

## Development of a method based on chemometric analysis of Raman spectra for the discrimination of heterofermentative lactobacilli

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In this work, a method based on Raman spectroscopy in combination with Principal Component Analysis (PCA) and Partial Least Square-Discriminant Analysis (PLS-DA) has been developed for the rapid differentiation of heterofermentative related lactobacilli. In a first approach, *Lactobacillus kefir* strains were discriminated from other species of heterofermentative lactobacilli: *Lb. parakefir* and *Lb. brevis*. After this first approach, PCA allowed for a clear differentiation between *Lb. parakefir* and *Lb. brevis*. For the first level of discrimination, PCA was performed on the whole spectra and also on delimited regions, defined taking into consideration the loading values. The best regions allowing a clear differentiation between *Lb. kefir* and non-*Lb. kefir* strains were found to be: the 1700–1500 cm<sup>-1</sup>, 1500–1185 cm<sup>-1</sup> and 1800–400 (whole spectrum) cm<sup>-1</sup> Raman ranges. In order to develop a classification rule, PLS-DA was carried out on the mentioned regions. This method permitted the discrimination and classification of the strains under study in two groups: *Lb. kefir* and non-*Lb. kefir*. The model was further validated using lactobacilli strains from different culture collections or strains isolated from kefir grains previously identified using molecular methods. The second approach based on PCA was also performed on the whole spectra and on delimited regions, being the regions 1700–1500 cm<sup>-1</sup>, 1500–1185 cm<sup>-1</sup> and 1185–1020 cm<sup>-1</sup>, i.e., those allowing the clearest discrimination between *Lb. parakefir* and *Lb. brevis*. The results obtained in this work, allowed a clear discrimination within heterofermentative lactobacilli strains, proteins being the biological structures most determinant for this discrimination.

**Keywords:** *Lactobacillus kefir*, multivariate analysis, PCA, PLS-DA, Raman spectroscopy.

The selection of bacteria for probiotic formulations requires a continuous isolation of new strains, in spite of the large number of commercially available starters, to ensure the elaboration of products with the greatest benefits for health.

Commercially available strains are generally isolated from milk, vegetables, human or animal microbiota and natural fermented products. However, there are other sources of probiotic bacteria that have not been extensively studied to date. In this sense, kefir grains could be considered as a source of potentially probiotic microorganisms. These grains are composed by a complex microbiota including lactobacilli, lactococci, acetic acid bacteria and yeast, to which several health promoting properties have been attributed

(Saloff Coste, 1996, Garrote et al. 2000, Golowczyc et al. 2007). These beneficial properties of some individual kefir strains and also of kefir as a whole product, together with the great variety of microorganisms present in kefir make this fermented milk an appropriate source of strains for probiotic formulation. However, the complexity of kefir microbiota makes the selection of potentially probiotic microorganisms a tedious task.

When such an amount of isolates is to be analyzed, the use of reliable techniques becomes crucial. For this reason, an appropriate method of analysis for this kind of sample should be non-destructive, non time-consuming and if possible, should not require chemical reagents. Besides that, the use of appropriate techniques of analysis should allow the maximum advantage to be made from the results obtained.

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**Table 1.** Data set preparation and results of performance of PLS-DA

Parameter	Raman data /cm <sup>-1</sup>		
	1800–400	1700–1500	1500–1185
Range	1800–400	1700–1500	1500–1185
i) N <sup>o</sup> Samples calibration set ( <i>Lb. kefir</i> and non- <i>Lb. kefir</i> )	7	7	7
ii) N <sup>o</sup> Spectra collected calibration set ( <i>Lb. kefir</i> and non- <i>Lb. kefir</i> )	45	45	45
iii) N <sup>o</sup> Spectra removed from the calibration	0	5	5
iv) N <sup>o</sup> Samples validation set ( <i>Lb. kefir</i> and non- <i>Lb. kefir</i> )	15	15	15
v) N <sup>o</sup> Spectra collected validation set ( <i>Lb. kefir</i> and non- <i>Lb. kefir</i> )	130	130	130
vi) Mathematical treatment (calibration and validation set)	Raw	Raw	Raw
vii) Pre-processing (calibration and validation set)	BLC <sup>a</sup> , MSC <sup>b</sup> , VN <sup>c</sup>	BLC <sup>a</sup> , MSC <sup>b</sup> , VN <sup>c</sup>	BLC <sup>a</sup> , MSC <sup>b</sup> , VN <sup>c</sup>
viii) PLS-factors for prediction	7	1	6

i) Number of strains *Lb. kefir* and non-*Lb. kefir* selected to calibrate the classification model. ii) Number of spectra collected and used for the calibration of the classification model. iii) Number of spectra identified as “outliers” and removed from the calibration stage. iv) Number of strains *Lb. kefir* and non-*Lb. kefir* selected to validate the calibrated model. v) Number of spectra collected and used for the validation of the calibrated model. vi) Mathematical treatment applied to the data set. In this case no mathematical treatment was applied to the data set, using the raw data. vii) Data pre-processing applied to the spectra prior to the multivariate calibration. viii) Number of components used in the calibrated model for the classification of the validation set. This number of PLS-factors correspond to the first minimum in the RMSEP values versus number of PLS factors or principal components.

<sup>a</sup>BLC: Base line correction

<sup>b</sup>MSC: Multiplicative scatter correction

<sup>c</sup>VN: Vector normalization

In order to achieve these desirable outcomes, physical methods appear to be the most appropriate. In this regard, vibrational spectroscopic techniques, namely Fourier transform infrared (FTIR) and Raman spectroscopies have been widely used for bacteria differentiation (Naumann et al. 1991a & b; Maquelin et al. 2002a). The information included in the vibrational spectra gives a global picture of the whole molecular composition (i.e.: nucleic acids, proteins, lipids and cell wall components) of bacteria. Therefore, the obtained spectroscopic data represent highly specific fingerprints enabling accurate microorganism identification at different taxonomic levels (Naumann et al. 1991a & b; Naumann & Meyers, 2000; Naumann, 2001).

In regard to lactic acid bacteria, FTIR and Raman spectroscopies have been used with different goals. Several articles reported the use of FTIR-based techniques for the differentiation and/or identification of streptococci

(Amiel et al. 2000a, 2001), lactococci (Lefier et al. 2000) *Lactobacillus sakei*, *Lb. plantarum*, *Lb. paracasei* and *Lb. curvatus* (Oust et al. 2004a & b), the *Lb. acidophilus* group (Weinrichter et al. 2001; Luginbühl et al. 2006), the *Lb. casei* group (Amiel et al. 2000b, 2001) and different species of lactobacilli isolated from kefir (Bosch et al. 2006).

Raman spectroscopy has also been demonstrated to be an adequate analytical tool for the characterization of LAB at different taxonomic levels (*Streptococcus thermophilus*, *Lb. acidophilus* and *Lb. delbrueckii* ssp. *bulgaricus* isolated from yogurt) (Rösch et al. 2003; López-Diez & Goodacre, 2004; Zhu et al. 2004; Gaus et al. 2006; Oust et al. 2006). However, to our knowledge no attempt has been reported regarding the employment of Raman spectroscopy for the differentiation of heterofermentative related lactobacilli species.

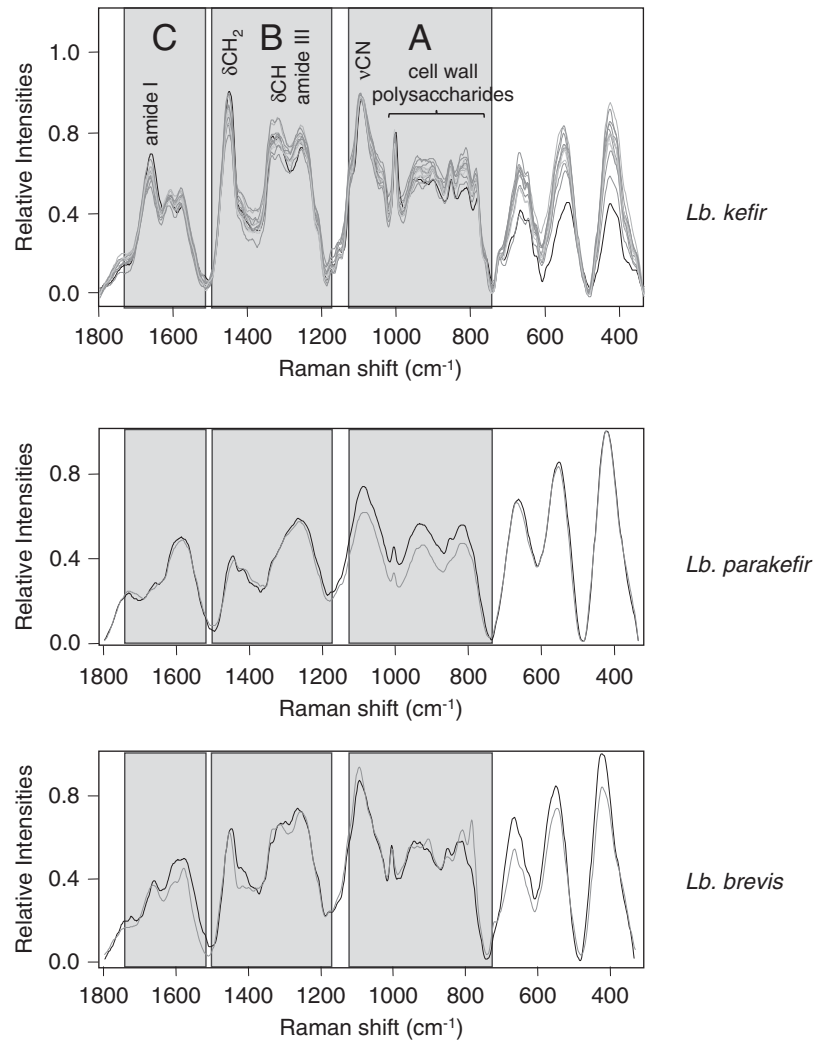
For this reason, the aim of the present work was to combine Raman spectroscopy with multivariate analysis (PCA and PLS-DA) for the discrimination of heterofermentative lactobacilli species generally present in kefir grains, identifying the biological structures that are responsible for this discrimination. In a first level of classification, *Lb. kefir* has been discriminated from *Lb. parakefir* and *Lb. brevis* and in a second level, multivariate analysis allowed for discrimination between *Lb. parakefir* and *Lb. brevis*.

## Materials and Methods

### Bacterial strains and growth conditions

Thirteen wild strains of *Lb. kefir* isolated from different kefir grains at CIDCA (Garrote et al. 2001) (CIDCA 83110; CIDCA 83111; CIDCA 83113; CIDCA 83115; CIDCA 8317; CIDCA 8321; CIDCA 8325; CIDCA 8332; CIDCA 8335; CIDCA 8343; CIDCA 8344; CIDCA 8345; CIDCA 8348), two of *Lb. parakefir* (CIDCA 8322 and CIDCA 8328), three of *Lb. plantarum* (CIDCA 8323, CIDCA 8331, CIDCA 83114) and the commercial strains *Lb. kefir* ATCC 8007, *Lb. brevis* (ATCC 8287 and JCM 1059) and *Lb. plantarum* DSMZ 20174 were cultured in MRS broth (de Man et al. 1960) (Biokar Diagnostics, Beauvais, France) at 30 °C for 48 h. ATCC strains were purchased from the American Type Culture Collection (Manassas, VA 20108, USA) DSMZ strain, from the Deutsche Sammlung von Mikroorganismen und Zellkulturen (Germany), and JCM strain from the Japanese Collection of Microorganisms (Japan). The wild strains belong to the CIDCA culture collection and were identified by molecular methods (whole cell protein profile, 16S-23S rRNA, ARDRA, RAPD-PCR, FTIR) as previously published (Garrote et al. 2001; Bosch et al. 2006; Delfederico et al. 2006; Golowczyc et al. 2008).

The microorganisms were harvested in the stationary phase, collected by centrifugation (10 000 g at 10 °C for 5 min), washed once with 50 mM-phosphate buffered saline (PBS, pH 7) and then twice with bidistilled water. Then, bacterial cells were lyophilized in a FD4 Heto freeze drier



**Fig. 1.** Raman spectra corresponding to all the strains of *Lb. kefir*, *Lb. parakefir* and *Lb. brevis* under study. Scripts A, B and C denote the regions where the main differences among spectra are observed. These regions correspond to 1700–1500  $\text{cm}^{-1}$ ; 1480–1180  $\text{cm}^{-1}$  and 1150–750  $\text{cm}^{-1}$ , respectively.

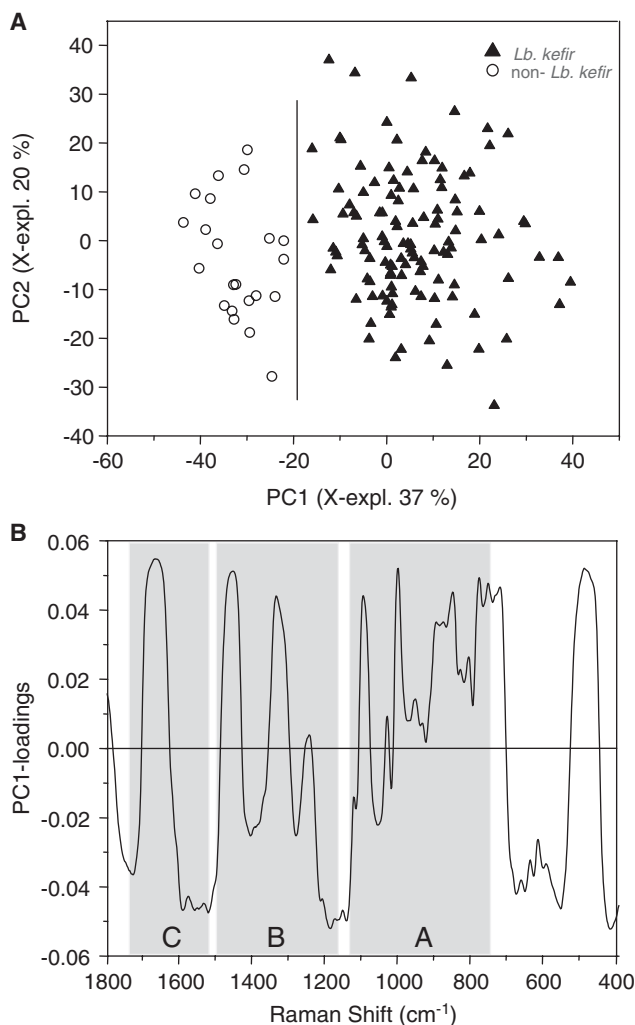
(Lab Equipment, Denmark) and conserved at room temperature.

#### Raman spectra

The Raman Spectra of the bacterial samples were measured by placing them onto an aluminium substrate and then under a Leica microscope (DMLM) integrated to the Raman system (Renishaw 1000B). In order to retain the most important spectral information from each strain, multiple scans were conducted at different points of the sample by moving the substrate on an X-Y stage. The Raman system was calibrated with a silicon semiconductor using the Raman peak at 520  $\text{cm}^{-1}$ , and further improved using samples of chloroform ( $\text{CHCl}_3$ ) with bands at 261, 364 and 667  $\text{cm}^{-1}$  and cyclohexane ( $\text{C}_6\text{H}_{12}$ ) with bands at 383, 426, 801, 1028, 1157, 1265 and 1347  $\text{cm}^{-1}$ , 1443  $\text{cm}^{-1}$ . The wavelength of

excitation was 830 nm and the laser beam was focused (spot size of approximately 2.0  $\mu\text{m}$ ) on the surface of the sample with a  $\times 50$  objective. The laser power irradiation over the samples was 45 mW. Principal component analysis (PCA) was conducted over 140 Raman spectra collected from 14 *Lb. kefir* and 4 non-*Lb. kefir* strains. On the other side, a total of 45 Raman spectra acquired from 4 *Lb. kefir* and 3 non-*Lb. kefir* strains were used for the calibration of the model (Table 1 i & ii). Another set of 130 spectra acquired from 10 *Lb. kefir* and 5 non-*Lb. kefir* strains that had not been used for the calibration were used to validate the model (see Table 1 iv & v for details). Each spectrum was registered with an exposure of 30 s, two accumulations, and collected in the 1800–400  $\text{cm}^{-1}$  region with 2  $\text{cm}^{-1}$  spectral resolution.

The fluorescence contribution was removed by approximating a polynomial function to the spectra and then subtracting it from the spectra.



**Fig. 2.** **A:** Score plots from the PCA analysis performed on the whole range of the Raman spectra (1800–400  $\text{cm}^{-1}$ ) for the lactobacilli samples under study (PC2 vs PC1). ▲ corresponds to *Lb. kefir* strains, ○, to non-*Lb. kefir* strains. **B:** 1-D Loading plot from the PCA analysis performed on the whole range of the Raman spectra (1800–400  $\text{cm}^{-1}$ ) for the lactobacilli samples under study.

### Data Analysis

The recorded Raman spectra were analyzed using GRAMS software (version 3.04, Thermo Galactic, USA). Multivariate analysis and data pre-processing as baseline correction (BLC), multiplicative scatter correction (MSC) and vector normalization (VN) were performed on the Raman spectra, using The Unscrambler™ software (version 8.0, CAMO, Norway).

Principal component analysis (PCA) was performed over the pre-processed Raman spectra in order to evaluate the spectral differences among *Lactobacillus* heterofermentative species in the PC space. In addition, PLS-discriminant analysis (PLS-DA) was used to develop models allowing discrimination of *Lb. kefir* strains from strains belonging to other related species (Martens & Næs, 1989; Esbensen, 2005).

### Brief description of the Methods

PCA is a multivariate technique that operates in an unsupervised manner (each number of the groups under study is not known *a priori*) and it is used to analyze the inherent structure of the data. PCA reduces the dimensionality of the data set by finding an alternative set of coordinates, the principal components (PCs) (Martens & Næs, 1989; Esbensen, 2005). PCs correspond to a linear combination of the original variables, which are orthogonal to each other and designed in such a way that each one successively accounts for the maximum variability of the data set.

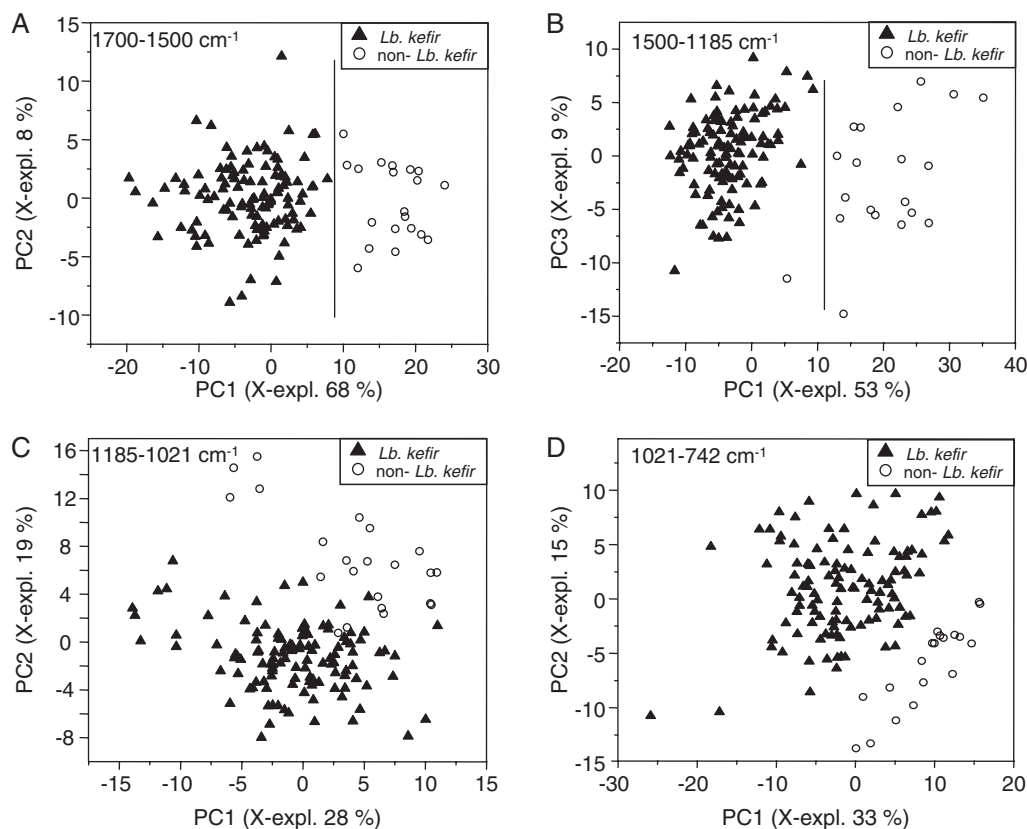
The first principal component (PC1) accounts for as much of the variability in the data as possible, and each succeeding component accounts for as much of the remaining variability as possible. When PC scores are plotted, for example PC1 vs PC2 or any combination of the PCs, they can reveal relationships between samples (grouping). PCA provides insight into which is the percentage of variance explained by each PC and how many PCs should be kept to maintain the maximum information from the original data (Martens & Næs, 1989; Esbensen, 2005). In addition, when the PC loadings are plotted as function of the variables, the plot reveals the most important diagnostic variables or regions related with the differences found in the data set.

PLS-DA is a version of PLS in which one or several Y-variables are modelled simultaneously, thus taking advantage of possible correlations or co-linearity between Y-variables. The discriminant analysis approach assumes that a sample has to be a member of one of the classes included in the analysis. Each class is represented by an indicator variable (a binary variable with a value equal to 1 for members of the class and 0 for non-members). This way, by building a PLS model with indicator variables (Y), it is possible to directly predict the class membership from the X-variables describing any given sample.

The model output corresponds to the predicted value for an unknown sample. A correct prediction should have, ideally, a Y value equal to 1 for the members of the class and 0 for the non-members. All predicted values are accompanied by a deviation, which counts for the reliability of the prediction.

### Results and Discussion

In kefir grains, two large groups of lactobacilli can be found: homo- and heterofermentative. As previously reported for the characterization of kefir lactobacilli using biochemical assays (Garrote et al. 2001) and FTIR spectroscopy (Bosch et al. 2006), discrimination between these groups represents a preliminary step in their characterization. Due to the huge differences between these two groups, Raman spectroscopy also depicted completely different spectra for the strains belonging to each of these large groups. For this reason, we decided to centre the scope of this work on the discrimination within the heterofermentative group, composed



**Fig. 3.** Score plots from the PCA analysis performed on the Raman spectra of the lactobacilli samples under study in the following regions: **A:** 1700–1500  $\text{cm}^{-1}$ , **B:** 1500–1185  $\text{cm}^{-1}$ , **C:** 1185–1021  $\text{cm}^{-1}$  and **D:** 1021–742  $\text{cm}^{-1}$ .  $\blacktriangle$  corresponds to *Lb. kefir* strains,  $\circ$ , to non-*Lb. kefir* strains.

of the phylogenetically related *Lb. kefir*, *Lb. parakefir* and *Lb. brevis* species (Bosch et al. 2006).

The pre-processed average raw Raman spectra of these three species allowed the observation of clear differences between heterofermentative lactobacilli in the whole range analyzed (Fig. 1). The major differences were observed between *Lb. kefir* strains and the other lactobacilli species. The regions where these differences were observed are related to the following bands: the various oligo- and polysaccharides of the cell wall (script A), amide III,  $\text{CH}_2$  and  $\text{CH}$  deformations (script B), and nucleotidic bases ring stretching and amide I (script C) (De Gelder et al. 2007).

### Multivariate Analysis

**Principal Component Analysis (PCA)** In order to take advantage of the huge amount of information provided by the spectroscopic data, a multivariate analysis was carried out on the raw Raman spectra.

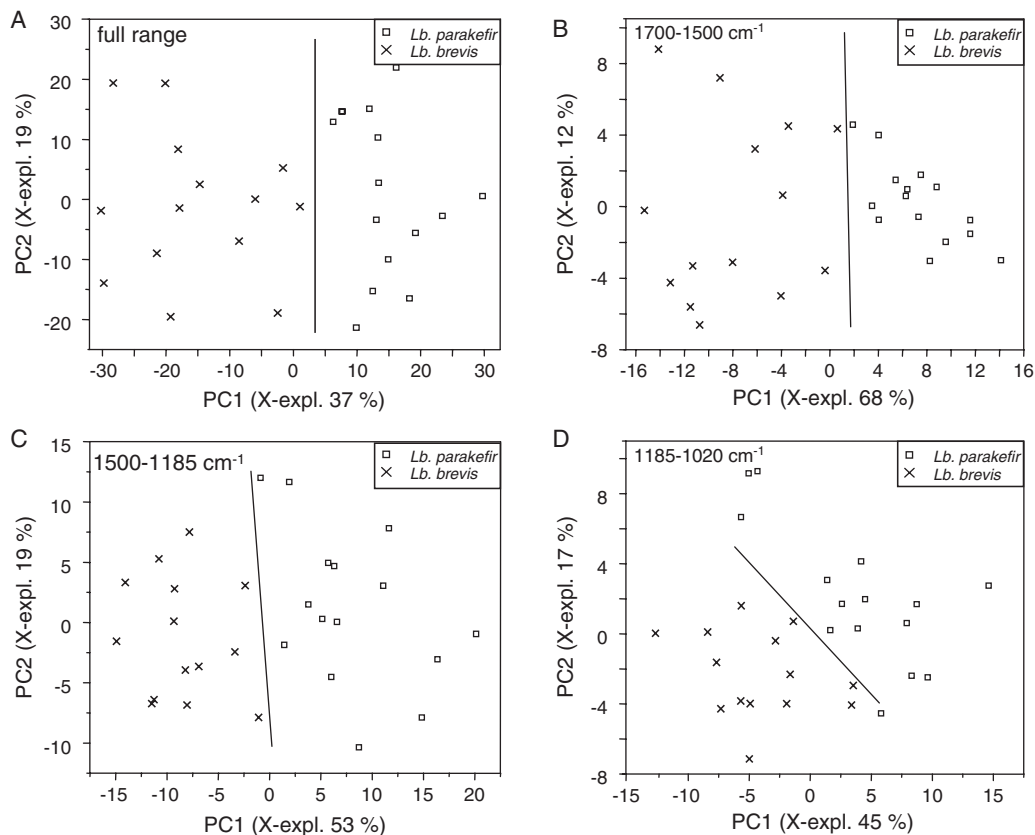
PCA was performed independently on the Raman data sets, with the aim of comparing, in an unsupervised manner, the inherent structure of the spectral data in terms of similarities and differences. For this analysis, 18 different bacterial strains were considered, 14 strains of *Lb. kefir* and 4 strains belonging to the species *Lb. parakefir* and *Lb. brevis*, from now on denominated as non-*Lb. kefir* strains.

In the scores plot obtained from principal component analysis conducted over the raw Raman spectra in the whole spectral range registered (1800–400  $\text{cm}^{-1}$ ) a clear separation between the *Lb. kefir* and non-*Lb. kefir* groups can be observed over the PC1-axis, which explains 37% of the total variance of the data set (Fig. 2A). This separation is the result of the spectral differences observed in the raw data. According to PC1-loadings plot (Fig. 2B), the highest loadings values correspond to the spectral regions where the differences between groups are more evident. These regions are the ones denoted as Scripts A, B and C in Fig. 1.

In order to enhance the differences between the two lactobacilli groups, eliminating irrelevant information in terms of differentiation, the PCA-scores were calculated for those regions where the differences between species were more evident: 1700–1500, 1500–1185, 1185–1021 and 1021–742  $\text{cm}^{-1}$ . The PCA-scores plots corresponding to these regions are depicted in Fig. 3A–D.

From Fig. 3, it can be observed that PCA in the 1700–1500  $\text{cm}^{-1}$  and 1500–1185  $\text{cm}^{-1}$  regions (Scripts A & B) allows for a clear discrimination between *Lb. kefir* and non-*Lb. kefir* strains over PC1-axis. Although certain grade of separation can be observed along PC2-axis in the 1185–1021  $\text{cm}^{-1}$  region (Script C), the overlap of symbols does not allow a clear differentiation between the two groups under consideration. In the 1021–742  $\text{cm}^{-1}$  region (Script D),





**Fig. 4.** Score plots from the PCA analysis performed on the Raman spectra of the *Lb. brevis* and *Lb. parakefir* in the following regions: **A:** 1800–400  $\text{cm}^{-1}$  (full range)  $\text{cm}^{-1}$ , **B:** 1700–1500  $\text{cm}^{-1}$ , **C:** 1500–1185  $\text{cm}^{-1}$  and **D:** 1185–1021  $\text{cm}^{-1}$ . X corresponds to *Lb. brevis* strains,  $\square$ , to *Lb. parakefir* strains.

a separation can also be observed along both PC1 and PC2, which is not reliable to perform the calibration of a classification model. Besides that, this region is known as the “fingerprint region” and therefore, it is strain dependent.

These results allowed us to conclude that the 1700–1500  $\text{cm}^{-1}$  and 1500–1185  $\text{cm}^{-1}$  regions are the best to differentiate *Lb. kefir* from non-*Lb. kefir* strains (Fig. 3A & B). In these regions, the discrimination between both groups could be observed entirely along PC1, which explains 68% and 53% of X-explained variance, respectively. The main spectral differences giving rise to this discrimination in the 1700–1500  $\text{cm}^{-1}$  region correspond to the bands at 1580  $\text{cm}^{-1}$  (arising from the adenine and guanine ring stretching) and at 1660  $\text{cm}^{-1}$  (arising from amide I vibrations) (De Gelder et al. 2007). Within the 1500–1185  $\text{cm}^{-1}$  region, the main bands contributing to discrimination are at 1452, 1320 and 1265  $\text{cm}^{-1}$ , corresponding to  $\text{CH}_2$  and CH deformations, and amide III, respectively (De Gelder et al. 2007).

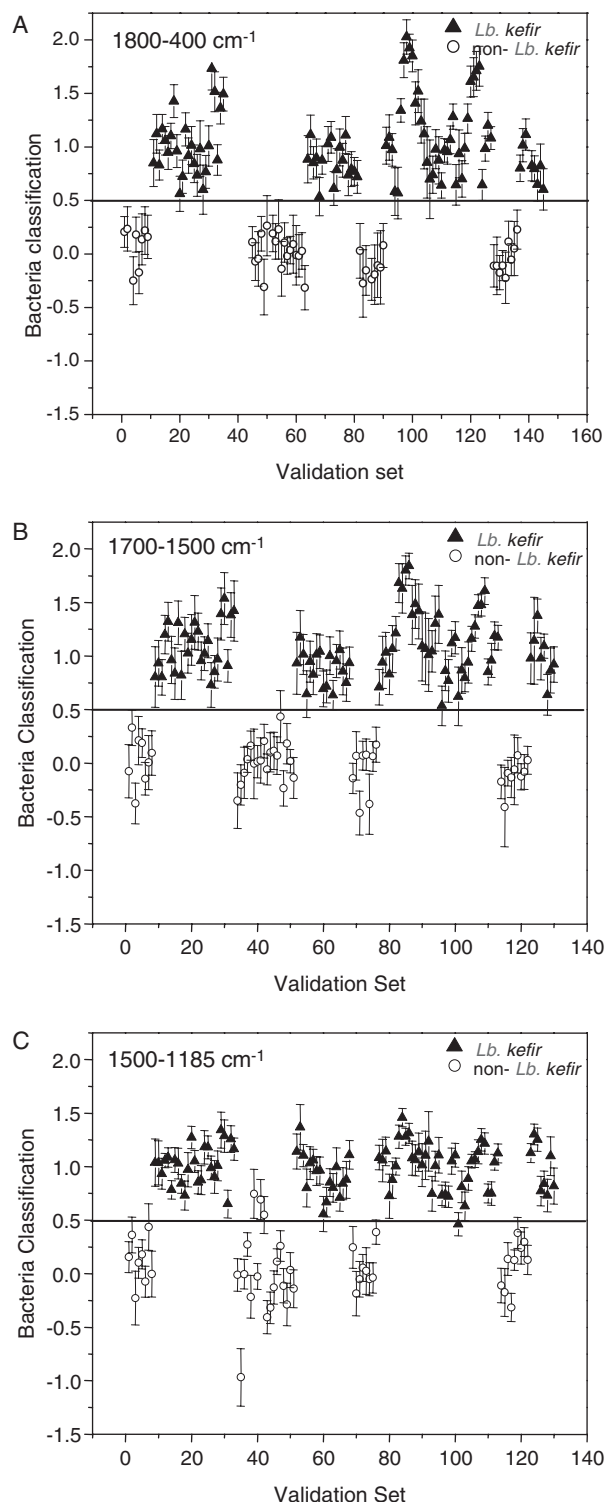
Once it was established that *Lb. kefir* strains can be distinguished from non-*Lb. kefir* strains, an effort was made to discriminate within the latter group, which includes *Lb. parakefir* and *Lb. brevis* strains. Figure 4 depicts the score plots corresponding to the following regions: 1800–400  $\text{cm}^{-1}$  (full range) (Script A); 1700–1500  $\text{cm}^{-1}$  (Script B);

1500–1185  $\text{cm}^{-1}$  (Script C) and 1185–1020  $\text{cm}^{-1}$  (Script D). Discrimination between these two species could be observed along PC1, in all the ranges analyzed. However, discrimination is better when restrained regions are considered. In fact, this PC explains 68, 53 and 45% of variance for the regions considered in Scripts B, C and D, respectively. These results indicate that the main differences between *Lb. brevis* and *Lb. parakefir* can be ascribed to amide I and III and to the  $\text{CH}_2$  and CH deformations (the regions defined in Scripts B to D). This allows us to conclude that the protein profile is one of the main factors contributing to the discrimination between *Lb. brevis* and *Lb. parakefir* species.

In summary, the information depicted in Fig. 3 & 4 allowed us to conclude that proteins are the biological structures the most determinant for the discrimination between heterofermentative species of lactobacilli.

Taking into account the efficiency of the 1800–400  $\text{cm}^{-1}$  (whole range), 1700–1500  $\text{cm}^{-1}$  and 1500–1185  $\text{cm}^{-1}$  regions to discriminate between *Lb. kefir* and non-*Lb. kefir* strains, they were further selected to build three models allowing the classification of unknown samples.

The PCA-scores plots for these three regions also allowed the differentiation between *Lb. parakefir* and *Lb. brevis*. However, the limited number of strains belonging to these



**Fig. 5.** PLS-DA prediction results for Raman spectra obtained from lactobacilli. ▲ corresponds to *Lb. kefir* strains, ○, to non-*Lb. kefir* strains. **A:** whole range (1800–400  $\text{cm}^{-1}$ ), **B:** 1700–1500  $\text{cm}^{-1}$  and **C:** 1500–1185  $\text{cm}^{-1}$ .

two groups makes impossible the calibration of a robust classification model and its further independent validation.

**Table 2.** Statistical values calculated in the validation step for the three classification models calibrated in the regions showing the best discrimination in PCA

Range	1800– 400 $\text{cm}^{-1}$	1700– 1500 $\text{cm}^{-1}$	1500– 1185 $\text{cm}^{-1}$
	Validation	Validation	Validation
Correlation	0.84	0.89	0.87
R-square*	0.55	0.69	0.73
RMSEP*	0.32	0.26	0.24
SEP*	0.32	0.25	0.24

\*R-square: Coefficient of determination; RMSEP: Root Mean Square Error of Prediction; SEP: Standard Error of Performance

**Table 3.** Sensitivity and Specificity values calculated for the classification of bacteria in terms of its Raman spectra using three classification models calibrated in three different regions

Range	1800– 400 $\text{cm}^{-1}$	1700– 1500 $\text{cm}^{-1}$	1500– 1185 $\text{cm}^{-1}$
Parameters			
<i>Lb. kefir</i> spectra classified correctly	77	83	85
<i>Lb. kefir</i> spectra classified erroneously	10	4	2
Non- <i>Lb. kefir</i> spectra classified correctly	41	41	36
Non- <i>Lb. kefir</i> spectra classified erroneously	2	2	7
Sensitivity value (%)	88	95	97
Specificity value (%)	95	95	83

#### Partial least square-discriminant analysis (PLS-DA)

PLS-discriminant analysis was used to develop the classification rules potentially useful for a quick discrimination of unknown samples.

This method operates in a supervised manner, meaning that a prior knowledge of the class membership is required (Martens & Næs, 1989). For the model, two classes were defined: *Lb. kefir* and non-*Lb. kefir*. A correct prediction should have, ideally, a Y value equal to 1 for the samples belonging to the *Lb. kefir* class and 0 for the non-members (non-*Lb. kefir*) (Martens & Næs, 1989). A Y value of 0.5 is the decision value: values higher than 0.5 will indicate *Lb. kefir* strains whereas values lower than 0.5 will denote strains belonging to the non-*Lb. kefir* group.

The PLS-DA was carried out over the pre-processed Raman spectra (including replicates) in an independent set of calibration (Martens & Næs, 1989). The number of samples used and spectra collected for the calibration and validation sets, number of spectra removed from the analysis, ranges analyzed, mathematical treatment, data pre-processing and PLS- factors are presented in Table 1. For the definition of a classification model, seven strains were selected for its calibration (4 *Lb. kefir* and 3 non-*Lb. kefir*) and fifteen for the validation set (10 *Lb. kefir* and 5 non-*Lb. kefir*), being calibration and validation sets independent of each other. Figures 5. A, B & C depict the classification models

performed in the whole spectral range (1800–400  $\text{cm}^{-1}$ ), as well as in the 1700–1500 and 1500–1185  $\text{cm}^{-1}$  ranges.

It is clear from the data that the 1700–1500  $\text{cm}^{-1}$  region is the one where the most accurate prediction can be made (Fig. 5B, Table 2). This result confirms those obtained with the PCA and the visual observation of raw spectra. Note that the plot corresponding to the classification model performed in this region is very similar to that showing the classification using the whole spectra (Fig. 5A). This indicates that the bands allowing discrimination in the 1700–1500  $\text{cm}^{-1}$  range are the most relevant in lactobacilli whole spectra. Besides that, the results also suggest that PC1 in the 1700–1500  $\text{cm}^{-1}$  range is the component containing species specific information. Since the amide I (arising mainly from amide C=O stretching vibration) has the main contribution in this region, these results indicate that the discrimination between *Lb. kefir* and non-*Lb. kefir* strains is somehow influenced by their protein composition and/or structure. Further studies should investigate in which way the composition of proteins determines these species differences.

In order to evaluate the potential for classification of the three regions showing the best discrimination capacity between *Lb. kefir* and non-*Lb. kefir* groups in PCA (1800–400, 1700–1500 and 1500–1185  $\text{cm}^{-1}$ ), the three classification models (one for each of the three regions) were calibrated and validated. The values of correlation, R-square, RMSEP and SEP computed for each model, are displayed in Table 2.

In spite of the occurrence of a few sample mismatches, observed in Fig. 5, a good discrimination between *Lb. kefir* and non-*Lb. kefir* groups was observed, in particular in the 1700–1500  $\text{cm}^{-1}$  region. Table 3 depicts the number of spectra correctly and incorrectly classified as *Lb. kefir* and non-*Lb. kefir*, as well as the calculated sensitivity and specificity percentages. It is important to point out the high specificity and sensitivity achieved for the classification of *Lb. kefir* and non-*Lb. kefir* strains (95% in the 1700–1500  $\text{cm}^{-1}$  region), which denotes the accuracy of the developed model to correctly classify the spectra belonging to *Lb. kefir* strains.

As a final remark, it must be underlined that the supervised method used in this work represents a useful tool for the quick discrimination of unknown samples. The high specificity and sensitivity of the method, in particular in the 1700–1500  $\text{cm}^{-1}$  region, confirm this model as an important tool for the classification of new strains. This way, the registration of a Raman spectrum would be enough to: a) decide if the sample belongs or not to *Lb. kefir* group, and in a second step, b) investigate if it belongs to *Lb. parakefir* or *Lb. brevis* group.

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## References

- Amiel C, Mariey L, Curk-Daubie MC, Pichon P & Travert J 2000a Potentiality of Fourier transform infrared spectroscopy (FTIR) for discrimination and identification of dairy lactic acid bacteria. *Le Lait* **80** 445–459
- Amiel C 2000b Potentialités de la spectroscopie infrarouge à transformée de Fourier (IRTF) pour l'identification et la caractérisation de bactéries d'implication laitière. Ph.D Thesis. University of Caen, pp. 71–95
- Amiel C, Mariey L, Denis C, Pichon P & Travert J 2001 FTIR spectroscopy and taxonomic purpose: Contribution to the classification of lactic acid bacteria. *Le Lait* **81** 249–255
- Bosch A, Golowczyc MA, Abraham AG, Garrote GL, De Antoni GL & Yantorno O 2006 Rapid discrimination of lactobacilli isolated from kefir grains by FT-IR spectroscopy. *International Journal of Food Microbiology* **111** 280–287
- de Man JC, Rogosa M & Sharpe ME 1960 A medium for the cultivation of lactobacilli. *Journal of Applied Bacteriology* **23** 130–135
- De Gelder J, De Gussem K, Vandenabeele P, Vancanneyt M, De Vos P & Moens L 2007 Methods for extracting biochemical information from bacterial Raman spectra: An explorative study on *Cupriavidus metallidurans*. *Analytica Chimica Acta* **603** 167–175
- Delfederico L, Hollmann A, Martínez M, Iglesias NG, De Antoni GL & Semorile L 2006 Molecular identification and typing of lactobacilli isolated from kefir grains. *Journal of Dairy Research* **73** 20–27
- Esbensen KH 2005 In: *Multivariate Data Analysis – In Practice*, CAMO Process AS, Esbjerg, Denmark, 5<sup>th</sup> Edition
- Garrote GL, Abraham AG & De Antoni GL 2000 Inhibitory power of kefir: the role of organic acids. *Journal of Food Protection* **63** 364–369
- Garrote GL, Abraham AG & De Antoni GL 2001 Chemical and microbiological characterisation of kefir grains. *Journal of Dairy Research* **68** 639–652
- Gaus K, Rösch P, Petry R, Peschke KD, Ronneberger O, Burkhardt H, Baumann K & Popp J 2006 Classification of lactic acid bacteria with UV-resonance Raman spectroscopy. *Biopolymers* **82** 286–290
- Golowczyc MA, Mobili P, Garrote GL, Abraham AG & De Antoni GL 2007 Protective action of *Lactobacillus kefir* carrying S-layer protein against *Salmonella enterica* serovar *Enteritidis*. *International Journal of Food Microbiology* **118** 264–273
- Golowczyc MA, Gugliada MJ, Hollmann A, Delfederico L, Garrote GL, Abraham AG, Semorile L & De Antoni GL 2008 Characterization of homofermentative lactobacilli isolated from kefir grains: potential use as probiotic. *Journal of Dairy Research* **75** 211–217
- Lefier D, Lamprell H & Mazerolles G 2000 Evolution of *Lactococcus* strains during ripening in Brie cheese using Fourier transform infrared spectroscopy. *Le Lait* **80** 247–254
- López-Díez EC & Goodacre R 2004 Characterization of microorganisms using UV resonance Raman spectroscopy and chemometrics. *Analytical Chemistry* **76** 585–591
- Luginbühl W, Jimeno J & Zehntner U 2006 Identification of seven species of the *Lactobacillus acidophilus* group by FT-IR spectroscopy. *Lebensmittel-Wissenschaft und Technologie* **39** 152–158
- Martens H & Næs T 1989 In: *Methods for calibration*, in *Multivariate Calibration*, Wiley, Chichester, England. Chapter 3, pp. 97
- Maquelin K, Kirschner C, Choo-Smith LP, Van den Braak N, Endtz H, Naumann D & Puppels GJ 2002a Identification of medically relevant microorganisms by vibrational spectroscopy. *Journal of Microbiological Methods* **51** 255–271
- Maquelin K, Choo-Smith LP, Endtz HP, Bruining HA & Puppels GJ 2002b Rapid Identification of Candida Species by Confocal Raman Microspectroscopy. *Journal of Clinical Microbiology* **40** 594–600
- Mello C, Ribeiro D, Novaes F & Poppi RJ 2005 Rapid differentiation among bacteria that cause gastroenteritis by use of low-resolution Raman



- spectroscopy and PLS discriminant analysis. *Analytical and Bioanalytical Chemistry* **383** 701–706
- Naumann D, Helm D, Labischinski H & Giesbrecht P** 1991a In *Modern Techniques for Rapid Microbiological Analysis*, pp. 43–96 (Ed. WH Nelson). VCH Publishers, New York
- Naumann D, Helm D & Labischinski H** 1991b Microbiological characterizations by FT-IR spectroscopy. *Nature* **351** 81–82
- Naumann D & Meyers RA** (Eds.) 2000 *Infrared spectroscopy in microbiology*. In *Encyclopedia of analytical chemistry*. Wiley, Chichester, UK, pp. 102–131
- Naumann D** 2001 In *Practical Spectroscopy*, pp. 323–378 (Eds. HU Gremlich & B Yan). Dekker, New York, vol. 24
- Oust A, Møretrø T, Kirschner C, Narvhus JA & Kohler A** 2004a. FT-IR spectroscopy for identification of closely related lactobacilli. *Journal of Microbiological Methods* **59** 149–162
- Oust A, Møretrø T, Kirschner C, Narvhus JA & Kohler A** 2004b Evaluation of the robustness of FT-IR spectra of lactobacilli towards changes in the bacterial growth conditions. *FEMS Microbiology Letters* **239** 111–116
- Oust A, Møretrø T, Naterstad K, Sockalingum GD, Adt I, Manfait M & Kohler A** 2006 Fourier transform infrared and Raman spectroscopy for characterization of *Listeria monocytogenes* strains. *Applied and Environmental Microbiology* **72** 228–232
- Rösch P, Schmitt M, Kiefer W & Popp J** 2003 The identification of microorganisms by micro-Raman spectroscopy. *Journal of Molecular Structure* **661–662** 363–369
- Saloff Coste CJ** 1996 Kefir. *Danone World Newsletter*, No: 11
- Weinrichter B, Luginbühl W, Rohm H & Jimeno J** 2001 Differentiation of facultatively heterofermentative lactobacilli from plants, milk, and hard type cheeses by SDS-PAGE, RAPD, FTIR, Energy Source Utilization and Autolysis Type. *Lebensmittel-Wissenschaft und Technologie* **34** 556–566
- Zhu Q, Quivey RG & Berger AJ** 2004 Measurement of bacterial concentration fractions in polymicrobial mixtures by Raman microspectroscopy. *Journal of Biomedical Optics* **9** 1182–1186