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Authentication of the Botanical and Geographical Origin of Honey by Front-Face Fluorescence Spectroscopy

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Front-face fluorescence spectroscopy, directly applied on honey samples, was used for the authentication of 11 unifloral and polyfloral honey types (n=371 samples) previously classified using traditional methods such as chemical, pollen, and sensory analysis. Excitation spectra (220–400 nm) were recorded with the emission measured at 420 nm. In addition, emission spectra were recorded between 290 and 500 nm (excitation at 270 nm) as well as between 330 and 550 nm (excitation at 310 nm). A total of four different spectral data sets were considered for data analysis. Chemometric evaluation of the spectra included principal component analysis and linear discriminant analysis; the error rates of the discriminant models were calculated by using Bayes' theorem. They ranged from <0.1% (polyfloral and chestnut honeys) to 9.9% (fir honeydew honey) by using single spectral data sets and from <0.1% (metcalfa honeydew, polyfloral, and chestnut honeys) to 7.5% (lime honey) by combining two data sets. This study indicates that front-face fluorescence spectroscopy is a promising technique for the authentication of the botanical origin of honey and may also be useful for the determination of the geographical origin within the same unifloral honey type.

KEYWORDS: Honey; unifloral; polyfloral; botanical origin; geographical origin; authenticity; adulteration; front-face fluorescence spectroscopy

INTRODUCTION

According to the Codex Alimentarius Standard (I) and the European Union Council Directive (2) relating to honey, the use of a botanical designation of honey is allowed if it originates predominately from the indicated floral source. Honey may also be designated by the name of a geographical region if it was produced within the area referred to (I, 2).

The vast majority of the honeys on the market contain significant nectar or honeydew contributions from several plant species and are therefore called polyfloral or multifloral honeys. Normally, they are just designated with the word "honey". Probably no honey produced by free-flying bees is purely unifloral. The term unifloral honey is used to describe honey in which the major part of the nectar or honeydew is derived from a single plant species. Honey composition, flavor, and color vary considerably depending on the botanical source it originates from (3).

The physical, chemical, and pollen analytical characteristics of the most important European unifloral honeys have been described in various papers (3-7). Contrary to the unifloral honeys, the polyfloral honeys do not exhibit distinct physical or chemical characteristic apart from a huge variability, which makes their authentication particularly difficult.

The interest in the production of unifloral honeys is caused by higher consumer preference for some honey types, generating a commercial concern of the beekeepers. The recent interest in the therapeutic or technological use of certain honey types may also contribute to the demand for a reliable determination of the botanical origin.

Botanical Origin. A number of new analytical techniques combined with multivariate data analysis have been proposed for the determination of the botanical origin of honey. They are, for example, based on physical and chemical measurands determined during quality control of honey (8, 9) or the former combined with the determination of mineral content (10), as well as carbohydrate composition (11), amino acid composition (12), mass spectrometry or metal oxide semiconductor based gas sensors (13, 14), differential scanning calorimetry (15),

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pyrolysis mass spectrometry (16), and Raman (17) and near-infrared spectroscopy (18).

Many of the methods mentioned above allow one to clearly discriminate between several types of unifloral honeys, but none of these methods accounts for the polyfloral honeys that represent the majority of the honeys produced. This means that these methods may not be useful in analytical practice, as the great challenge in honey analytics is not to distinguish between several unifloral honey types but to discriminate the minority of unifloral honeys from the overwhelming majority of polyfloral honeys on the market. This also explains why until now none of these proposed methods are commonly used for the determination of the botanical origin of honey.

Only a single ion chromatographic method has been tested in the presence of polyfloral honeys and showed a potential to discriminate between several unifloral as well as polyfloral honey samples by first classifying the honey samples into two groups by color measurements (19). However, only very few samples were analyzed in this study, and it remains to be verified if this methodology is useful in analytical practice.

Currently, a reliable determination of the botanical and geographical origin can be achieved only by a global interpretation of sensory, pollen, and physicochemical analyses carried out by experts (4, 20, 21). However, the uncertainty related to the interpretation of pollen analytical results, originating from a number of different factors, demands the development of new analytical methods (22).

Geographical Origin. Pollen analysis is currently used to determine the geographical origin of honey as pollen in honey reflects the vegetation type where the nectar has been collected by the bees. In the past many analytical methods such as amino acid composition (23, 24), Raman spectroscopy (17), mineral content (25, 26), and sugar or mineral composition combined with common chemical quality control data (27-29) together with multivariate data evaluation have been proposed for the determination of the geographical origin.

Unfortunately, in most of the above quoted studies the botanical origin of the honey samples was not determined or the discrimination between the geographical origins was not verified on samples of the same botanical origin. Generally, the sample sets analyzed were small or limited to a small geographical area. The distinctions found are therefore rather due to differences of the vegetation type between the geographical regions and thus to the botanical origin of honey (30). A geographical discrimination will therefore be found when the differences are related to the vegetation type present in these areas.

As several analytical methods have to be used together for a reliable authentication of the botanical origin, such work is time-consuming and costly. Very specialized expertise is needed for the interpretation of the pollen spectrum used for the determination of the geographical origin of honey. Thus, there is a real need for new methods that allow a rapid and reproducible authentication of the botanical and geographical origin of honey at low cost (21, 31).

Fluorescence Spectroscopy. Compared to spectroscopic techniques based on absorption, fluorescence spectroscopy offers a 100–1000-fold higher sensitivity. It provides information on the presence of fluorescent molecules and their environment in inorganic and organic materials. In addition, front-face fluorescence spectroscopy allows an investigation of fluorophores in powders as well as in concentrated or opaque samples (32, 33).

Honey is known to contain fluorophores such as polyphenols (34-37) and amino acids (38, 39). Fluorescence spectroscopy should therefore be helpful for authenticating the botanical origin of honey. More detailed information on fluorescence spectroscopic applications to honey and other food can be found in our previous study, which already showed that front-face fluorescence spectroscopy is a promising approach for the determination of the botanical origin of honey (40).

The aim of the current work was to study the fluorescence spectroscopic characteristics of 11 honey types and to develop a rapid, low-cost, and reliable method for the authentication of unifloral and polyfloral honeys. As the physical and chemical characteristics of honey may be changed by adulteration, the potential of fluorescence spectroscopy was also studied on this subject. As minor nectar contributions from plant species other than the unifloral source may contribute to regional characteristics of unifloral honeys, the potential of fluorescence spectroscopy for the determination of the geographical origin of honey was studied as well.

MATERIALS AND METHODS

Sampling and Botanical Classification by Reference Methods. A total of 371 honey samples produced between 1998 and 2004 were collected and stored at 4 °C until analysis. They originated predominately from Switzerland (CH), but samples from Germany (D), Italy (I), Spain (E), France (F), Slovenia (SLO), and Denmark (DK) were

also included.

To classify these honey samples corresponding to their botanical origin, the following measurands were determined according to the harmonized methods of the European Honey Commission (41): electrical conductivity, sugar composition, fructose/glucose ratio, pH value, free acidity, and proline content. Pollen analysis was carried out according to DIN 10760 (42, 43).

On the basis of these analytical results, the honey samples were assigned to one of the following 11 honey types, according to the criteria of Persano and Piro (3): acacia (Robinia pseudoacacia) (CH, n=14; D, n=4; F, n=3); alpine rose (Rhododendron spp.) (CH, n=14; I, n=5); sweet chestnut (Castanea sativa) (CH, n=21; I, n=5; F, n=3); rape (Brassica napus var. oleifera) (CH, n=22); fir honeydew (Abies and Picea spp.) (CH, n=56; D, n=63; SLO, n=2); oak honeydew (Quercus spp.) (E, n=8); honeydew from Metcalfa pruinosa (I, n=14); heather (Calluna vulgaris) (D, n=21; DK, n=2); lime (Tilia spp.) (CH, n=14; D, n=9; I, n=4); dandelion (Taraxacum s.l.) (CH, n=10; D, n=7; I, n=2); and polyfloral honeys (CH, n=68). In the heterogeneous group of the polyfloral honeys, nectar or honeydew contributions from all of the above-mentioned sources were represented.

Adulterated Honeys. To evaluate the potential of fluorescence spectroscopy to detect beet sugar adulteration, an artificial honey was produced by feeding two colonies after the nectar flow, in autumn, with a sucrose solution of 62.5 g/100 g, generally used as winter feed for bee colonies in Switzerland. The sucrose solution was converted into artificial honey by the bees and left to ripen in the combs until extraction. To evaluate the possibility to detect honey adulteration by fluorescence spectroscopy, six chestnut and six acacia honey samples were adulterated with 50% of the artificial honey produced.

Fluorescence Spectroscopy. An aliquot of 20 g of the honey samples was liquefied at 55 °C for 8 h, allowed to cool to room temperature, and poured into a 1 cm quartz cuvette. The latter was placed into the sample holder of a Perkin-Elmer LS 50 B luminescence spectrometer (Perkin-Elmer, Beaconsfield, U.K.) equipped with a variable-angle front-surface accessory, with the incident angle of the excitation radiation set to 56°. Spectra were recorded at a scan rate of 150 nm/ min and saved as ASCII textfiles. Instrumental artifacts were corrected in excitation using a rhodamine cell in the reference channel.

Method Development. To find additional wavelength ranges with specific emission or excitation for the honey types of interest in addition

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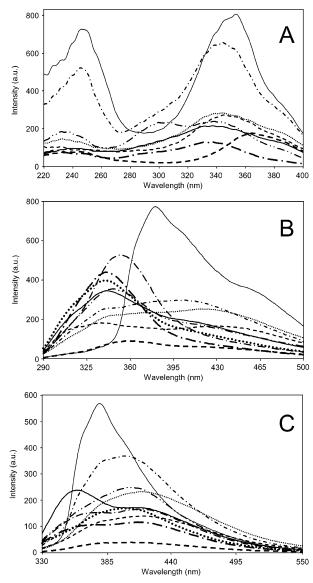


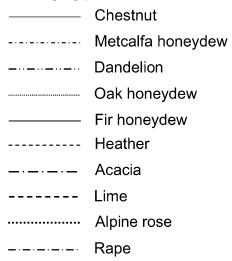
Figure 1. Fluorescence spectra of different honey types.

to those already used in the preliminary study (40), the following ranges were studied. An excitation scan between 220 and 440 nm and recording of the fluorescence intensity at 420 and 490 nm was carried out. Six further emission scans were recorded with wavelengths between 220 and 600 nm, with excitation wavelengths being 210, 270, 310, 350, 390, and 440 nm, respectively. The following three instrumental settings yielded the most discriminating fluorescent spectra for the 10 types of unifloral honeys studied: excitation scan between 220 and 440 nm with the fluorescence emission measured at 420 nm (method A); using the excitation wavelengths of 270 and 310 nm, fluorescence emission spectra were recorded from 290 to 500 nm (method B) and from 330 to 550 nm (method C), respectively The excitation slit width was set to 10 nm and the scan speed to 150 nm/min for all three methods. Two spectra were recorded using different aliquots of each sample. The spectra of the honey types studied are shown in **Figure 1**.

A control honey sample for the evaluation of instrumental stability and determination of the intermediate precision of the method was prepared by heating an acacia honey for 20 min up to $100~^{\circ}$ C; the sample was then filtered to remove the pollen grains, partitioned into 2 mL glass vials, and then stored at $-20~^{\circ}$ C until analysis. The intermediate precision was determined by recording spectra of the control honey sample on 18 days of analysis within 1.5 months. The small coefficients of variation indicate that instrumental conditions were reasonably stable over the duration of the measurements (**Table 1**).

Processing of Spectra and Multivariate Analysis. The spectra were converted into the GRAMS spc-format (GRAMS/32 AI vs. 6.0, Thermo

Honey type



Method

A: emission at 420 nm

B: excitation at 270 nm

C: excitation at 310 nm

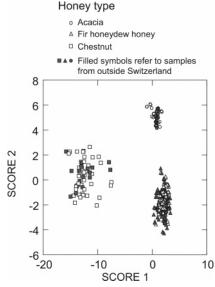


Figure 2. Scatterplot of canonical discriminant scores from method A (for better legibility, the scores of only three honey types are displayed).

Galactic, Salem, NH) for more convenience in the visual examination and data reduction. It was found that a normalization of the spectra

Table 1. Repeatability and Intermediate Precision of the Three Fluorescence Spectroscopic Methods

	method				
repeatability ($n = 6$)	A	В	С		
av I _{max} (au) ^a	179.8	185.8	100.9		
reproducibility s _r (au)	1.9	2.1	2.6		
coefficient of variation v_r (%)	1.1	1.1	2.6		
repeatability limit (r) (au)	5.4	5.9	7.5		
rel repeatability limit (%)	3	3.2	7.4		
intermediate precision ($n = 18$)					
av I _{max} (au)	203	195	106		
lab reproducibility s _L (au)	7.2	2.7	2.4		
rel lab reproducibility v ₁ (%)	3.6	1.4	2.2		

^a Arbitrary units.

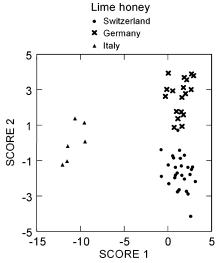


Figure 3. Scatterplot of canonical discriminant scores of lime honeys of different geographic origins (method A)

was not necessary and that the consideration of the fluorescence intensities can even improve the possibilities in discriminating the different unifloral honeys (40).

To avoid random noise resulting from instrumental effects, only the following spectral ranges were used for multivariate analysis: method A, 224–398 nm; method B, 290–500 nm; and method C, 333–547 nm. These ranges were also used for the combination of the spectra.

After elimination of spectral outliers, principal component analysis (PCA) was applied to eliminate the spectral collinearity and to reduce the number of variables to 20 PCs (using the PLSplus/IQ Add-on of GRAMS/32 AI vs. 5.09). This was performed separately for each type of spectra and each combination of different types of spectra.

In linear discriminant analysis (LDA), the 20 initial PCs were further reduced by backward elimination of PCs on the basis of their partial F values in the discriminant models (SYSTAT version 11, Systat Software Inc., Richmond, VA). The models were then optimized for maximum correct classification in jackknife classification. To account for the limited precision of single measurements, both spectra of each sample were used in the model of single types (A, B, and C) of spectra rather than the average. In the models using combined spectra, averaged spectra were used. The validation was carried out using spectra of one-third of the samples selected randomly and not present in the group of samples used to build the model.

Geographical Origin. The applicability of fluorescence spectroscopy for the determination of the geographical origin of honey was evaluated for the honey types when samples originating from different countries were available. The differences resulting from the geographic origin were studied within the groups of unifloral honeys by using MANOVA (SYSTAT version 11) as well as LDA and are visualized by plots of the canonical discriminant scores (Figures 3 and 4).

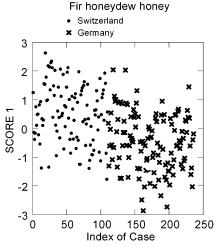


Figure 4. Scatterplot of the canonical discriminant score of fir honeydew honeys of German and Swiss provenance (method A).

RESULTS AND DISCUSSION

Repeatability. The repeatability of the three different methods was determined by a 6-fold measurement at the maximum intensity (I_{max}) of an acacia honey. With coefficients of variation (v_r) between 1.1 and 2.6%, the methods showed a good repeatability, which stayed in the same range over 43 days while the intermediate precision was determined (**Table 1**).

Fluorescence Spectra of Different Honey Types. The recorded fluorescence spectra at three different excitation and emission wavelengths for the 10 unifloral honey types considered are displayed in Figure 1 (for better legibility, the spectra of alpine rose and rape honey are not shown in Figure 1A as they are visually very similar to that of fir honeydew honey). Every spectrum is typical for a given honey type. The spectra obtained by the different methods were recorded using different aliquots of the same sample.

Excitation spectra were scanned from 220 to 400 nm with the emission measured at 420 nm (method A, **Figure 1A**). For most honey types two maxima at \approx 240 nm and between 340 and 360 nm, respectively, were observed, whereas lime honey exhibits its second maximum at \approx 365 nm. Most of the honey types investigated had their intensity within the same order of magnitude except for chestnut honey, which showed a nearly 2-fold intensity at the maximum. Metcalfa honeydew honey is also characterized by a more intense fluorescence. Dandelion honey shows an additional shoulder at \approx 300 nm.

For the spectra recorded using excitation at 270 nm (method B, Figure 1B), all honey types except chestnut, rape, and lime honeys exhibited broad and overlapping emission bands including at least two maxima located between 330 and 350 nm and between 400 and 440 nm, respectively. The very characteristic fluorescence spectrum of chestnut honey showed a much narrower band with two shoulders and a maximum at \approx 380 nm. Rape and lime honeys showed both maxima at \approx 350 nm, whereas the latter had a broader emission between 400 and 500 nm. Alpine rose honey showed a shoulder at \approx 310 nm and a maximum at \approx 340 nm. For heather, fir honeydew, dandelion, acacia, rape, and alpine rose honeys the intensities at the maxima ranged between 150 and 520 arbitrary units (au), whereas chestnut exhibited a considerably higher intensity of \approx 800 au. The lowest intensity was detected for lime honey. However, the intensities were found to vary considerably within the honey

Using an excitation wavelength at 310 nm (method C, **Figure 1C**) the spectra of chestnut honey again clearly differed from

Table 2. Percentage of Correct Classification by Using Single Data Sets at Different Excitation and Emission Wavelengths and by Combining of the Data of the Methods (Jackknife Classification by the Leave One Out Method and Validation with Independent Samples)

		A		В	C		combination of spectra from methods A and	
honey type jackknife	validation	jackknife	validation	jackknife	validation	jackknife	validation	
acacia	95	100	90	79	85	75	90	100
alpine rose	87	100	50	80	63	50	93	100
heather	98	88	100	100	91	100	100	100
chestnut	97	100	96	100	96	100	96	100
lime	96	100	98	100	98	100	95	100
dandelion	100	100	97	100	100	100	100	100
rape	88	100	100	100	93	43	95	100
fir honeydew	92	86	91	84	84	76	96	97
metcalfa honeydew	93	80	100	100	100	75	92	100
oak honeydew	100		100		78		100	
polyfloral	57	50	47	50	42	43	63	55
av (weighted)	87	85	84	83	80	73	90	91

^a Method A, excitation scanned 220 and 400 nm, emission measured at 420; method B, excitation at 270 nm, emission measured between 290 and 500 nm; method C, excitation at 310 nm, emission measured between 330 and 500 nm.

those of the other honey types investigated, especially by the 2-fold intensity compared to the others having a maximum at $\approx\!380$ nm. Most of the honey types exhibited a maximum at $\approx\!400$ nm and an intensity in the range from 100 to 200 au. Lime honey showed again the lowest intensity. Rape and acacia honeys were characterized by a shoulder at 365 nm. The maximum of the spectra of fir honeydew honeys was located at $\approx\!355$ and showed a shoulder at 420 nm. The band of the chestnut honey spectrum was narrower than that found by using method B and less intense but was, nevertheless, the most intense among the spectra recorded by using method C. The spectra of metcalfa honeydew honeys expressed a broad band with an intensity of $\approx\!350$ au, being thus the second most intense spectra.

It has been reported that chestnut honey, compared to the other honey types analyzed in this study, contains high amounts of hydroxycinnamates such as caffeic, *p*-coumaric, and ferulic acids as well as unidentified flavonoids (34, 44). Chestnut honey may also contain more phenylalanine than the other honey types analyzed in this study (39). The fluorescence of 2-aminoacetophenone, the main volatile component of chestnut honey, may also explain the characteristic spectra (45, 46).

Interestingly heather honey, commonly known to contain high amounts of phenolic compounds (47), does not show spectra of high fluorescence intensity compared to the other honey types. This may be due to scattering, reflection, and interference effects resulting from the numerous air bubbles present in heather honey.

LDA Applied to the Fluorescence Spectra. *Botanical Origin.* LDA was performed on the PCs of each type of spectra as well as on the combination of the two most significant types of spectra.

In the evaluation of single spectra the highest average classification rate (weighted according to the number of samples) of 85% in validation was obtained for method A (**Table 2**). The rates of correct classification were similar in both jackknife classification and validation, demonstrating that the models used were robust. Throughout the three methods studied, the classification rate for the polyfloral honeys was, at only 42–63%, very low. This can be explained by the lack of specific physical and chemical characteristics of this honey type. Thus, the polyfloral honeys are classified into the groups of unifloral honeys with the smallest Mahalanobis distance (**Table 3**).

For method A the lowest classification rate of 80% was observed for metcalfa honeydew honey. Twenty percent of the samples were misclassified as chestnut honey. This can be explained by the important nectar contribution of chestnut often present in metcalfa honeydew honeys. Due to the low number of samples (n = 5) used for validation, the 20% of misclassification arises from a single misclassified sample. In the validation step all samples of acacia, alpine rose, chestnut, lime, dandelion, and rape honeys were correctly classified. No validation was done for the oak honeydew honey due to the low number of samples available. In the jackknife classification some difficulties occurred in assigning alpine rose and acacia honeys (Table 3). Some samples of heather honey were also misclassified to rape and polyfloral honeys. Interestingly, a few samples of fir honeydew honey were classified as polyfloral or lime honeys. This could indicate that the value of 0.8 mS cm⁻¹ in electrical conductivity is not always adequate to discriminate between polyfloral and honeydew honeys. Lime honeys very often contain some honeydew honey, which complicates their characterization.

Even though samples originated from different geographical origins, they were correctly classified according to their botanical origin. Irrespective of their geographical origin the fluorescent characteristics of honey from various botanical origins seem to be uniform, as samples from outside Switzerland group among the samples from Switzerland (**Figure 2**; for better legibility, the scores of only three different honey types are displayed).

The overall discriminating potential of method B is comparable to that of method A (**Table 2**). However, for the discrimination between alpine rose and acacia honeys more difficulties were encountered using method B than method A. In spite of the fact that the two groups were mingled, some samples of alpine rose honey were even misclassified as polyfloral honeys (data not shown).

The potential of method C for the classification of both unifloral and polyfloral honeys by using a single discriminant model was clearly inferior to that of methods A and B. Besides the difficulties already mentioned for alpine rose and acacia honeys, a considerable number of samples belonging to the groups of rape and honeydew honeys were not correctly classified in validation (**Table 2**).

To evaluate whether the rate of correct classification could be further increased by combining two of the most promising types of spectra, the ones of methods A and B were averaged

Table 3. Jackknife Classification and Validation Table for the Honey Samples Classified by LDA on the Spectra of Method A

		jackknife classification rate for method A (%)									
	acacia	alpine rose	heather	chestnut	lime	dandelion	rape	fir honeydew	metcalfa honeydew	oak honeydew	polyfloral
		1036	Healifei	CHESTILL	IIIIIC	dandellon	таре	noneyaew	noneyaew	noneyaew	Polylloral
acacia ($n = 21$)	95	5	0	0	0	0	0	0	0	0	0
alpine rose ($n = 19$)	11	87	0	0	0	0	0	0	0	0	3
heather $(n=23)$	0	0	98	0	0	0	0	0	0	0	2
chestnut ($n = 29$)	0	0	0	97	0	0	0	0	2	2	0
lime $(n=26)$	0	0	0	0	96	0	0	0	0	0	4
dandelion ($n = 18$)	0	0	0	0	0	100	0	0	0	0	0
rape $(n = 24)$	0	0	0	0	0	0	88	0	0	0	12
fir honeydew ($n = 120$)	0	0	0	0	1	0	0	92	0	0	8
metcalfa honeydew ($n = 14$)	0	0	0	7	0	0	0	0	93	0	0
oak honeydew $(n = 8)$	0	0	0	0	0	0	0	0	0	100	0
polyfloral ($n = 65$)	0	9	2	3	9	5	7	9	0	0	57

	classification rate in validation for method A (%)										
		alpine fir metcalfa									
	acacia	rose	heather	chestnut	lime	dandelion	rape	honeydew	honeydew	polyfloral	
acacia (n = 7)	100	0	0	0	0	0	0	0	0	0	
alpine rose $(n = 6)$	0	100	0	0	0	0	0	0	0	0	
heather $(n=8)$	0	0	88	0	0	0	6	0	0	6	
chestnut $(n = 10)$	0	0	0	100	0	0	0	0	0	0	
lime $(n=9)$	0	0	0	0	100	0	0	0	0	0	
dandelion ($n = 6$)	0	0	0	0	0	100	0	0	0	0	
rape $(n=7)$	0	0	0	0	0	0	100	0	0	0	
fir honeydew ($n = 40$)	0	0	0	0	3	0	0	86	0	11	
metcalfa honeydew ($n = 5$)	0	0	0	20	0	0	0	0	80	0	
polyfloral ($n = 22$)	0	7	0	9	0	0	20	14	0	50	

and concatenated for each sample. The rate of correct classification increased for alpine rose, fir, and spruce honeydew and even for polyfloral honeys compared to the results obtained by using the individual methods A and B (**Table 2**).

The classification tables revealed that polyfloral honeys were very often classified into the groups of the unifloral honeys, whereas the latter were rarely misclassified into the one of the polyfloral honeys. This observation led to the development of a two-step procedure. In the first step the sample was attributed to one of the 11 honey types considered using an overall discriminant model including all honey types. In the second step this classification was verified by using one or several twogroup models consisting of a group formed by samples of a given unifloral honey versus a group called "non-unifloral", consisting of all the other samples. Each two-group model was separately built using LDA backward elimination and forward selection. For the verification of the classification by the first model, at least the two-group model of the corresponding honey type was used. In addition, one to six two-group models (entries in boldface type in Table 3) were used when a misclassification rate of >3% was calculated in jackknife classification or validation tables of the overall model.

The classification rates for the unifloral honeys in the two-group models were generally >90%, whereas the classification rate for the polyfloral honeys ranged between 48 and 75% (**Table 4**). However, as far as the polyfloral honeys are concerned, this is not very important, as we are principally interested in the authentication of unifloral honeys. The high rates of correct classification for both the unifloral and non-unifloral groups considered by the two-group models indicate that the botanical origin can be reliably determined according to this procedure. The respective error rates of this two-step procedure using methods A and B as well as the combination of the two former types of spectra were calculated by applying Bayes' theorem on the conditional probabilities of disjoint events.

Method A gave again the most promising results with an error probability (wrong classification of a sample of unknown botanical origin) of <5% for all honey types except for fir honeydew, for which it was 10% (**Table 5**). The error probabilities by using method B were higher for all honey types compared to those using method A except for the determination of metcalfa honeydew honey. By using the combination of the spectra of methods A and B, the error probability could be reduced to <5% (in validation) for the 11 honey types studied. It is interesting to note that the error probabilities of the honey types that express the highest variability in physical and chemical characteristics such as lime and fir honeydew honeys are the highest in fluorescence spectroscopy as well. This can be interpreted to indicate that fluorescence spectroscopy reproduces well the characteristics of classical criteria.

Geographical Origin. Differences in geographical origin were studied within the groups of samples of the same botanical origin when samples were available from at least two countries. Interestingly, a statistically significant difference was found by MANOVA between the geographical origins of all honey types studied (Table 6). The lime honey samples originating from Switzerland, Germany, and Italy formed groups in the plot of discriminant scores according to their geographical origin (Figure 3). The samples could also be correctly classified by LDA according to their geographical origin except for one Swiss sample that was classified to German provenance (data not shown). However, the classification according to geographical origin could be observed only within the groups of honeys of the same botanical origin. An LDA model of acacia, lime, dandelion, and fir honeydew honeys of German and Swiss origin failed to classify the samples according to their geographical provenance (Table 7). This clearly indicates that the characteristics resulting from the botanical source are considerably stronger than the geographical aspects. The sample set of the lime honeys was small; a larger sample set would possibly lead to a less pronounced difference. This may be illustrated on the

Table 4. Jackknife and Validation Tables for the Honey Samples Classified by the Two-Group Discriminant Models of Methods A and B and the Combination of These Spectra

		jackknife cl	assifica	ation	١	/alidation
	·	unifloral	noi	n-unifloral		unifloral
	n	correct classifi- cation (%)	n	correct classifi- cation (%)	n	correct classifi- cation (%)
		Metho	d A			,
acacia	21	100	343	96	7	100
alpine rose	19	100	345	90	6	83
heather	23	98	341	99	8	88
chestnut	29	97	335	99	10	90
lime	26	100	338	97	9	100
dandelion	18	100	346	98	6	100
rape	24	91	343	93	7	100
fir honeydew	120	95	244	93	40	88
metcalfa honeydew	14	100	350	100	5	100
oak honeydew polyfloral	8 65	94 74	356 300	99 65	22	48
polylloral	00			63	22	40
		Metho				
acacia	21	100	341	96	7	100
alpine rose	16	97	346	89	5	80
heather	23	100	340	95	8	100
chestnut	28	96	335	99	9	100
lime dandelion	26 19	100	336 343	95	8	100
	22	100		94	6 7	100
rape	120	100 92	340 242	98 92	40	100 84
fir honeydew metcalfa honeydew	120	100	350	92 98	40	88
oak honeydew	8	100	354	96 95	4	00
polyfloral	68	65	294	72	22	75
. ,				. –		70
				ethods A and B	7	100
acacia alpine rose	20 15	100 100	327 332	97 91	7	100 100
heather	23	100	324	100	8	100
chestnut	23 27	96	320	100	9	100
lime	26	100	321	96	9	100
dandelion	19	95	328	98	6	100
rape	21	100	326	98	7	100
fir honeydew	117	97	230	93	39	100
metcalfa honeydew	12	100	335	98	4	100
oak honeydew	8	100	339	99	•	
polyfloral	59	69	288	71	15	75

example of the fir honeydew honeys from Germany and Switzerland, for which a classification according to geographical origin was not possible (**Figure 4**). However, the samples of fir honeydew honeys originated from an area of ≈ 300 km in diameter belonging to Switzerland and Germany and therefore having very similar vegetation.

In future studies it should be verified if the geographical origin of honey could be determined by fluorescence spectroscopic techniques on the basis of the minor contributions of accompanying flora that may be different in areas distant enough. The chemometric models should also be validated with samples of polyfloral provenance.

Adulteration by Feeding of Bees. The acacia and chestnut honey samples adulterated with as much as 50% of artificial honey did not show any comprehensible changes in the spectra compared to the pure samples in any of the three methods studied. Generally, the spectra of the adulterated samples remained in the range of the natural variation of the corresponding unifloral honeys. A detection of honey adulteration is therefore not possible except if the adulterant contains a characteristic fluorophore.

Conclusion. Although absolutely pure unifloral honeys do not exist, the definition of unifloral honey is in fact based on the points of view and the descriptions of different analysts. However, a consensus has been reached using the physical, chemical, and pollen analytical characteristics of the unifloral honeys considered as internationally recognized criteria already published (3-6).

Of capital importance is certainly to ensure a uniform honey quality that can be recognized by consumers preferring a given type of honey. Currently, the determination of the botanical origin of honey relies on the judgment of experienced experts who base their decision on the criteria of several analytical measurands. The challenge of new analytical methods that do not need such an expertise is to mathematically model and reproduce this decision-making process. As the definition of a unifloral honey is ultimately a matter of opinion, absolutely correct classification by chemometric models can therefore not be expected as these models are trained by uncertain sample sets as reference.

As the characteristic physical and chemical differences between unifloral and polyfloral honeys are small and only a very few compounds are specific to a given type of honey, the chemometric approach based on a fingerprint seems to be more promising than the search for individual marker compounds.

This study shows that front-face fluorescence spectroscopy combined with chemometrics offers a promising approach to the authentication of the botanical origin of honey and that the problems related to the determination of the polyfloral honeys can be overcome by the successive use of at least two mathematical models. The current results show that classical criteria commonly used for the determination of the botanical origin of honey can be very well reproduced by front-face fluorescence spectroscopy and chemometrics. Depending on the certainty needed, one may base the classification on the single spectra of type A or combine the spectra of methods A and B.

Table 5. Error Probabilities for the Classification of Unifloral and Polyfloral Honeys by the Different Methods

	meth	nod A	meth	nod B	combination of spectra from methods A and B		
honey type	jackknife	validation	jackknife	validation	jackknife	validation	
acacia	0.029	0.006	0.109	0.030	0.022	0.005	
alpine rose	0.016	0.003	0.058	0.018	0.009	0.001	
heather	0.044	0.013	0.050	0.051	0.003	0.003	
chestnut	0.003	<10 ⁻³	0.034	0.053	<10 ⁻³	<10 ⁻³	
lime	0.067	0.037	0.096	0.054	0.075	0.039	
dandelion	0.037	0.008	0.075	0.072	0.021	0.019	
rape	0.003	0.002	0.046	0.070	0.033	0.042	
fir honeydew	0.088	0.099	0.107	0.090	0.047	0.045	
metcalfa honeydew	0.040	0.004	0.004	0.002	<10 ⁻³	<10 ⁻³	
oak honeydew	0.044		0.050		0.006		
polyfloral	<10 ⁻³	<10 ⁻³	0.034	0.031	<10 ⁻³	<10 ⁻³	

Table 6. Results from MANOVA for the Geographical Origin of the Different Unifloral Honeys (Method A)

honey type, country ^a	Wilks' λ	ρ
acacia (CH, D, F)	0.009	<10 ⁻³
alpine rose (CH, I)	0.027	<10 ⁻³
fir honeydew (CH, D)	0.696	<10 ⁻³
chestnut (CH, F, I)	0.001	<10 ⁻³
lime (CH, D, I)	0.004	<10 ⁻³
dandelion (CH, D, I)	0.023	<10 ⁻³

^a Country codes: CH, Switzerland; D, Germany; F, France; I, Italy.

Table 7. Percentage of Correct Classification According to the Geographical Provenance by Using the Data Set of Method A

	jackknife classification matrix ^a						
	Switzerland	Germany	correct (%)				
Switzerland	117	69	63				
Germany	59	109	65				
total	176	178	64				

^a Jackknife classification by the leave one out method on samples from acacia, lime, dandelion, and honeydew honeys from spruce and fir.

Of course, the proposed fluorescence spectroscopic method needs a considerable amount of preliminary work to establish the chemometric models based on samples of known botanical origin. Once the classification models have been set, the technique enables a rapid determination of the botanical origin without particular sample preparation and special qualification of laboratory personnel. It remains to be tested by future studies if these models can be transferred from one instrument to another, as in infrared spectroscopy when normalized fluorescence spectra are used (40) or the instruments are calibrated with reference materials.

In addition, the present work clearly shows that fluorescence characteristics of honey are much more dependent on their botanical origin than on the geographical origin. Therefore, the former should be determined before a method is proposed for the determination of the geographical origin of honey. Such a method must be tested as well with samples of the same botanical origin.

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