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Optimisation of a sample preparation procedure for the screening of fungal infection and assessment of deoxynivalenol content on maize using mid-infrared attenuated total reflection spectroscopy

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**Optimisation of a sample preparation procedure for the screening of
fungal infection and assessment of deoxynivalenol content of maize
using mid-infrared attenuated total reflection spectroscopy**

For Peer Review Only

Abstract

A sample preparation procedure for the determination of deoxynivalenol (DON) with mid-infrared spectroscopy is presented. Attenuated total reflection was employed with a mid-infrared spectrometer. Repeatable spectra were obtained from samples featuring a narrow particle size distribution. Samples were ground with a centrifugal mill and analysed with an analytical sieve shaker. Particle sizes of < 100 μm , 100-250 μm , 250-500 μm , 500-710 μm and 710-1000 μm were obtained. Repeatability, classification and quantification abilities for DON were compared with non-sieved samples. The fraction "100-250 μm " showed the best repeatability: The relative standard deviation of spectral measurements improved from 20% to 4.4%. 100% of sieved samples were correctly classified compared with 79% of non-sieved samples. The DON content in analysed fractions was a good estimate of the overall toxin content.

Keywords

attenuated total reflection, fungi, cereals, mycotoxins, sample preparation

Introduction

The determination of fungal invasion of grain samples is of vital interest to food and feed producers as well as consumers alike. Food and feed that is contaminated by fungi may not only be unsuitable for food and feed processing, but can also be dangerous to human and animal health (Council for Agricultural Science and Technology 2003).

Secondary fungal metabolites are often toxic and can cause serious health problems. In the moderate climatic zones of North America and Europe *Fusarium* fungi are among the most prevalent agriculturally important fungal species (Charmley et al. 1994). The main toxic secondary metabolite produced by these genera during the growth period of the host commodity is deoxynivalenol (DON), which causes feed refusal by swine, skin damage, vomiting and hemorrhages, if consumed (D'Mello et al. 1999).

Besides regulations for aflatoxins and ochratoxin A (EC 466/2001), maximum levels for *Fusarium* toxins in selected foodstuffs were recently set in the European Commission Regulation No 856/2005 which applies from July 1st 2006. Maximum levels for DON have been set e.g. at 700 $\mu\text{g kg}^{-1}$ for cereal flour including maize flour. In the US, the Food and Drug Administration (FDA) has set advisory levels for DON to as low as 1000 $\mu\text{g kg}^{-1}$ (Trucksess 1995).

The determination of DON itself (Krska et al. 2001) is well established, but tedious and labour intensive. Several sample preparation steps are necessary before separation and detection can take place using chromatographic methods such as LC-MS or GC-ECD.

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4 Determination could be sped up considerably with a rapid method. There is indeed a
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6 need for such a method, when considering the amounts of cereals that are processed by
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8 the food and feed industry each year. Food imports are subjected to frequent checks by
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10 authorities in Europe and the US and major entry points such as harbours and border
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12 crossings are facing an increased workload. A method for the determination of fungal
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14 contaminants within minutes is highly desirable in order to avoid long waiting times and
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16 increase the overall percentage of controlled goods.
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23 So far, a small number of rapid methods have been proposed in order to assess the
24
25 amount of fungal contamination in a given sample and not always the amount of toxic
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27 DON which is the more relevant parameter. Visual inspection is widely used for the
28
29 assessment of maize cobs (Bechtel et al. 1985). The method is quick and cheap to
30
31 perform, but labour-intensive and requires a lot of experience. DON levels are not
32
33 directly estimated and the amount of fungus present on the cob is not always a reliable
34
35 indicator for quantification, as it often resides inside the kernel (Greene et al. 1992).
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37 Samples that were already ground need more sophisticated investigation techniques,
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39 because a visual inspection is not possible.
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47 The use of infrared spectroscopic techniques has already been proposed in previous
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49 studies as spectra can be easily recorded (Gordon et al. 1997; Dowell et al. 1999;
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51 Gourama and Bullerman 1995; Petterson and Aberg 2002). Delwiche and Gaines
52
53 (2005) investigated visible and near-infrared (NIR) wavelengths for the determination
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55 of *Fusarium* head blight in winter wheat kernels. Near infrared instruments are popular
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4 and well-established in the analysis of foods and feeds, because the radiation penetrates
5 through glass containers (Berardo et al. 2005). With Fourier Transform infrared
6 (FTIR) instruments in the mid-infrared range, measurements can be performed very fast
7 with an excellent signal to noise ratio and high-resolution power. In recent years
8 several accessories have been developed that enable the measurements of solid samples
9 without the need for extensive sample preparation (Harrick 1967). Photoacoustic
10 infrared (IR-PAS) and diffuse reflection spectroscopies (IR-DR) have been employed
11 for the detection of fungal species on maize and cereals (Greene et al 1992) and method
12 performance was discussed previously (Kos et al. 2003; Kos et al. 2002). In these
13 papers, the successful application of mid-infrared spectroscopy with attenuated total
14 reflection (ATR) for the detection of fungal contamination on maize was described. The
15 sample is placed on a horizontally mounted diamond ATR crystal and the mid infrared
16 ATR spectrum is recorded. Maize was chosen as a commodity, because it is almost
17 exclusively infected by *Fusarium* fungi during the growth period in moderate climates,
18 thus minimizing contamination from other fungal species and making it a good model
19 system to study.
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45 The optimisation of the sample preparation procedure is a crucial point in obtaining
46 reproducible and representative spectra. Cereal samples that are infected with fungal
47 species have certain properties that make ATR measurements a challenging task with
48 respect to repeatability of spectral measurements and prediction ability of the
49 contaminant:
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4 The distribution of fungal contamination (and therefore its toxin) is not homogenous
5 over the sample batch (Park and Pohland 1989). As consequence a small number of
6 highly infected kernels can make a whole batch unsafe for food processing. The
7 differences between the blank and the contaminated samples are small and usually not
8 visible by the eye. PAS investigation of single kernels with high contamination levels
9 revealed changes to the spectrum and in ATR spectra only highly contaminated samples
10 show visible changes. Several spectral regions have to be excluded from calculations,
11 because they either show high variability not related to fungal contamination (e.g. the
12 OH moisture band and the ubiquitous, non-specific methylene bands)
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28 Instrumental limitations can also cause the following problems:
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30 After sample grinding a rather broad particle size distribution from $> 1000 \mu\text{m}$ to
31 $< 100 \mu\text{m}$ in diameter is observed, which lead to repeatability problems in spectral
32 measurements without any further pre-treatment (see Figure 1). Because of the small
33 area of the ATR crystal, large sample particles may take up large amounts of space on
34 the crystal, leading to over-representation in resulting spectra.
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50 Large particles could also cause varying pressure conditions on the ATR crystal, when
51 using a pressure applicator, which ensures full contact between the sample and the
52 crystal. The range between 1800 and 2200 cm^{-1} is dominated by absorption of the
53 diamond crystal and has to be removed from all datasets.
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7 This paper describes a sample preparation procedure that was developed in order to
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9 obtain representative samples that are suitable for ATR measurements and contain
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11 information about infection, which is accessible by mid infrared measurements. Results
12
13 are compared to samples, which have not been treated, but were directly measured. The
14
15 problems mentioned above are discussed and in order to overcome these problems
16
17 sample preparation steps are presented.
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23 **Materials and methods**

24 *Sample preparation*

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26 All maize samples were chosen from the same genotype (RWA2) and were
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28 predominantly naturally infected with *Fusarium graminearum* in the field during the
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30 growth period. Thus, samples were as close as possible to real life samples, but still
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32 represented a very defined model system that eliminated any variability from
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34 parameters such as genotype variation and variation due to differing fungal species,
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36 providing a solid basis for the investigation of the repeatability of measurements.
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44 After harvest, maize cobs were dried for 3 days by blowing air at 40°C through the
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46 storage container, similar to storage protocols used in commercial storage facilities.
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48 Thus, the moisture content was reduced to <15%, which stopped fungal growth
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50 (Schlegel 1993). After drying, whole kernels were separated from the cob and ground
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52 with a ZM 100 centrifugal mill (Retsch Haan, Germany) based on impact and shearing
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54 effects between a rotor and a fixed ring sieve. The sample kernels are pre-crushed on the
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4 wedge-shaped rotor teeth before further milling takes place between rotor and screen
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7 (see Figure 2).
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11 [Insert Figure 2 about here]
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16 Grinding of 50-100 g of sample was performed very quickly, usually in less than a min.
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18 The hole width of the ring sieve determined the upper particle size limit after grinding,
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20 which was collected in the cassette. Rotor speed was 14 000 rpm. In this study a 1 mm
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22 stainless steel ring sieve was employed for all experiments.
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28 After grinding, parts of the sample were directly investigated with Fourier Transform
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30 Infrared (FTIR) Spectroscopy (see below) and for the second fraction (usually 30-50 g)
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32 the particle size distribution was determined by sieving with an analytical sieve shaker
33
34 (Retsch AS200/Haan, Germany). Sieves with five different mesh sizes (< 100 μm , 100-
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36 250 μm , 250-500 μm , 500-710 μm and 710-1000 μm) were employed. The sieving
37
38 module was operated at an amplitude of 2 mm for a duration of 30 min using a 3-
39
40 dimensional throw movement. Two latex spheres were added to each sieve to minimise
41
42 clogging. The particle size distribution was calculated prior to FTIR measurements.
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49 *Attenuated total reflection spectroscopy*

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51 All FTIR measurements were performed with a Bruker Vector 22 mid-infrared (MIR)
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53 spectrometer (Bruker Optics/Karlsruhe, Germany). A global light source emitted
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55 broadband radiation that was coupled into a pendulum interferometer. After passing
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4 through the sample chamber, the radiation was detected with a L-alanine doted
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6 DTGS detector. The instrument is portable, because the chosen set-up requires no
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8 maintenance or service lines, making the spectrometer suitable for in-the field
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10 measurements.
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15 Because of the unique properties of ATR spectroscopy for the easy measurement of
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17 solid samples, a horizontal ATR unit (SensIR-Technologies/Danbury, CT, US) was
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19 mounted in the sample chamber. It was fitted with a diamond ATR crystal with
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21 3 internal reflections aligned to a ZnSe-crystal, which acted as a focusing element. The
22
23 robustness and hardness of the crystal allowed for the use of a pressure applicator with a
24
25 torque knob to ensure close contact between sample and crystal and the application of a
26
27 reproducible pressure.
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35 Spectra were collected at a resolution of 4 cm^{-1} and with a spectral range between
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37 650 and 4500 cm^{-1} . Background measurements were made against air by co-adding
38
39 32 scans for each spectrum. The interferometer was operated at a mirror speed of
40
41 10 kHz and in the double-sided-bidirectional-mode. Fourier transformation was
42
43 performed with a Mertz phase correction, a Blackman-Harris 3 term apodisation
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45 function and a zero filling factor of 2. No filters were applied. Each spectrum consisted
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47 of 1997 data points and was stored in a separate data file for further data processing.
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53 54 *Reference methods*

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56 In order to evaluate results from spectral measurements, especially for the estimation of
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4 differences between contaminated and blank samples, data from conventional methods
5 for DON was obtained. Extraction, clean-up, separation and detection were performed
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7 with a method established using GC-ECD detection (Weingaertner et al. 1997): After
8
9 extraction of ground wheat samples using an acetonitrile/water mixture, the filtered
10
11 extracts were cleaned-up using Mycosep #225 columns (Romerlabs, Tulln, Austria).
12
13 After evaporation, the residue was derivatized using Sylon BTZ (Supelco, Vienna,
14
15 Austria) and re-extracted into isooctance followed by analysis with a gas chromatograph
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17 with electron capture detection (GC-ECD) with Mirex as an internal standard. The limit
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19 of detection for DON in wheat was $65 \mu\text{g kg}^{-1}$. The mean recovery was 96.6% with a
20
21 standard deviation of 9.5% for 47 samples spiked with DON. All samples were
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23 measured in a quality controlled environment.
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33 *Data analysis*

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35 Differences between spectra of blank and contaminated spectra are not easily detected,
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37 because of their high similarity. Therefore, it is necessary to reduce the variability of
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39 measurements that can not be related to fungal infection and additionally use the
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41 potential of multivariate data analysis for the detection of differences by employing
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43 several variables (i.e. wavelengths) for spectral calculations (Naes et al 2002).
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49 All calculations were performed with Unscrambler® 7.5 (Camo/Oslo, Norway) and
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51 Datalab (H. Lohninger/Vienna, Austria) (Lohninger 2000). Repeatability calculations
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53 were carried out by the spectrometer software and in MS Excel (Microsoft/Redmond,
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55 OR, US). A minimum of 10 individual spectra was recorded from each sample, always
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4 using a new subsample. The fingerprint area (the range between 1800 and 800 cm^{-1})
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6 was chosen for all calculations. The final number of datapoints varied from 300-500 for
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8 each spectrum depending on the experiment. After appropriate data treatment
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10 (averaging, normalisation, derivative calculations) and principal component analysis
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12 (PCA) for the de-correlation of variables (Danzer et al. 2001), classification was
13
14 performed using cluster analysis (CA). A quantification model was built with a partial
15
16 least squares (PLS1) regression model. In order to assess the quality of the classification
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18 full cross validation was calculated, which is especially suited for the evaluation of
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20 small data sets, and the root mean square error of cross validation (RMSECV)
21
22 determined (Naes et al. 2002). All results were cross-checked with trends obtained from
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24 raw data in order to detect artefacts caused by data treatment. A detailed description of
25
26 the employed multi-variate methods can be found in an earlier publication (Kos 2003).
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35 **Results and discussion**

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37 The investigation of single kernels with ATR spectroscopy is theoretically possible, but
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39 not recommended. Because of the inhomogeneous distribution of fungus and toxin in a
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41 sample batch, a very large number of kernels would need investigation before any
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43 sound conclusions could be made for the whole batch and therefore sampling remains
44
45 one of the central problems for representative mycotoxin analysis (Champeil et al.
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47 2004). Therefore it is advisable to grind the sample in order to make it more
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49 representative of the lot (Garfield 1989).
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56 In a first approach spectra were directly recorded from ground samples without any
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4 further sample preparation. Spectra for 10 repetitive measurements (each measured
5 from a new sub-sample) are displayed in Figure 1. Band assignments are displayed in
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7 Table I. The rectangle marks the spectral range used for all calculations. The plots
8
9 clearly illustrate the need for improvement, which is also reflected in a PCA score/score
10 plot (Figure 3) and subsequent classification using cluster analysis (CA). Score plots
11 already reveal a trend by the formation of two clusters, but correct classification of these
12 clusters with CA is not possible (Figure 4): Contaminated samples with DON
13 concentrations of 557, 1024 and 2596 $\mu\text{g kg}^{-1}$ could not be classified correctly. Sample
14 names in plots give DON concentrations in $\mu\text{g kg}^{-1}$ obtained from reference
15 measurements. Individual blank measurements are marked with “bl_” and an
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17 identification number.
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33 [Insert Table I about here]

34 [Insert Figure 3 about here]

35 [Insert Figure 4 about here]

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42 The lack of repeatability could be due to the broad particle size distribution of the
43 sample. A number of factors that might influence the repeatability of ATR
44 measurements of particles that differ in diameter, when carrying out particle size
45 measurements of silica particles have been described (Yoshidome et al. 1998). In order
46
47 to overcome these limitations, the samples were subjected to particle size analysis.
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56 The weight of fractions after sieving was determined and is displayed in Figure 5. The
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4 diagram also displays results for different contamination levels of DON in order to
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6 detect influences of the fungal contamination on the particle size distribution after
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8 grinding. The largest particle fraction had a particle size between 100 and 250 μm in
9
10 diameter, containing between 40 and 50% of the particles (w/w). The fraction between
11
12 250 and 500 μm contained an additional 20 to 30%, which was still enough for a
13
14 suitable number of repetitions for spectral measurements. The remaining fractions held
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16 only small amounts of material, usually not enough for ATR measurements.
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23 The improvement in the repeatability of spectral measurements was estimated by
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25 calculating the relative standard deviation (RSD, %) at five different wavelengths in the
26
27 spectrum. From Figures in Table II it can be concluded that the repeatability of spectral
28
29 measurements improves with a narrowing particle size distribution and smaller particle
30
31 diameters. For the non-sieved sample a relative standard deviation (RSD) of almost
32
33 20% was calculated. This RSD decreased with the particle size reaching a low at 1.7%
34
35 for the fraction with a particle diameter $< 100 \mu\text{m}$. From this point of view it was
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37 therefore obvious to choose the 250 -100 μm fraction for all subsequent measurements
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39 in this study, because it provided enough sample material for multiple repetitions (up to
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41 15-25 g) and additional reference measurements and featured a satisfying repeatability
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43 of measurements (4.4%). The smallest fraction ($< 100 \mu\text{m}$) did not yield enough sample
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45 in order to carry spectral measurements for all samples.
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4 A score/score plot after PCA shows a good separation between the clusters of blank and
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6 contaminated samples (Figure 6), following the trend of the non-sieved sample. The
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8 dendrogram after CA (Figure 7) calculated from the first two principal components
9
10 reflected this improvement and all samples could be assigned correctly into two classes
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12 (contaminated/not contaminated).
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18 [Insert Figure 6 about here]
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21 [Insert Figure 7 about here]
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26 Five samples with their sieving fractions were chosen from all available samples and, if
27
28 possible the DON content was determined in each of the fractions in order to assess,
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30 which fraction contained which amount of the reference analyte. Blank and
31
32 contaminated samples with low (approx. $300\mu\text{g kg}^{-1}$), mid (approx. $1000\mu\text{g kg}^{-1}$) and
33
34 high (approx. $2600\mu\text{g kg}^{-1}$) DON content were selected as a model system. All
35
36 fractions that had enough sample left (2-10 g) after performing IR measurements were
37
38 analysed for their DON content using the conventional procedure described above.
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41 Table III summarises the results and illustrates that the DON content in the fraction
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43 chosen for ATR measurements gives a good estimate for the DON in the original (non-
44
45 sieved) sample.
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52 [Insert Table III about here]
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56 **Conclusions and future work**

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4 The results obtained here clearly illustrate the improvements that were achieved by
5
6 applying particle size analysis to maize samples before ATR mid infrared measurements
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8 for the detection of fungal infection were carried out. Comparability is ensured, because
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10 all sub-samples in both datasets (not sieved and sieved) came from the same sample.
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12 The sample set was very defined (from a single maize genotype [RWA2] and fungal
13
14 infection with *F. graminearum*), minimising variation from other sources than effects
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16 on spectral measurements by the particle size distribution.
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23 By using sieved samples with a particle size fraction between 100 and 250 μm , it was
24
25 possible to classify all 14 samples of the dataset in a concentration range of the
26
27 reference analyte DON from 310 $\mu\text{g kg}^{-1}$ to 2596 $\mu\text{g kg}^{-1}$. With the dataset containing
28
29 non-sieved samples only 11 out of 14 samples were assigned correctly. The relative
30
31 standard deviation of spectral measurements could be reduced from almost 20% to
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33 < 5% for the chosen fraction, which is suitable and sufficient from an analytical point of
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35 view. A suitable size fraction has to be chosen after grinding or alternatively milling to
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37 the desired particle size has to be performed. It was shown for which particle sizes
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39 measurement variability is lowest, leading to stable and repeatable spectra. From
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41 reference measurements it was shown that the DON content in the investigated fraction
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43 is a good estimate of the overall DON content.
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51 The current approach is not intended to be used directly in feed and food control, but
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53 rather demonstrates the effect of grinding and particle size for the determination of
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55 DON. The sieving procedure and the selection of a suitable particle size fraction
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4 eliminated a major source of unwanted variation in spectral measurements. Future work
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6 can therefore focus on the further optimisation of the ATR method. The investigation of
7
8 additional sources that influence mid-infrared spectra is now possible. The (much
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10 smaller) variation caused by different maize genotypes and infection by several fungal
11
12 species can now be investigated more easily, leading to a stable model for the
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14 estimation of the DON content.
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21 **Acknowledgements**

22
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24
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26
27 study.
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Tables

Table I. Band assignments for a mid-infrared ATR spectrum of maize.

Table II. Relative standard deviation [%] of spectral measurements from a single sample.

For each fraction datapoints were taken from 5 different wavelengths and 10 repetitions of the ATR measurement, each of which was performed with a new subsample.

Table III. DON reference data [$\mu\text{g kg}^{-1}$] for particle size fractions. Detection limit of the reference method was $65 \mu\text{g kg}^{-1}$. (n.d. – not detected, n.m. – not measured, lack of sample).

Figures

Figure 1. Repeatability of mid-infrared ATR spectra of maize (n=10). The sample was ground only, without any further pre-treatment. Band assignments are displayed in Table I. The rectangle marks the spectral range used for all calculations.

Figure 2. Set-up of the ZM 100 centrifugal mill (view from top). (1) feed, (2) rotor teeth, (3) fixed ring sieve, (4) collector, (5) mill housing.

Figure 3. PCA of averaged ATR spectra that were not sieved before spectral measurements. Sample numbers in plot give DON concentrations in [$\mu\text{g kg}^{-1}$]. Each datapoint was averaged from 10 repetitive measurements.

Figure 4. Cluster Analysis of the first 2 principal components calculated from the PCA results obtained from Figure 3 (samples not sieved).

Figure 5. Particle size distribution after sieving of ground maize samples. Different levels of fungal contamination (expressed as DON content) were investigated for their behaviour during the grinding process.

Figure 6. PCA of averaged ATR spectra that were sieved before spectral measurements. Particle size fraction was between 250 and 100 μm . Sample numbers in plot give DON concentrations in [$\mu\text{g kg}^{-1}$]. Each datapoint was averaged from 10 repetitive

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4 measurements.
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9 Figure 7. CA of the first 2 principal components calculated from the PCA results
10 obtained from Figure 6 (sieved samples).
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Table I. Band assignments for a mid-infrared ATR spectrum of maize.

Wavenumber [cm ⁻¹]	No in fig. 2	Assignment of Bands	Relative Intensity
3300	(1)	Hydroxyl stretching	Strong
2930, 2855	(2)	Methylene stretching	Strong
2325 - 1900	(3)	Diamond crystal absorptions	Strong
1745	(4)	C=O stretching	Weak
1540	(5)	NH bending/CN stretching	Weak
990	(6)	C-O stretching	Strong

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Table II. Relative standard deviation [%] of spectral measurements from a single sample. For each fraction datapoints were taken from 5 different wavelengths and 10 repetitions of the ATR measurement, each of which was performed with a new subsample.

Wavelength [cm ⁻¹]	Fraction: Particle size [μm]					
	Not sieved	>710	710-500	500-250	250-100	<100
815.8	18.4	7.05	6.93	3.37	3.42	1.33
997.1	20.0	23.1	25.4	13.6	4.12	4.71
1336	18.8	5.52	5.79	2.86	3.61	0.85
1643	20.4	6.81	6.68	2.22	3.61	1.34
1839	21.8	0.92	0.72	0.47	7.08	0.21
Mean	19.9	8.68	9.10	4.49	4.37	1.69

Table III. DON reference data [$\mu\text{g kg}^{-1}$] for particle size fractions. Detection limit of the reference method was $65 \mu\text{g kg}^{-1}$. (n.d. – not detected, n.m. – not measured, lack of sample).

Sample	Fraction: Particle size [μm]					
	Original	>710	710-500	500-250	250-100	<100
bl_2	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.
bl_5	n.d.	n.d.	n.d.	n.d.	n.d.	140
309	309	375	488	380	263	n.m.
1173	1173	n.m.	1290	1179	1185	n.m.
2596	2596	1979	2251	3452	4529	n.m.

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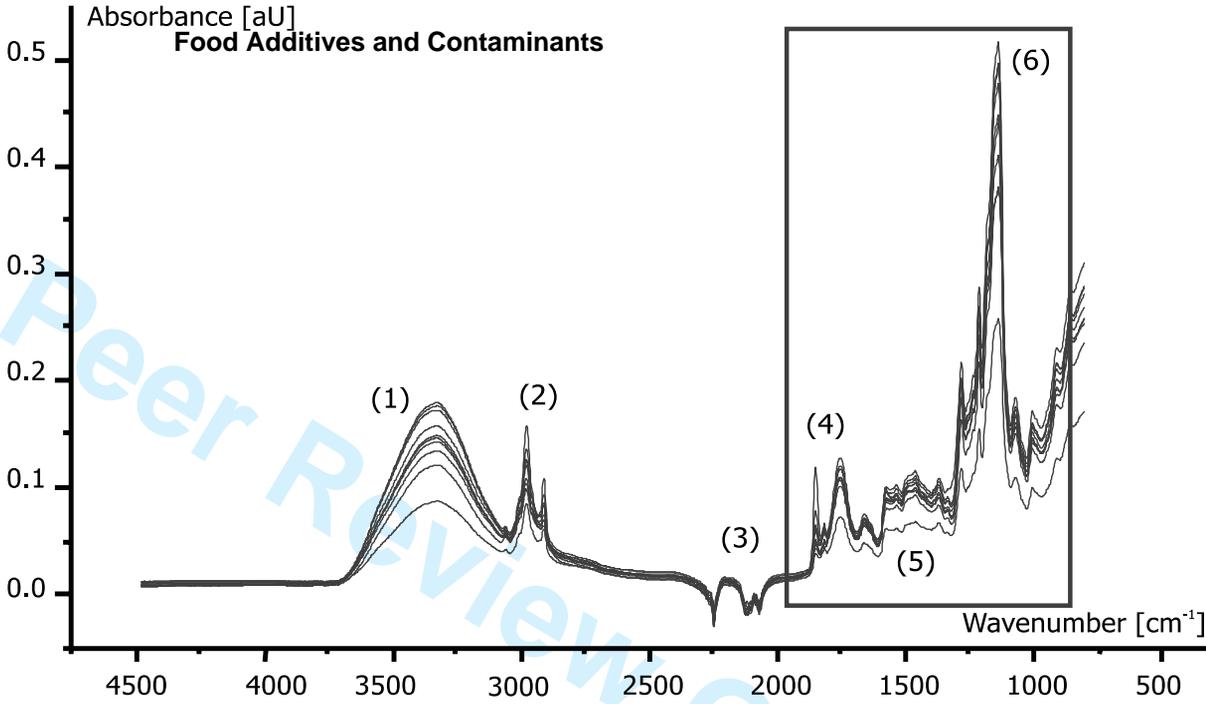
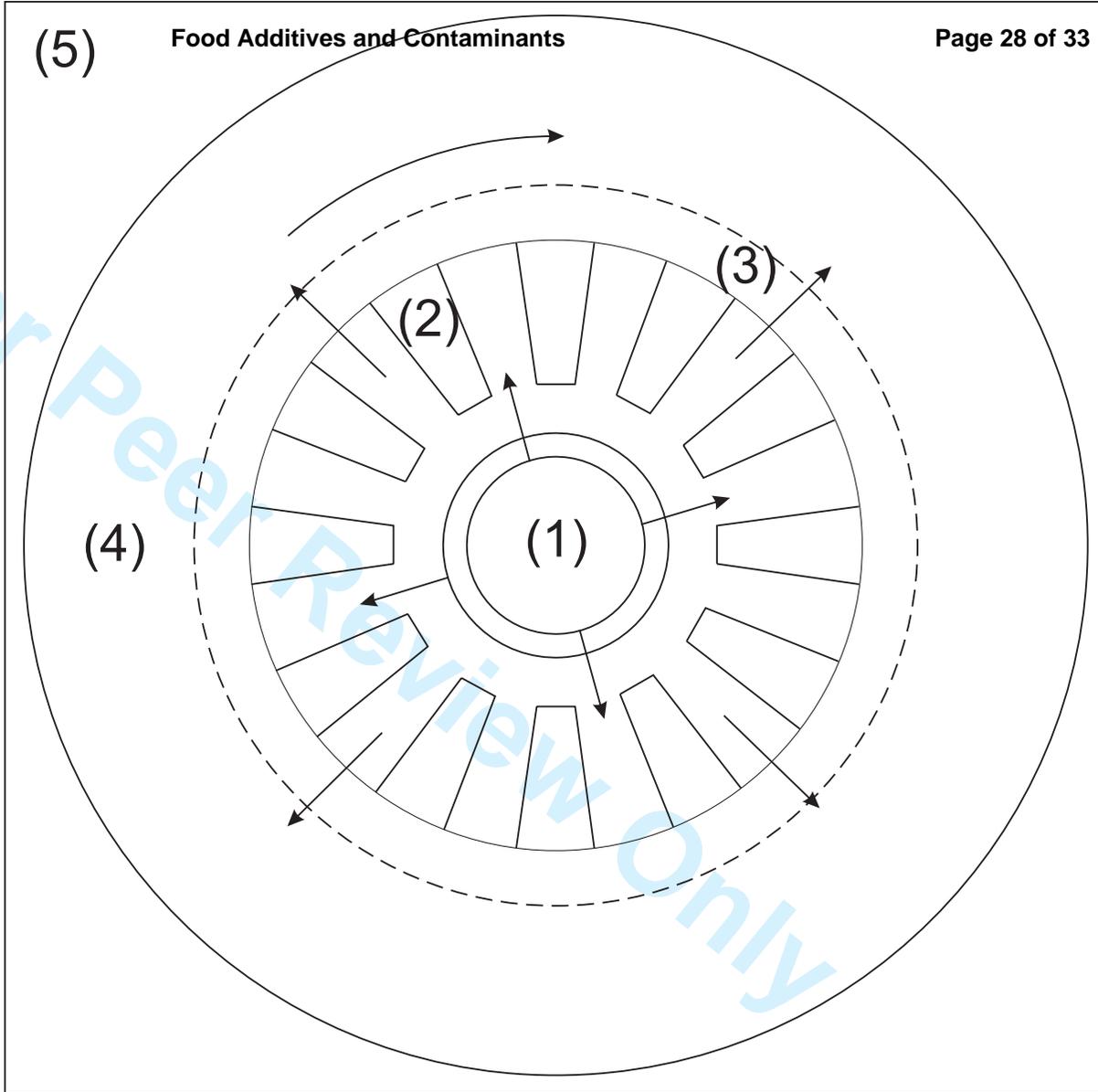


Figure 2



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Figure 1

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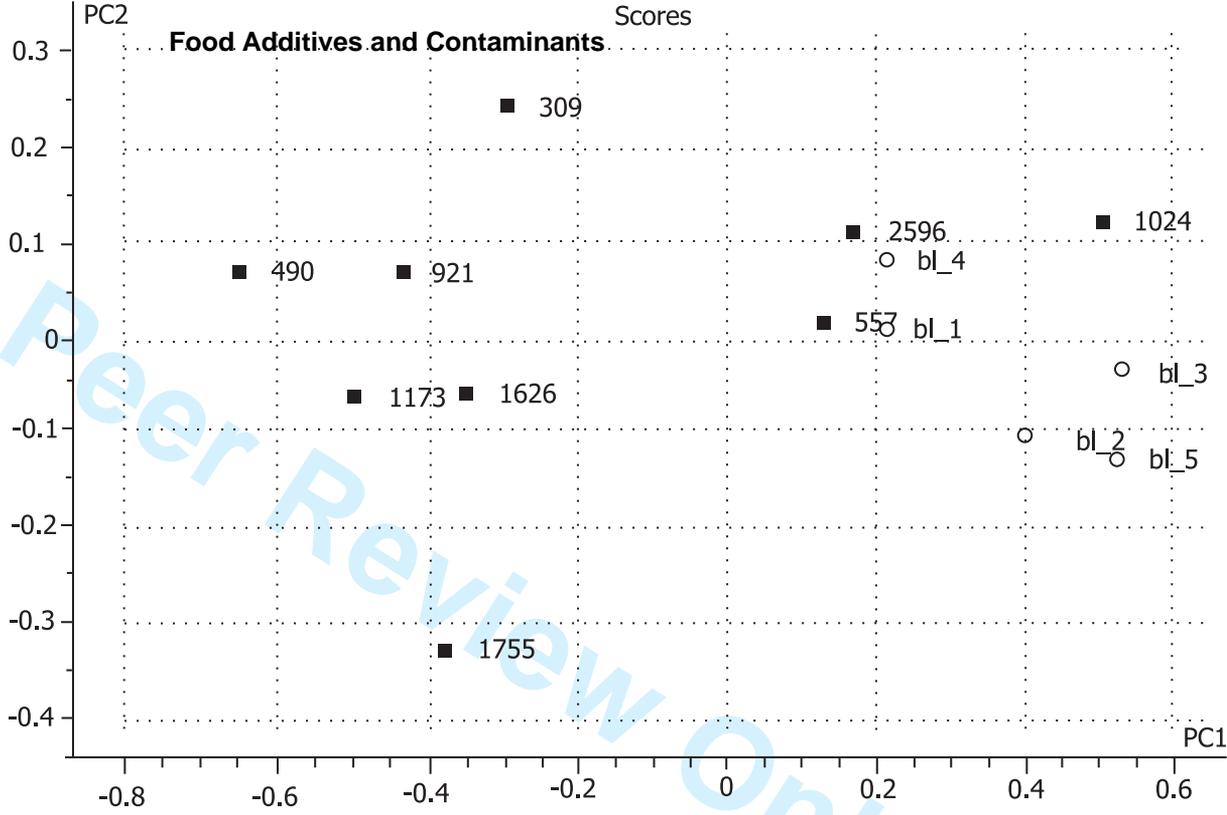


Figure 3

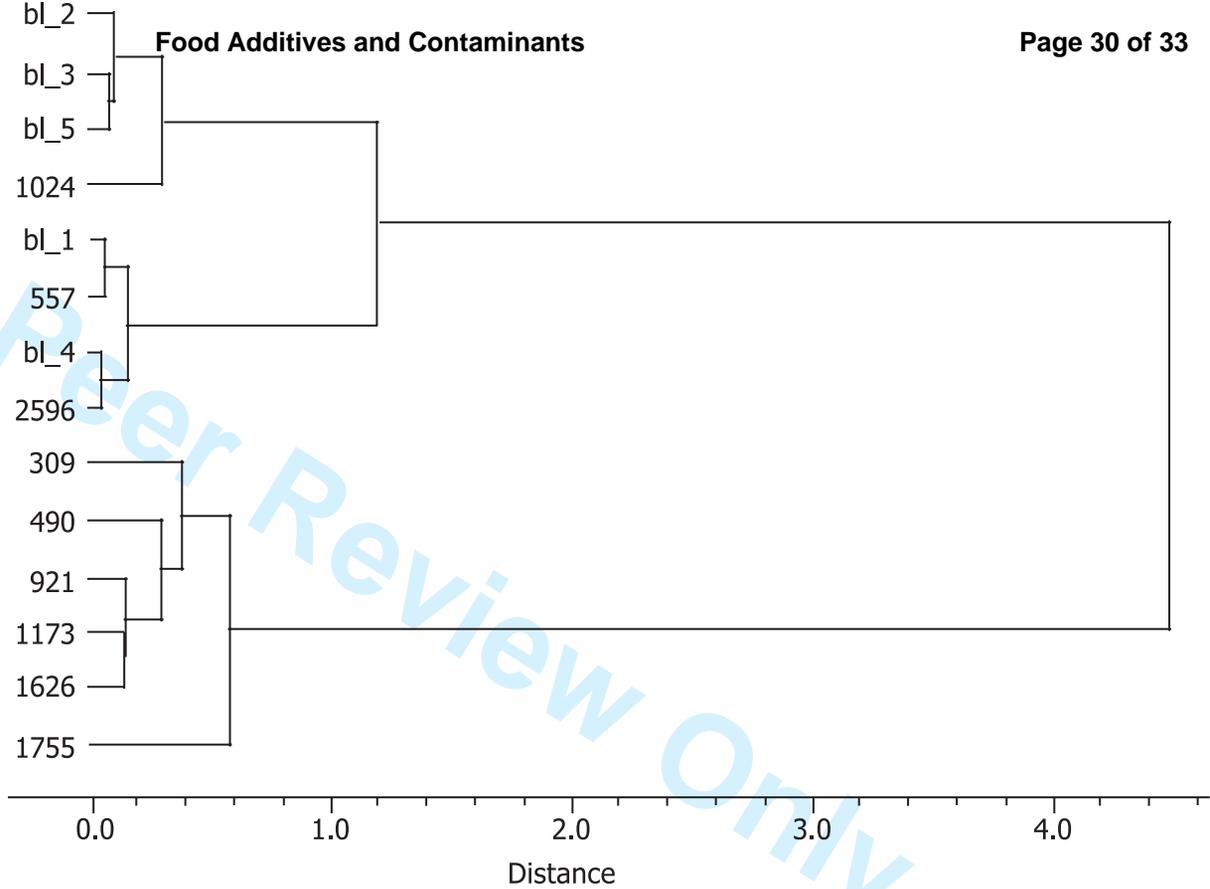


Figure 4

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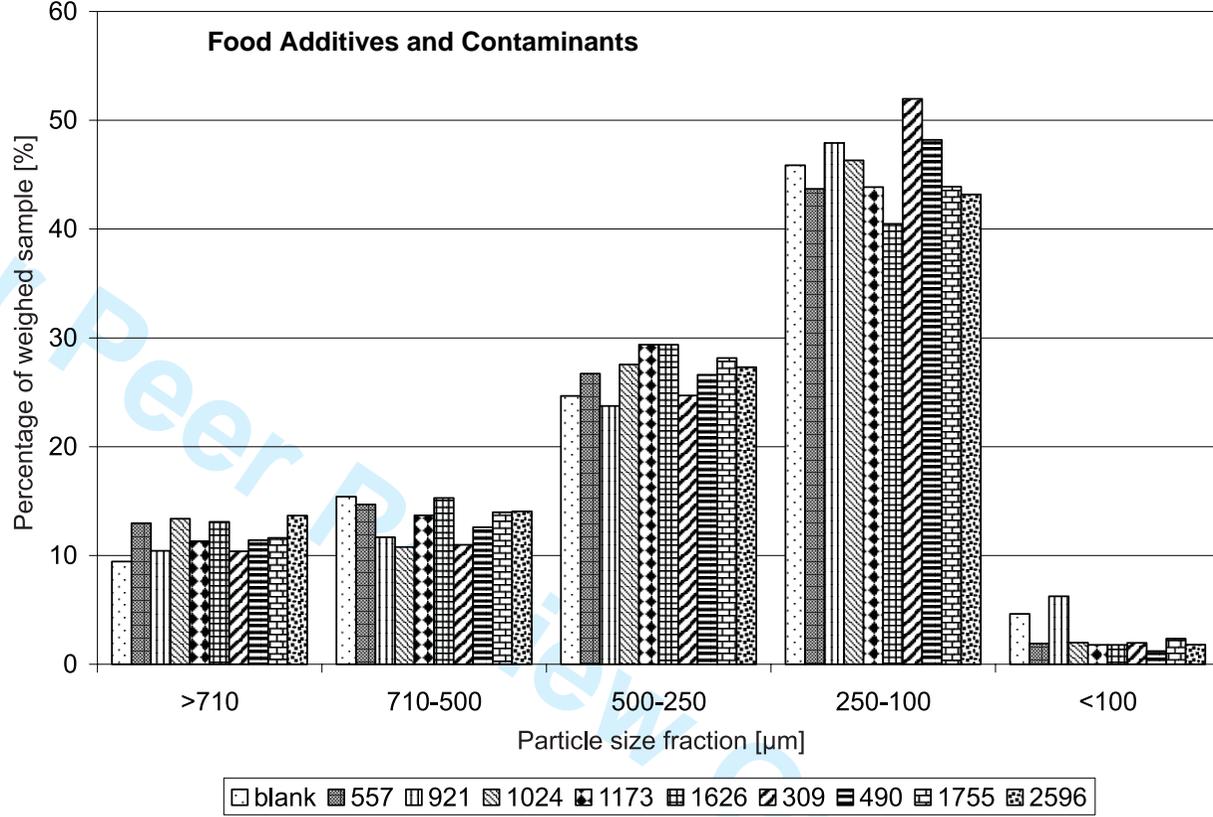


Figure 5

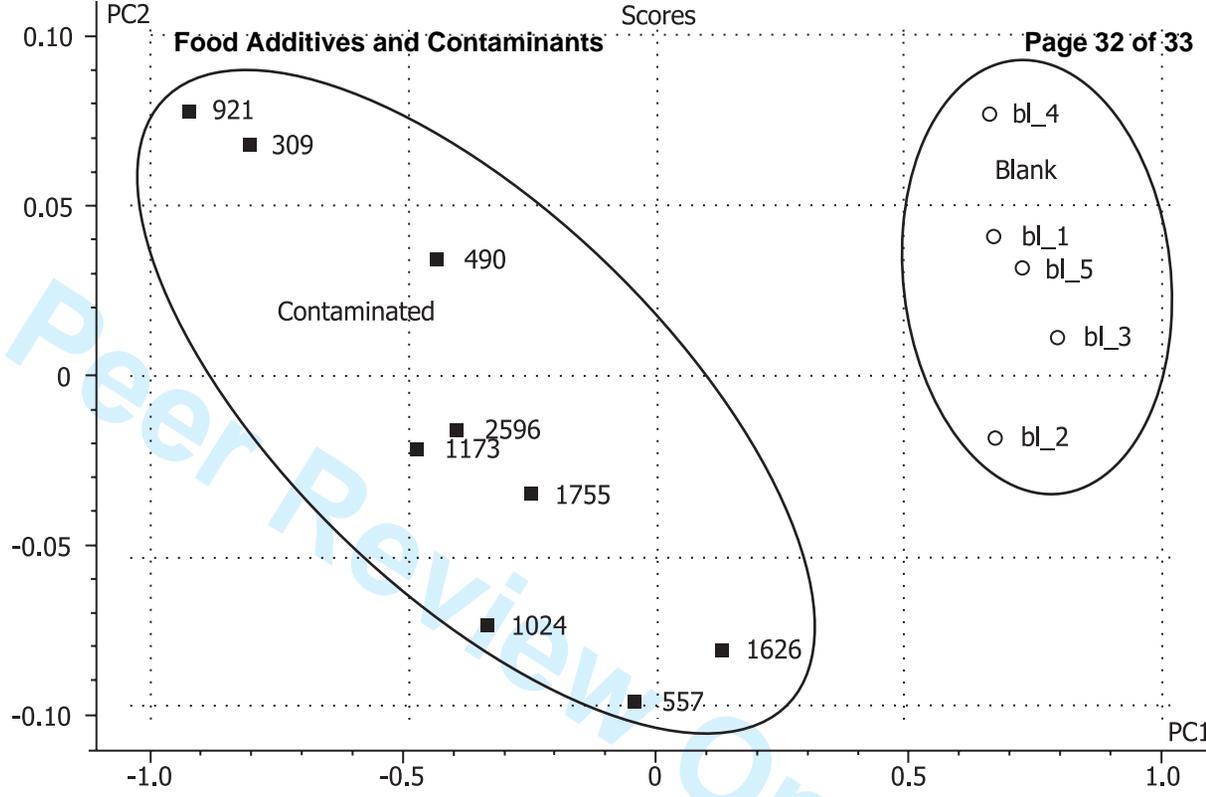


Figure 6

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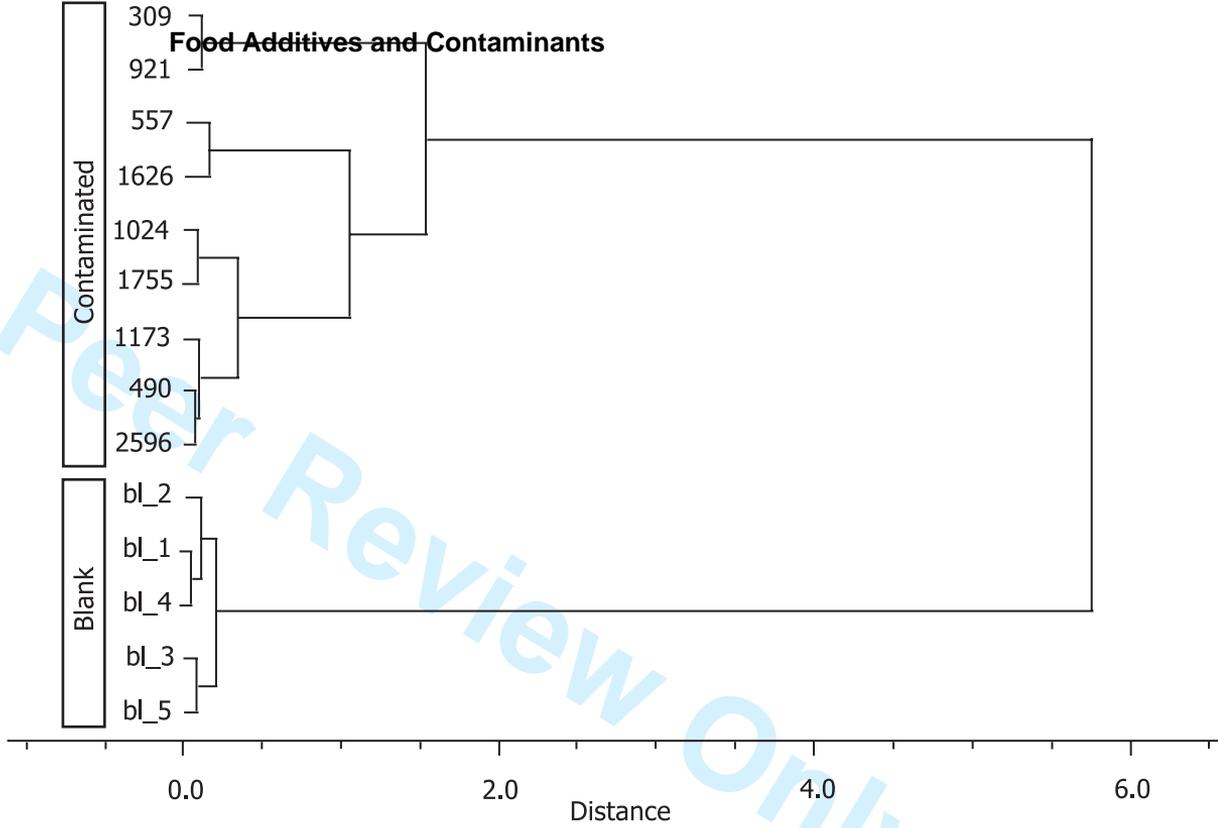


Figure 7