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Fast determination of beef quality parameters with time-domain nuclear magnetic resonance spectroscopy and chemometrics

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ABSTRACT

The noteworthy of this study is to predict seven quality parameters for beef samples using time-domain nuclear magnetic resonance (TD-NMR) relaxometry data and multivariate models. Samples from 61 Bonsmara heifers were separated into five groups based on genetic (breeding composition) and feed system (grain and grass feed). Seven sample parameters were analyzed by reference methods; among them, three sensorial parameters, flavor, juiciness and tenderness and four physicochemical parameters, cooking loss, fat and moisture content and instrumental tenderness using Warner Bratzler shear force (WBSF). The raw beef samples of the same animals were analyzed by TD-NMR relaxometry using Carr-Purcell-Meiboom-Gill (CPMG) and Continuous Wave-Free Precession (CWFP) sequences. Regression models computed by partial least squares (PLS) chemometric technique using CPMG and CWFP data and the results of the classical analysis were constructed. The results allowed for the prediction of aforementioned seven properties. The predictive ability of the method was evaluated using the root mean square error (RMSE) for the calibration (RMSEC) and validation (RMSEP) data sets. The reference and predicted values showed no significant differences at a 95% confidence level.

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1. Introduction

For most consumers, a sensory response reflects their first impression about the quality of foods [1,2]. In general, the flavor, juiciness, and tenderness are the main properties used to evaluate the quality of meat on the consumer's plate [3].

The sensory analysis of food is determined by trained assessors who award scores for different attributes. These scores have a specific point scale to describe the tested property [4]. However, the problem with this procedure is the inherent subjectivity in the results [5]. Additionally, these panel evaluations are time-consuming and expensive [6].

Several studies have shown that meat sensory attributes such as juiciness and tenderness depend on the water content and its distribution. Bertram et al. [7] showed the relationship between water mobility and its distribution using transverse relaxation time (T_2) measured by the Carr-Purcell-Meiboom-Gill (CPMG) pulse sequence (T_2 relaxometry) and the sensory attributes in pork slaughtered between the ages of 90 and 180 days. Time-domain nuclear magnetic resonance (TD-NMR) relaxometry

showed three different water environments in the meat linked to (i) proteins, (ii) myofibrils, and (iii) the myofibrillar lattice. Each region has a distinct range of relaxation times based on its aforementioned water distribution [7–9]. The meat with the highest juiciness score was from the 90-day old pork, and it had a longer relaxation time from the extramyofibrillar water corresponding to more mobile water than that observed in the older pigs [7].

Properties such as cooking loss, fat and moisture content, and instrumental tenderness are the main physicochemical parameters commonly used to evaluate beef quality [10,11]. However, the drawback is that the content of fat and/or moisture is estimated by taking an average value for one carcass. In other words, the reported value does not belong to the product purchased by the consumer [12].

The practice of displaying the nutrients or chemical properties on the product's label could be improved by utilizing the results from analytical laboratory tests and other techniques [13]. The specific characteristics for each package should be displayed on the product's label to help the consumer's choice and should not be displayed as an average estimation [12].

Therefore, the goal of this study was to show the possibility of predicting sensory traits, such as beef flavor, juiciness, and tenderness, as well as the physicochemical parameters, such as

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cooking loss, fat and moisture content, and instrumental tenderness using Warner Bratzler shear force, and TD-NMR relaxometry, which depends on fat and water content and their distributions.

The relaxometry studies were performed using a standard CPMG sequence to measure the T_2 values and the sequence known as Continuous Wave-Free Precession (CWFP) to measure longitudinal relaxation time, T_1 and T_2 , in a single and fast experiment [14]. In our study, the entire TD-NMR signals obtained from the sequences were investigated to develop models to predict these seven properties. The main objective was not to replace the human opinion but to speed up the sensory tests and help the manufacturer to improve quality control.

2. Materials and methods

Animal Care and Use Committee approval was not obtained for this study because samples were taken from federally inspected slaughter facilities.

2.1. Procurement of samples

Sixty-one Bonsmara heifers were separated into five groups according to genetic (breeding composition) and feed system (grain and grass feed). After harvest and chilling, a portion of each left side strip loin (*Longissimus dorsi* muscle) was collected, vacuum packaged and sent to the Meat Lab at the State University of Campinas (UNICAMP, Campinas, São Paulo state, Brazil).

After 14 days of aging, for each portion, 4 steaks (2.5 cm thick) were prepared, overwrapped with parchment paper and frozen ($-20\text{ }^\circ\text{C}$) before being used for sensory analysis, shear force and proximate analysis at Meat Laboratory at UNICAMP (Campinas, São Paulo state, Brazil) and, for TD-NMR tests at Embrapa—Instrumentation (São Carlos, São Paulo state, Brazil). For cooking and TD-NMR measurements, each steak was first trimmed of any external fat and connective tissue.

2.2. Sensory analysis

Steaks were thawed ($4\text{ }^\circ\text{C}$ for 24 h) and cooked in a conventional electrical oven—FEC (Imequi, series 8–4000 W, São Paulo, Brazil) equipped with upper and lower electrical resistances. The oven was pre-heated with the thermostat adjusted to $170\text{ }^\circ\text{C}$, and the steaks' internal temperatures were individually monitored.

Each steak was placed on a metal rack over an aluminum tray and was turned over after reaching an internal temperature of $40\text{ }^\circ\text{C}$. After this point, only the upper resistance was left on. Steaks were removed from the oven when they reached an internal temperature of $71\text{ }^\circ\text{C}$.

Grilled steaks were immediately cut into 1 cm cubes, which were placed in glass flasks with metal lids. For sampling, cubes with apparently no internal connective tissue were used.

A yogurt maker with the thermostat adjusted to $40\text{ }^\circ\text{C}$ was used to keep the samples warm until evaluation, which was conducted in individual light- and temperature-controlled booths [15].

A trained sensory panel of eight members evaluated the tenderness, juiciness and flavor intensity of the samples on an 8-point scale (8=extremely tender, juicy and intense and 1=extremely tough, dry and bland) [16]. A total of 444 evaluations were performed, and they were reported by means of individual scores.

2.3. Cooking loss

Before cooking, steaks were trimmed to remove external fat and the thawed weight was recorded. Once steaks exited the oven, the cooked weight was recorded immediately. Cooking loss (%) was calculated using the following formula:

$$\text{Cookingloss(\%)} = \left[\frac{(\text{thawedweight(g)} - \text{cookedweight(g)})}{\text{thawedweight(g)}} \right] \times 100$$

2.4. Proximate analysis

Steaks were trimmed of fat and connective tissue, ground and oven dried to a constant weight (12 h). The moisture content was determined by weight difference [17]. The extraction of intramuscular lipids was performed using the Bligh and Dyer method [18], which is a recommended method for determining total lipid content in biological tissues [19].

2.5. Warner Bratzler shear force (WBSF)

After cooking, following the same procedure as performed for the sensory analysis, the steaks were placed on trays, covered with plastic film, and stored overnight at $4\text{ }^\circ\text{C}$ for WBSF analysis.

In the next day, the cores from the lateral, middle, and medial portions (for a total of 6 cores, 1.3 cm in diameter) of each steak were removed parallel to the longitudinal orientation of the muscle fibers. The cores were sheared using a texture meter TA-XT 2i (Texture Technologies Corp./Stable Micro Systems, Godalming, Surrey, UK) equipped with a 1 mm thick Warner–Bratzler blade, and the peak shear force was recorded, and the average was determined. The peak load (kg) for all 6 cores was averaged, and the mean peak load (kg) was analyzed for each sample [16].

2.6. Time-domain nuclear magnetic resonance measurements

After thawing, each raw sample was separated into three cylindrical slices using a cylinder cutter with a diameter of 2 cm. For measurements, a benchtop SLK 100 TD-NMR spectrometer (Spinlock Magnetic Resonance Solution, Cordoba, Argentina) equipped with a 0.23 T permanent magnet (8.9 MHz for ^1H) and a 13×30 mm probe head was applied to collect CPMG and CWFP decay signals. The CPMG sequence was executed using $\pi/2$ and π pulses of 6.28 and 12.56 μs , respectively, and echo times of $\tau = 300\text{ } \mu\text{s}$ with a total of 1500 echoes. The dead time was approximately 50 μs . The CWFP [14] sequence also used $\pi/2$ and π pulses of 6.28 and 10.6 μs , echo times of $\tau = 141.56\text{ } \mu\text{s}$ and 1501 echoes. The frequency offset was 5 KHz. Each signal for both sequences was the result of an average of 4 scans. The room temperature was held constant at $23\text{ }^\circ\text{C}$.

2.7. Data set evaluations

For the construction of PLS multivariate regression models, two separate data matrices were organized into 61 lines, which correspond to the samples, and 1500 and 1501 columns for variables (time) of CPMG and, of CWFP, respectively. TD-NMR signals are related to the independent variables (**X** matrix). The flavor, juiciness, tenderness, cooking loss, fat content (total fat), moisture content and instrumental tenderness (Warner Bratzler shear force) were the dependent variables (**Y** matrix) used to compute the regression models. In the calibration data set, 49 samples were included, and the validation data (randomly chosen) set was comprised of 12 samples. The PLS chemometric technique is available on the Pirouette 4.0 rev. 2 software (Infometrix, Bothell, Washington, USA). An electronic spreadsheet

available at Microsoft Excel™ software was applied for statistical analysis.

3. Results and discussion

The sensory tests were executed following the international standard methods, and the results were considered to be adequate for evaluating the beef samples' quality. The intra-sample variability shown by the RSD values in Table 1 is considered to be accurate because 12 or more slices from the same steak were used. In addition, there is intrinsic subjectivity from each individual assessor's perception of the sample's sensory property.

The physicochemical parameters were determined using at least three replicates. The instrumental tenderness (kg F^{-1}) was performed with six replicates and the cooking loss (g g^{-1}) was measured for each steak. These results are shown in Table 2.

The reference value of each parameter for sensorial and physicochemical properties used to calculate the regression models was the mean from the other analyzed steaks. These tests require different instrumentation and are time-consuming. To overcome this limitation, TD-NMR can help to construct calibration models based on the chemical features of food. The main objective is to quickly visualize the repeatability of the quality parameters.

The advantage shown in this study is the possibility of measuring raw beef samples with benchtop TD-NMR spectrometry. The sample needs no preparation, and the method is non-destructive. The TD-NMR signals from the benchtop spectrometer are shown in Fig. 1 where differences between the relaxation decay signals for the CPMG (1500 variables) and CWFP (1501 variables) sequences can also be visualized. TD-NMR has sufficient sensitivity to measure the differences in water content in several meat samples due to variations in a signal's amplitude [8,20].

The TD-NMR data were evaluated using the PLS chemometric technique to create multivariate models [21,22]. The advantages of PLS regression when compared with fitting procedures are: (i) it is not necessary to assign the T_2 values (or another parameter) to the water in a specific environment; (ii) it is not biased by the operator and (iii) it can be easily automated.

The raw data were mean-centered for modeling the independent variables. The number of latent variables (LV) needed to perform the prediction is showed in Table 3 for both sequences (CPMG and CWFP).

With one or two LV (case of cooking loss and moisture content), CPMG signals required the first 56 variables of the TD-NMR signals for 99% of the explained variances for mostly of the seven investigated parameters. For the CWFP signals, the values of the explained variances ranged up to 40 to 98% and the LV needed was between 1 and 2 (for fat content). In this case, the first 58 variables were used.

The selection of variables for both sequences was based on the highest regression coefficients for the data. The signals of the CWFP sequence were also transformed with the first derivative (15-point window) and mean-centered, according to our previous study [23], thus generating models with best results. This pre-processing also

allowed more information to be extracted from the tested data in this study.

The ranges for sensory and physicochemical parameters are also shown in Table 3. The magnitude of the RMSE values depends on the property under analysis. The sensory parameters are non-dimensional because the assessors followed a numerical point scale (scores).

The noteworthy result from this study was that lower variations are observed for the RMSE values for signals from the CPMG and CWFP sequences. This finding provides good evidence that the TD-NMR signals are sufficiently accurate or even more accurate than an individual assessor's evaluation and the standard tests for the physicochemical parameters supported by the high correlation between the water content and the amplitude of CPMG and of CWFP signals. Then, the ability of TD-NMR to distinguish the differences between samples was verified with a paired Student's *t*-test that showed no significant difference at 95% confidence levels. The same result was observed for the validation data set.

Table 2

Results from the reference methods for cooking loss, proximate analysis and instrumental tenderness (WBSF).

Parameter	Cooking loss (g g^{-1})	Fat content ($\text{g} \times 100 \text{g}^{-1}$)	Moisture Content ($\text{g} \times 100 \text{g}^{-1}$)	Instrumental tenderness (WBSF) (kg F^{-1})
Set of breeding composition	Mean \pm SD			
1 ($n=10$)	22.03 \pm 3.01	7.43 \pm 0.86	70.97 \pm 2.59	3.88 \pm 0.46
2 ($n=16$)	23.18 \pm 2.39	4.45 \pm 0.41	73.24 \pm 2.23	3.71 \pm 0.27
3 ($n=10$)	22.60 \pm 3.14	3.81 \pm 0.18	73.56 \pm 0.81	3.96 \pm 0.89
4 ($n=14$)	23.02 \pm 2.03	3.70 \pm 0.39	73.32 \pm 1.64	4.00 \pm 0.65
5 ($n=11$)	22.03 \pm 3.17	3.08 \pm 0.50	72.55 \pm 3.35	4.11 \pm 0.41

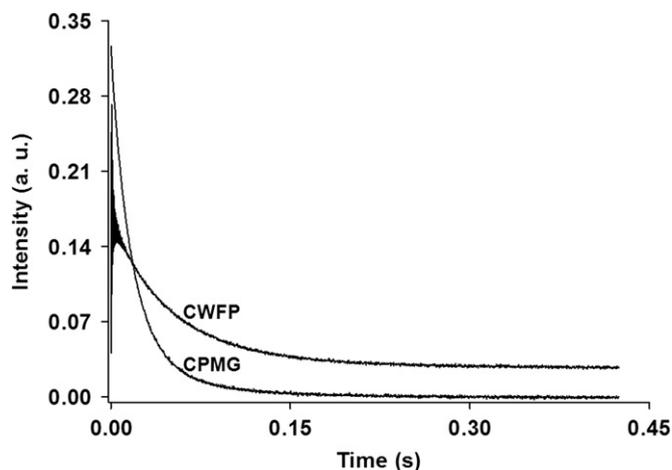


Fig. 1. Profiles of the TD-NMR signals for CPMG and CWFP recorded for one beef sample.

Table 1

Values of relative standard deviation in percentage (%RSD) computed from the trained assessors' evaluations for sensory properties.

Sensory property Set of breeding composition	Flavor Mean \pm SD (Median and %RSD range)	Juiciness	Tenderness
1 ($n=10$)	6 \pm 2 (6 and 6–36%)	6 \pm 1 (6 and 11–40%)	7 \pm 1 (7 and 6–22%)
2 ($n=16$)	6 \pm 1 (6 and 8–35%)	6 \pm 1 (6 and 8–36%)	7 \pm 1 (7 and 7–27%)
3 ($n=10$)	5 \pm 2 (6 and 17–36%)	6 \pm 1 (6 and 12–44%)	6 \pm 1 (6 and 10–24%)
4 ($n=14$)	6 \pm 2 (6 and 10–42%)	6 \pm 1 (6 and 8–35%)	6 \pm 1 (6 and 7–33%)
5 ($n=11$)	5 \pm 1 (6 and 9–45%)	6 \pm 1 (6 and 12–33%)	6 \pm 1 (6 and 8–28%)

Table 3
Results for prediction of sensory and physicochemical parameters of beef samples using PLS multivariate model and TD-NMR signals.

Parameter	Sensory			Physicochemical			
	Flavor	Juiciness	Tenderness	Cooking loss (g g ⁻¹)	Fat content (g × 100 g ⁻¹)	Moisture Content (g × 100 g ⁻¹)	Instrumental tenderness (WBSF) (kg F ⁻¹)
Range	4–7	5–7	4–8	15.06–30.48	1.96–8.91	65.40–75.76	2.70–5.26
CPMG sequence							
% Explained variance (LV)	97.2 (1)	99.1 (1)	99.2 (1)	99.9 (2)	94.3 (1)	99.9 (2)	99.1 (1)
RMSEC (n=49)	0.6	0.7	0.7	2.81	1.61	1.78	0.57
RMSEP (n=12)	0.6	0.5	0.6	2.45	1.16	2.40	0.56
CWFP sequence							
% Explained variance (LV)	39.8 (1)	52.2 (1)	45.4 (1)	41.9 (1)	92.1 (2)	59.5 (1)	79.2 (1)
RMSEC (n=49)	0.6	0.7	0.8	2.79	1.34	1.93	0.58
RMSEP (n=12)	0.6	0.5	0.5	2.11	1.55	2.57	0.55

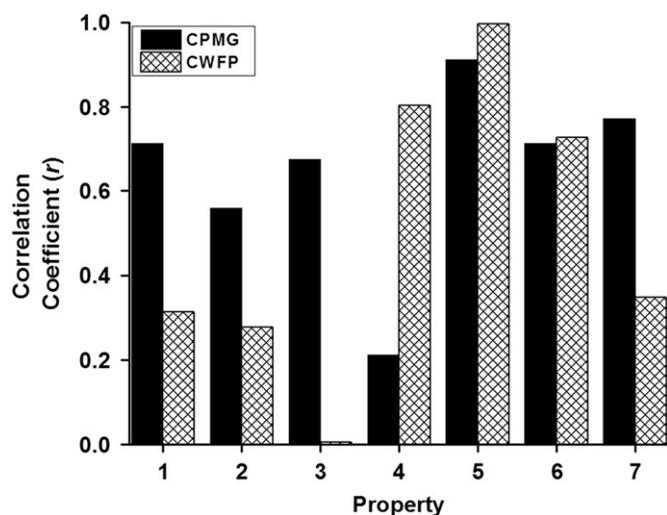


Fig. 2. Values of the correlation coefficients between the classical analysis results and those predicted by PLS multivariate model with TD-NMR signals. From left to right (1) flavor, (2) juiciness, (3) tenderness, (4) cooking loss, (5) fat content, (6) moisture content and, (7) instrumental tenderness (WBSF).

The potential of these multivariate models was also confirmed by the low values for RMSEP, as shown in Table 3. For the calculations, these samples were not included in the calibration data set. Thus, the multivariate models can help predict the studied properties by measuring of fat and water content using TD-NMR.

The correlation coefficients (r) between the values of reference methods and those predicted by the PLS models can be verified in Fig. 2. CPMG is very promising for most properties with r values up -0.56 to -0.91 , except for cooking loss ($r = -0.21$). Otherwise, CWFP signals have a good correlation for cooking loss ($r = 0.80$), moisture ($r = 0.73$) and fat content ($r = 0.99$). For better comparisons, the modulus values of correlation coefficients are shown in Fig. 2.

4. Conclusions

The findings of this study showed the possibility of simultaneously measuring several sensory properties, such as flavor, juiciness and tenderness, as well as cooking loss, fat and moisture content, and instrumental tenderness (WBSF) with the TD-NMR signals from two different sequences, CPMG and CWFP using a PLS multivariate model. The low root mean square errors for the calibration and validation data sets were verified in the models

and support the notion that the models can predict these seven parameters. The information from the models provides a reliable and fast analytical method for commercial applications; the TD-NMR measurement takes less than 1 s per beef sample and it can be easily automated.

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