreTM i7 940 CPU Extreme Edition operating at 2.93 GHz with 8.00 GB RAM. Total run time for the total antioxidant prediction of *Prunella* L. species was ~3 h of CPU time for all the four representations for the 10 bootstraps.

5. Results and discussion

The total antioxidant activity and the total phenolic content of the Prunella L. species were previously determined with the analysis of 12 different extraction solvents by three different antioxidant methods, the ABST, the DPPH, and the FC methods. The total antioxidant activity and phenolic content of the Prunella L. species from the ABST, the DPPH, and the FC methods were used from the published articles [32,33]. These antioxidant activities are reported in Table 1. Based on the 4 different Prunella L. species, along with 12 different extraction solvents, the total antioxidant activity by the ABST and DPPH methods were measured as Trolox equivalent (TE). The values ranged from 0.44 + 0.01 to 344 + 2 mg TE/g of dried plants and from 0.05 ± 0.01 to 256 ± 3 mg TE/g dried plants, respectively for the ABST and the DPPH. The total phenolic content was measured as the gallic acid equivalent (GAE) by the FC method, and the values ranged from 1.0 ± 0.2 to 87 ± 1 mg GAE/g of the dried plant material. From Table 1, the most efficient solvent was acidic methanol for extracting the phenolic compounds because it yielded the greatest antioxidant activities ranging from 219 mg TE/g dried sample to 344 mg TE/g dried sample. However, acidic ethyl acetate gave the highest extraction efficiency for the phenolic compounds (Table 1).

Fig. 3 is a surface plot of the chromatographic and UV spectral image of the *Prunella grandiflora* L. extracted with acetonitrile. High absorbance can be clearly seen at the retention time of 5 min where the wavelength ranged between 220 and 275 nm in Fig. 3. Fig. 4 gives the total absorbance chromatograms and the average UV spectra across the retention times of the 4 different *Prunella* L. species prepared with the 12 extraction solvents as reported in Table 1. The UV spectra (bottom) in Fig. 4 comprise 48 different

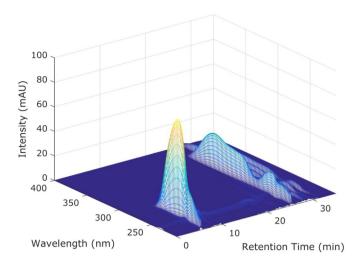


Fig. 3. The two-way representation of the dataset of *Prunella grandiflora* L. species extracted with acetonitrile solvent based on chromatographic and UV spectral measurements after retention time alignment.

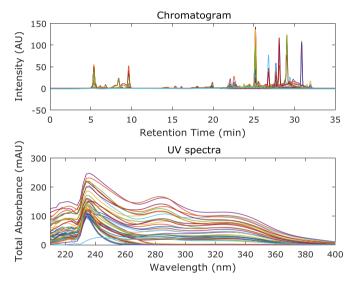


Fig. 4. Chromatogram (top) and UV spectrum (bottom) of 4 different *Prunella* L. species prepared with 12 extraction solvents (total 48 samples).

Prunella L. samples. *Prunella vulgarus* L. when extracted with water gave different peaks than the other species and solvents.

Table 1

Total antioxidant activities of Prunella L. extracts with extraction solvents by antioxidant methods (adapted from Ref. [32,33]).

	Prunella grandiflora L.			Prunella laci	niata L.	Prunella orientalis Bornm.		Prunella vulgarus L.				
	M1	M2	M3	M1	M2	М3	M1	M2	M3	M1	M2	M3
A	18.2 ± 0.5	32.7 ± 1.3	1.8 ± 0.0	22.5 ± 0.6	24.2 ± 0.4	2.0 ± 0.1	14.4 ± 0.2	26.2 ± 1.5	1.7 ± 0.1	22.3 ± 0.7	24.1 ± 1.4	4.2 ± 0.3
AA	$\textbf{37.8} \pm \textbf{0.1}$	73.5 ± 0.4	$\textbf{79.6} \pm \textbf{0.7}$	37.2 ± 0.3	$\textbf{73.8} \pm \textbf{1.1}$	$\textbf{77.3} \pm \textbf{0.8}$	35.4 ± 0.1	$\textbf{72.2} \pm \textbf{0.8}$	63.5 ± 0.6	37.3 ± 0.2	73.2 ± 1.2	$\textbf{76.5} \pm \textbf{0.9}$
AB	81.7 ± 1.8	92.9 ± 1.0	$68.0\ \pm 0.6$	102.5 ± 0.7	94.5 ± 2.0	55.8 ± 0.93	87.2 ± 2.4	86.0 ± 2.0	45.4 ± 1.2	102.1 ± 0.2	$\textbf{88.8} \pm \textbf{1.8}$	47.1 ± 0.7
AE	31.3 ± 0.1	67.5 ± 1.9	83.9 ± 0.4	$\textbf{33.9} \pm \textbf{0.1}$	67.2 ± 0.3	86.1 ± 0.9	$\textbf{33.6} \pm \textbf{0.6}$	67.2 ± 1.2	$\textbf{70.6} \pm \textbf{0.6}$	$\textbf{33.3} \pm \textbf{0.1}$	62.2 ± 0.7	$\textbf{87.3} \pm \textbf{0.9}$
AH	1.4 ± 0.0	0.1 ± 0.0	1.4 ± 0.4	0.7 ± 0.1	0.1 ± 0.1	1.3 ± 0.1	1.7 ± 0.0	0.2 ± 0.0	1.6 ± 0.1	1.5 ± 0.0	0.1 ± 0.0	1.2 ± 0.1
AM	298.2 ± 2.2	237.9 ± 2.1	$\textbf{78.0} \pm \textbf{0.5}$	$\textbf{329.6} \pm \textbf{0.8}$	220.5 ± 2.5	$\textbf{77.0} \pm \textbf{0.5}$	329.0 ± 1.7	256.5 ± 2.8	$63.0\ \pm 0.8$	343.6 ± 2.1	219.3 ± 4.7	$\textbf{76.3} \pm \textbf{1.0}$
AW	201.6 ± 0.7	103.0 ± 2.4	$\textbf{30.9} \pm \textbf{0.4}$	208.9 ± 1.5	107.2 ± 2.2	32.3 ± 1.2	248.4 ± 7.0	108.4 ± 4.1	$\textbf{27.1} \pm \textbf{0.9}$	215.2 ± 0.9	118.8 ± 1.7	26.5 ± 0.6
В	57.1 ± 3.2	76.6 ± 1.7	12.6 ± 0.5	48.3 ± 0.2	63.0 ± 0.4	13.0 ± 0.9	44.7 ± 1.0	65.1 ± 1.5	9.6 ± 0.6	53.3 ± 1.6	53.9 ± 1.5	13.7 ± 0.7
Е	5.7 ± 0.2	10.4 ± 0.2	2.5 ± 0.1	4.1 ± 0.5	5.6 ± 1.0	1.8 ± 0.0	10.7 ± 0.9	11.5 ± 1.1	1.9 ± 0.6	5.6 ± 0.4	7.5 ± 0.4	2.1 ± 0.3
Н	0.5 ± 0.0	0.1 ± 0.0	1.0 ± 0.1	0.4 ± 0.0	0.1 ± 0.0	1.0 ± 0.2	0.5 ± 0.0	0.1 ± 0.0	1.0 ± 0.2	0.8 ± 0.0	0.1 ± 0.0	1.0 ± 0.3
Μ	188.6 ± 3.4	122.5 ± 1.6	$\textbf{16.9} \pm \textbf{0.0}$	151.2 ± 3.5	134.2 ± 5.1	11.7 ± 1.4	201.0 ± 0.4	118.8 ± 1.6	$\textbf{8.4} \pm \textbf{0.6}$	166.1 ± 4.2	151.8 ± 1.6	11.5 ± 0.5
W	123.4 ± 0.9	$\textbf{97.2} \pm \textbf{1.9}$	24.6 ± 0.5	121.9 ± 1.0	84.0 ± 0.8	20.2 ± 1.3	131.8 ± 1.9	85.9 ± 1.2	11.0 ± 0.1	117.7 ± 0.6	92.3 ± 1.5	15.5 ± 0.0

Abbreviations: Methods: M1 (mg TE/g); The ABST Method, M2 (mg TE/g); The DPPH Method, M3 (mg GAE/g); The FC Method. Solvents: A; Acetonitrile, AA; Acidic Acetonitrile, AB; Acidic Butanol, AE; Acidic Ethyl Acetate, AH; Acidic Hexane, AM; Acidic Methanol, AW; Acidic Water, B; Butanol, E; Ethyl Acetate, H; Hexane, M; Methanol, W; Water.

une aignitette							
	ABST	ABST			FC		
	Before	After	Before	After	Before	After	
R1 R2 R3 R4	$\begin{array}{c} 0.936 \pm 0.003 \\ 0.48 \pm 0.04 \\ 0.873 \pm 0.005 \\ 0.935 \pm 0.002 \end{array}$	$\begin{array}{c} 0.943 \pm 0.008 \\ 0.74 \pm 0.08 \\ 0.92 \pm 0.01 \\ 0.93 \pm 0.02 \end{array}$	$\begin{array}{c} 0.911 \pm 0.003 \\ 0.60 \pm 0.03 \\ 0.866 \pm 0.006 \\ 0.903 \pm 0.004 \end{array}$	$\begin{array}{c} 0.91 \pm 0.01 \\ 0.76 \pm 0.06 \\ 0.89 \pm 0.02 \\ 0.89 \pm 0.02 \end{array}$	$\begin{array}{c} 0.969 \pm 0.002 \\ 0.919 \pm 0.003 \\ 0.957 \pm 0.002 \\ 0.885 \pm 0.004 \end{array}$	$\begin{array}{c} 0.963 \pm 0.006 \\ 0.944 \pm 0.006 \\ 0.954 \pm 0.006 \\ 0.90 \pm 0.01 \end{array}$	

Average coefficients of determination R^2 with confidence intervals for each assay from four different representations evaluated by PLS predictions before and after retention time alignment.

Abbreviations: R1; The entire two-way chromatographic-spectral images, R2; Averaged UV spectrum, R3; Total absorbance chromatogram, and R4; Lambda max (λ_{max}) chromatogram.

The PLS prediction of antioxidant activity of the set of 4 *Pru-nella* L. from four different representations; the two-way chromatographic-spectral images, the averaged UV spectrum, the total absorbance chromatogram and the lambda max (λ_{max}) chromatogram. The coefficients of determination (R^2) of the predicted and measured (experimental results in Table 1) antioxidant activities for the four different representations are reported in Table 2. The results are both before and after retention time alignment. Thus, the same trends are observed among the coefficients of correlation R^2 for the four different representations.

Table 2

The prediction results given in Fig. 5 were obtained from the two-way chromatographic-spectral images representations after retention time alignment. As clearly seen, the R^2 values vary among the three different assays. The best correlation result were obtained for the FC method that gave an average R^2 of 0.969 ± 0.002 before retention time alignment, and 0.963 ± 0.006 after retention time alignment. The ABST and the DPPH methods before retention time alignment gave an average R^2 of 0.936 ± 0.003 and 0.911 ± 0.003 , respectively. Retention time alignment had a slightly same effect on prediction of antioxidant activity of *Prunella* L. species with the ABST and the DPPH methods giving an average R^2 of 0.943 ± 0.008 and 0.91 ± 0.01 , respectively, for the two-way chromatographic-spectral images representation.

Often, the average or summed spectrum of a chromatogram is used because retention time drift and consequent alignment step can be mitigated by use of this representation. Fig. 6 is the

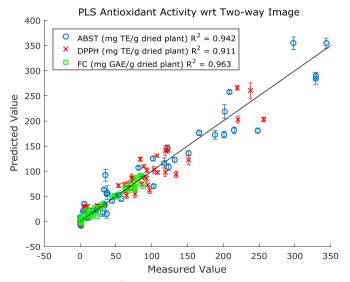


Fig. 5. Correlation values (R^2) of ABST (blue circle), DPPH (red multiplication), and FC (green square) antioxidant methods observed based on the two-way chromatographic-spectral images. Measured values of total antioxidant activity of Prunella L species are from experimental results. Predicted values of total antioxidant activity of *Prunella* L. species were obtained from 10:4 bootstrapped Latin partitions. The error bars are 95% confidence intervals. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

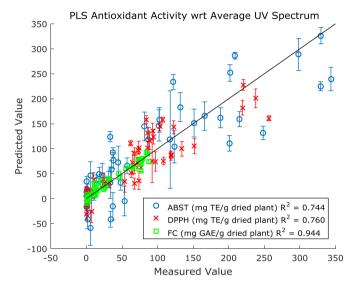
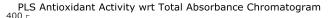


Fig. 6. Correlation values (R^2) of the ABST (blue circle), the DPPH (red multiplication), and the FC (green square) antioxidant methods observed based on UV spectra. Measured values of total antioxidant activity of *Prunella* L. species are from experimental results. Predicted values of total antioxidant activity of *Prunella* L. species were obtained from 10:4 bootstrapped Latin partitions. The error bars are 95% confidence intervals. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

predicted and observed antioxidant and phenolic content values of Prunella L. species for the UV spectral representation after retention time alignment. The decrease in predictability indicates that the UV spectral representation is insufficient to model the antioxidant content of the plant material. The average R^2 based on the UV spectra for the ABST and the DPPH methods before retention time alignment are 0.48 ± 0.04 and 0.60 ± 0.03 , respectively. The significant improvement in correlation values by 0.26 for ABST and by 0.16 for DPPH are observed after retention time alignment, but still quite low as compared to other correlation values of other representations. However, the UV spectral representation may be sufficient to model the phenolic content. The correlation values of the FC method were 0.919 + 0.003 before retention time alignment and 0.944 + 0.006 after retention time alignment providing an improvement for predicting the total phenolic content of Prunella L.

The total absorbance chromatogram is another representation that was evaluated and the prediction results are given in Fig. 7 after retention time alignment. For the ABST and the DPPH methods the average R^2 values are 0.873 ± 0.005 and 0.866 ± 0.006 , respectively, before retention time alignment while the average R^2 of the ABST method after retention time alignment shows an increase with the value of 0.92 ± 0.01 . The DPPH method also showed an increase in the average R^2 of 0.89 ± 0.02 after retention time alignment where it was the value of 0.866 ± 0.006 before retention time alignment. The FC method, on the other



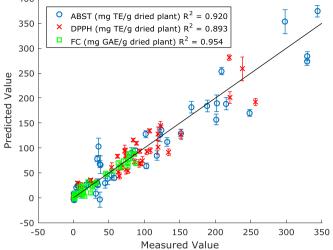


Fig. 7. Correlation values (R^2) of the ABST (blue circle), the DPPH (red multiplication), and the FC (green square) antioxidant methods observed based only on the total absorbance chromatogram. Measured values of total antioxidant activity of *Prunella L.* species are from experimental results. Predicted values of total anti-oxidant activity of *Prunella L.* species are 95% confidence intervals. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

hand, gave an average R^2 of 0.957 ± 0.002 before retention time alignment, but showed a very slightly decrease with an average R^2 of 0.954 ± 0.006 after retention time alignment, but it still falls in to the confidence interval range.

The last representation was the lambda max (λ_{max}) chromatogram after retention time alignment presented in Fig. 8. The ABST method gave the better average R^2 values than both the DPPH and the FC methods with both before and after retention time alignment. The average R^2 value of the ABST method is 0.93 ± 0.02 after retention time alignment while it was 0.935 ± 0.002 before retention time alignment. The DPPH method

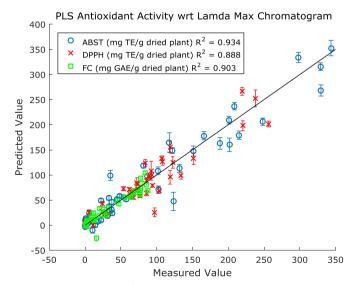


Fig. 8. Correlation values (R^2) of the ABST (blue circle), the DPPH (red multiplication), and the FC (green square) antioxidant methods observed based on PLS calibration of the lambda max (λ_{max}) chromatographic data. Measured values of total antioxidant activity of *Prunella* L. species are from experimental results. Predicted values of total antioxidant activity of *Prunella* L. species are 95% confidence intervals. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

Table 3

Average root mean square error of prediction (RMSEPs) with confidence intervals for each assay from four different representations evaluated by PLS predictions before and after retention time alignment.

RMSEP	ABST (mg Th		DPPH (mg TE/g)		FC (mg GAE/g)	
	Before	After	Before	After	Before	After
R1 R2 R3 R4	$\begin{array}{c} 26.8 \pm 0.4 \\ 85 \pm 2 \\ 27.9 \pm 0.5 \\ 26.7 \pm 0.4 \end{array}$	$\begin{array}{c} 24\pm 1 \\ 50\pm 6 \\ 28\pm 2 \\ 25\pm 3 \end{array}$	$\begin{array}{c} 20.6 \pm 0.3 \\ 47 \pm 1 \\ 25.2 \pm 0.5 \\ 21.7 \pm 0.3 \end{array}$	$\begin{array}{c} 19\pm 1 \\ 31\pm 3 \\ 21\pm 1 \\ 21\pm 2 \end{array}$	$\begin{array}{c} 5.9 \pm 0.1 \\ 9 \pm 1 \\ 6.9 \pm 0.1 \\ 10.9 \pm 0.1 \end{array}$	$\begin{array}{c} 5.9 \pm 0.4 \\ 7.3 \pm 0.4 \\ 6.6 \pm 0.4 \\ 10 \pm 1 \end{array}$

Abbreviations: R1; The entire two-way chromatographic-spectral images, R2; Averaged UV spectrum, R3; Total absorbance chromatogram, and R4; Lambda max (λ_{max}) chromatogram.

showed a decreasing trend from 0.903 ± 0.004 to 0.885 ± 0.015 after retention time alignment, but it still falled into the confidence interval range. However, the FC method resulted an increase in the average R^2 value from 0.885 ± 0.004 to 0.90 ± 0.01 after retention time alignment. Interestingly, the FC method resulted in the lowest average R^2 value with the lambda max (λ_{max}) chromatogram representation, where it yielded the best average R^2 value for the other 3 representations in the case of both before and after retention time alignment.

Table 3 reports the RMSEPs for each representation for both before and after retention time alignment. The entire two-way chromatographic-spectral image yielded the lowest RMSEPs all 3 assays among four different representations with and without retention time alignment. After retention time alignment, ABST method shows a lowered RMSEP from 26.8 ± 0.4 to 24 ± 1 mg TE/g for the entire two-way chromatographic-spectral images. The DPPH method also showed an improvement by lowering the RMSEP from 20.6 + 0.3 to 19 + 1 mg TE/g. The FC method, however, resulted in the same confidence interval range for the RMSEP of the same representation. The great improvements can be seen for UV spectra representation of ABST and DPPH methods where RMSEPs are decreased from 85 ± 2 to 50 ± 5 mg TE/g and from 47 ± 1 to 31 ± 3 mg TE/g, respectively. A slight improvement from 9 ± 1 to 7.3 ± 0.4 mg GAE/g is also observed for FC method of UV spectra representation. The total absorbance chromatogram representation resulted a better RMSEP for DPPH method from 25.2 ± 0.5 to 21 ± 1 mg TE/g after retention time alignment while ABST and FC methods showed similar trend with before retention time alignment. Lambda max (λ_{max}) chromatogram representation, on the other hand, gave the RMSEPs in the confidance intervals range of all three methods after retention time alignment. The FC method gave the lowest RMSEPs with respect to all data representations, with the minimum of 5.9 ± 0.1 mg GAE/g before retention time alignment and 5.9 ± 0.4 mg GAE/g after retention time alignment for the entire two-way chromatographic-spectral images. Based on the average prediction results in Table 3 and because the entire two-way chromatographic-spectral image representation contains the largest amount of information, it was selected to evaluate the extraction solvents and their corresponding antioxidant activities.

Tables 4–6 report the RMSEPs with confidence intervals for each solvent and four species obtained after retention time alignment from the two-way representations. Acidic hexane and acetonitrile had the lowest RMSEP for the ABST method in Table 4 for *Prunella orientalis Bornm*. at $2 \pm 1 \text{ mg TE/g}$, *Prunella vulgarus* L. at $3 \pm 2 \text{ mg TE/g}$ and *Prunella grandiflora* L. at $3 \pm 1 \text{ and } 3 \pm 2 \text{ mg}$ TE/g, respectively. The lowest RMSEP of *Prunella laciniata* L. was observed with acetonitrile at $2 \pm 1 \text{ mg TE/g}$ for the ABST method. In Table 5, acidic hexane gave the lowest RMSEPs for *Prunella grandiflora* L. at $2 \pm 1 \text{ mg TE/g}$, *Prunella orientalis Bornm*. at

Table 4
RMSEP for each solvent and the ABST method. RMSEPs are given with 95% con
fidence intervals after retention time alignment.

	Prunella grand- iflora L. (mg TE/g)	Prunella laci- niata L. (mg TE/ g)	Prunella orientalis Bornm. (mg TE/g)	
Α	3 ± 2	2 ± 1	2 ± 1	3 ± 2
AA	22 ± 11	23 ± 8	57 ± 12	17 ± 3
AB	25 ± 3	33 ± 3	4 ± 2	22 ± 4
AE	15 ± 7	2 ± 2	7 ± 4	30 ± 3
AH	3 ± 1	7 ± 2	2 ± 1	3 ± 2
AM	56 ± 13	40 ± 6	45 ± 12	16 ± 5
AW	20 ± 11	48 ± 3	68 ± 4	34 ± 7
В	12 ± 5	3 ± 1	5 ± 4	5 ± 2
Е	29 ± 5	16 ± 2	8 ± 2	17 ± 7
Н	4 ± 1	7 ± 2	2 ± 1	10 ± 3
Μ	17 ± 8	16 ± 5	28 ± 6	11 ± 4
W	15 ± 7	24 ± 7	9 ± 4	12 ± 5

Solvents: A; Acetonitrile, AA; Acidic Acetonitrile, AB; Acidic Butanol, AE; Acidic Ethyl Acetate, AH; Acidic Hexane, AM; Acidic Methanol, AW; Acidic Water, B; Butanol, E; Ethyl Acetate, H; Hexane, M; Methanol, W; Water.

Table 5

RMSEP for each solvent and the DPPH method. RMSEPs are given with 95% confidence intervals after retention time alignment.

	Prunella grand- iflora L. (mg TE/g)	Prunella laci- niata L. (mg TE/ g)	Prunella orientalis Bornm. (mg TE/g)	
A	4 ± 1	8 ± 2	3 ± 1	6±3
AA	11 ± 6	8 ± 3	14 ± 8	14 ± 3
AB	9 ± 4	22 ± 2	3 ± 1	20 ± 3
AE	9 ± 4	3 ± 2	5 ± 2	11 ± 4
AH	2 ± 1	5 ± 1	1 ± 1	2 ± 1
AM	28 ± 10	18 ± 7	54 ± 4	46 ± 4
AW	8 ± 4	24 ± 2	11 ± 3	21 ± 8
В	23 ± 7	5 ± 1	14 ± 6	17 ± 3
Е	19 ± 4	25 ± 2	6 ± 2	15 ± 7
Н	5 ± 2	3 ± 2	2 ± 1	7 ± 3
Μ	20 ± 6	41 ± 6	23 ± 8	29 ± 9
W	35 ± 7	40 ± 4	11 ± 4	8 ± 5

Solvents: A; Acetonitrile, AA; Acidic Acetonitrile, AB; Acidic Butanol, AE; Acidic Ethyl Acetate, AH; Acidic Hexane, AM; Acidic Methanol, AW; Acidic Water, B; Butanol, E; Ethyl Acetate, H; Hexane, M; Methanol, W; Water.

Table 6

RMSEP for each solvent and the FC method. RMSEPs are given with 95% confidence intervals after retention time alignment.

	Prunella grand- iflora L. (mg GAE/ g)	Prunella laci- niata L. (mg GAE/g)	Prunella orientalis Bornm. (mg GAE/ g)	Prunella vul- garus L. (mg GAE/g)
А	3 ± 1	3 ± 1	2 ± 1	2 ± 1
AA	2 ± 1	8 ± 3	2 ± 1	13 ± 1
AB	7 ± 2	2 ± 1	5 ± 1	2 ± 1
AE	7 ± 3	5 ± 1	3 ± 2	18 ± 1
AH	1 ± 1	1 ± 1	1 ± 1	2 ± 1
AM	9 ± 3	8 ± 3	5 ± 2	10 ± 4
AW	3 ± 1	1 ± 1	1 ± 1	3 ± 2
В	3 ± 2	4 ± 1	2 ± 1	2 ± 1
Е	1 ± 1	2 ± 1	0.8 ± 0.3	5 ± 2
Н	4 ± 1	2 ± 1	3 ± 1	4 ± 1
Μ	5 ± 3	4 ± 2	10 ± 1	5 ± 1
W	14 ± 2	8 ± 2	12 ± 3	7 ± 3

Solvents: A; Acetonitrile, AA; Acidic Acetonitrile, AB; Acidic Butanol, AE; Acidic Ethyl Acetate, AH; Acidic Hexane, AM; Acidic Methanol, AW; Acidic Water, B; Butanol, E; Ethyl Acetate, H; Hexane, M; Methanol, W; Water.

 1 ± 1 mg TE/g, and *Prunella vulgarus* L. at 2 ± 1 mg TE/g. In the case of *Prunella laciniata* L., the lowest RMSEP was observed both with acidic ethyl acetate and hexane at 3 ± 2 mg TE/g. In Table 6 for the

FC method, the lowest RMSEPs for different *Prunella* species varied with the different solvents. Ethyl acetate and acidic hexane for *Prunella grandiflora* L. gave the lowest RMSEP at $1 \pm 1 \text{ mg GAE/g}$, and *Prunella orientalis Bornm*. gave the lowest RMSEPs with acidic hexane and acidic water at $1 \pm 1 \text{ mg GAE/g}$, and with ethyl acetate at $0.8 \pm 0.3 \text{ mg GAE/g}$. Acetonitrile, acidic butanol, acidic hexane and butanol for *Prunella vulgarus* L. had the lowest RMSEP at $2 \pm 1 \text{ mg GAE/g}$. For *Prunella laciniata* L., while acidic hexane and acidic water gave both the lowest RMSEPs at $1 \pm 1 \text{ mg GAE/g}$. The RMSEPs favored different solvents with respect to the different species, based on the RMSEP for the four species, the optimal extraction solvent for phenolic activity may depend on the species.

6. Conclusion

Antioxidant activity assays of the ABST, the DPPH, and the FC were used to measure the total antioxidant activity and phenolic content of 4 different Prunella L. species with the use of 12 extraction solvents. The prediction values of the total antioxidant activity of these Prunella L. species were obtained by PLS calibration based on four different representations: the entire two-way chromatographic-UV spectral images, the average UV spectrum, the total absorbance chromatogram and the lambda max (λ_{max}) chromatogram. By comparing R^2 and RMSEPs results of these four different representations, the entire two-way chromatographic-UV spectral image gave the best predictions of the total antioxidant activity and phenolic content. The entire two-way chromatographic-UV spectral images are used for the first time for calibration. This representation preserves the most amount of information from the chromatographic run and multichannel detector.

Among the three assays, the FC method provided more reliable predictions both with the average R^2 values ranging between 0.90 ± 0.01 and 0.963 ± 0.006 after retention time alignment and with RMSEPs values ranging from 5.9 ± 0.4 mg GAE/g to 10 ± 1 mg GAE/g for all four representations after retention time alignment, with the average UV spectral representation performing the worst and representing the greatest loss of informing power. The ABST and the DPPH antioxidant activity assays failed to predict for this averaged UV spectrum representation.

Further evaluation using the prediction error demonstrated that acidic hexane may be the best solvent for all three assays. It can be concluded that acidic hexane combined with the entire two-way chromatographic-spectral representations provides a reasonable method to rapidly measure phenolic content and antioxidant measurements of *Prunella* extracts that may be extended to other botanical extracts.

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