

1 **Feed authentication by near infrared microscopy**

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10 **Keywords**

11 Spectroscopy, microscopy, NIR, authentication, meat, discrimination, classification.

12

13 **Abstract**

14 In this study, the use of NIR spectromicroscopy for the detection and the quantification of
15 Meat and Bone Meal (MBM) in compound feedstuffs is investigated.

16 A spectral survey of particles from commercial raw materials was conducted to assess spectral
17 differences among forbidden and allowed raw materials for feeding ruminants. The validation
18 of a discrimination rule obtained with more than 1800 particles showed that it is possible to
19 recognize animal particles in a ground compound feedstuff with an error rate of 0.64 %.

20 A basic, non-adulterated compound feedstuff and different meat and bone meals thoroughly
21 mixed in different weight proportions (0 to 10% in 2% intervals) were used to calibrate the
22 instrument response. Reflectance spectra (1112 to 2500 nm) were acquired with a FT-NIR
23 microscope from particles randomly chosen in these feedstuffs. A linear regression model was
24 constructed between the proportion of meat and bone meal in feed and the area proportion of
25 the meat and bone particles detected by spectromicroscopy ($r^2 = 0.86$). The validation of this
26 model with an independent test set made of a few commercial or artificial compound

27 feedstuffs indicated that the spectromicroscopic method could be as reliable as the currently
28 adopted optical microscopic method.

29

30 **Introduction**

31 Bovine Spongiform Encephalopathy (BSE) is a fatal degenerative disease affecting the central
32 nervous system of cattle. According to the generally accepted scientific explanation, the BSE
33 epizootic in the United Kingdom has its roots in the recycling of contaminated cattle carcasses
34 processed into animal feed in the form of meat and bone meal (MBM), as well as in changes
35 made (in 1981-82) in the technological processes used in the production of such meal
36 (reduction of drying temperatures and discontinuation of solvent defatting in order to optimise
37 the extraction of fats).¹

38

39 The Commission Decision 94/381/EC of 27 June 1994 has banned, with effect from 27th July
40 1994 in all the Member States, the use of proteins derived from ruminant tissue or - in the
41 event of difficulty of identification - from any mammalian tissue for feeding ruminants.
42 Moreover, the EU laid down compulsory manufacturing standards in all the Member States in
43 order to improve the safety of meal for other animals (pigs, poultry, fish, etc...). These
44 standards have been tightened since 1 April 1997 (Decision 94/449/EC of 18/7/96: minimum
45 parameters for the processing of animal waste from mammals, excluding fats : $\emptyset < 50$ mm, t°
46 $> 133^{\circ}$ C, t: 20', p: 3 bar).

47

48 The ban of the use of mammalian protein in the feeding of ruminants need fast and reliable
49 analytical methods to identify animal ingredients in compound feed.

50

51 In most of the European countries the microscopic method is currently adopted. The detection
52 limit of the microscopic method is approximately 0.1% or even smaller. When used for

53 quantification of animal ingredients in feedstuffs this method is dependant on the presence of
54 bones in the product.² Moreover, the accuracy is very dependant on the bone content in the
55 animal ingredient to be identified in a compound feed. Furthermore, the differentiation of
56 bones from mammals and poultry is very difficult and considerable expertise is necessary
57 to make this differentiation.

58
59 Contrary to the feedstuff microscopy the commercial Elisa can identify the different animal
60 species depending on the available antibodies. The detection limit of commercial Elisa used
61 for detection of constituents of animal origin in compound feedstuffs was at a level of
62 approximately 5% depending on the animal species. When increasing the temperature
63 treatment of the animal product the sensitivity of the detection decreases respectively.
64 Identification of products heated to above 130°C could not be achieved.^{3,4}

65
66 The DNA methodology is another approach for the identification of animal ingredients in
67 compound feedstuffs. By using PCR procedures and appropriate primer pairs the
68 methodology allows a rapid and sensitive detection of species specific DNA-sequences from
69 meat and bone meal. It allows detection of the presence of bovine derived meat and bone meal
70 in feedstuffs containing less than 0.125% meat and bone meal.⁵

71
72 NIRS is another possibility of identification of animal ingredients. The traditional application
73 of NIRS in the analysis of feeds has been focused on the development of predictive
74 calibration equations relating spectral data to chemical or nutritional parameters (e.g. crude
75 protein, crude fat, fibre fractions, starch, digestibility, energy, etc). In the particular case of
76 ingredients recognition in a mixture, NIRS has been used for a number of applications and
77 seems able to predict accurately the ingredient composition of binary mixtures.⁶ Further

78 research is needed for the quantitative prediction of meat and bone meal in compound
79 feedstuffs.⁷

80

81 In fact, presently, none of the methods described above is totally satisfactory to detect **and** to
82 quantify meat and bone meal in compound feed. We present hereafter a new method, based on
83 FT-NIR microscopy, to detect and to quantify meat and bone meal in compound feed. This
84 spectromicroscopic method consists in the analysis of several hundreds particles being the
85 result of the grinding of a compound feedstuff. These particles are then identify as
86 contaminant (meat and bone meal) particles or not by comparing their spectra with reference
87 libraries. Finally, the area proportion of meat and bone particles found is related to the meat
88 and bone meal percentage in the compound feedstuff.

89

90 **Materials and methods**

91 *NIR Perkin-Elmer microscope*

92 The AutoIMAGE Microscope is connected to a Perkin-Elmer FT-NIR and allows to collect
93 spectra from extremely small samples (up to $5\mu \times 5\mu$). The microscope includes a camera and
94 a viewing system that magnifies the visible-light image of the sample to observe, to position
95 (by means of a motorized sample stage with a minimum step size of 1μ), and to isolate a point
96 of interest. The image of the sample is displayed on a PC monitor. The AutoIMAGE software
97 enables to control the operation of the microscope, to map and to collect spectra from a
98 sample. Spectra can be collected in reflectance or transmittance mode.

99

100

101

102

103 *Feedstuffs*

104 1) Raw materials

105 Raw materials samples used to construct reference libraries were supplied principally by the
106 Belgian Ministry of Small Enterprises, Traders and Agriculture as well as by two Belgian
107 feed producers.

108 The complete set of forbidden raw materials for feeding ruminants consisted of :

109 - meat and bone meal (MBM) (15 samples)

110 - meat meal (MM) (13 samples)

111 - ground bones (4 samples)

112 - feather meal* (3 samples)

113 - poultry by-products* (8 samples)

114 * feather meal and poultry by-products are not forbidden for feeding ruminants but we can not
115 actually differentiate these products from forbidden raw materials by NIR microscopy.

116

117 The complete set of allowed raw materials for feeding ruminants consisted of :

118 - fishmeal (19 samples)

119 - peas (6 samples)

120 - manioc (6 samples)

121 - wheat (2 samples)

122 - blood meal (1 sample)

123 - rape extracted oil cake (3 samples)

124 - corn (3 samples)

125 - maize gluten feed (1 sample)

126 - maize germ oilcake (1 sample)

127 - soybean (5 samples)

- 128 - flax (3 samples)
129 - lucerne (alfalfa) (4 samples)
130 - milk by-product (2 samples)

131

132 2) Compound feedstuffs

133 Compound feedstuffs with known concentration in MBM were used to build and to validate
134 the model.

135 The training set consists in a basic, non-adulterated compound feedstuff composed by a
136 Belgian feedstuff producer (1A, see Table 1 for composition) and different MBM thoroughly
137 mixed in different weight proportions (0 to 10% in 2% intervals).

138 The test set consists in three compound feedstuffs. The first one was prepared by the State
139 Analysis Laboratory Tervuren and the two others were prepared by a Belgian feedstuff
140 manufacturer.

141

142 *Sample preparation*

143 Samples were ground with a 1mm hole sieve (Retsch mill, Germany).

144

145 *Sample scanning and spectra acquisition*

146 Analysis were made on particles displayed on a reference surface (spectralon) in reflectance
147 mode with an aperture size of 50 μ by 50 μ . Reflectance data as Log 1/R were recorded at 4
148 nm intervals over the region 1112 to 2500 nm giving 348 data points per spectrum. Spectra
149 were averaged from 100 scans for the libraries construction and from 10 scans for the
150 compound feedstuffs analysis.

151

152

153

154 *Data treatment*

155 Spectral data were processed using ISI software (NIRS 3 ver. 4.0 and WinISI, Infracsoft
156 International, Port Matilda, PA, USA) and SAS software, ver. 6.12 (SAS Institute Inc., Cary,
157 NC, USA). Images of particles were processed using Micro Image 3.0 (Olympus Optical Co.,
158 Hamburg, Europe).

159 Canonical discriminant analysis was used to derive canonical variates that summarize
160 between-group variation in much the same way that principal components summarize total
161 variation. The first canonical axis was used to visualise differences between groups.

162 Because each group (allowed particles and forbidden particles) is a mixture of different
163 populations (different raw materials) it was difficult to assume multi-normal distributions.

164 This was confirm by the results of a normality test done on raw spectra for each variable: for
165 the first group (allowed particles), the normality hypothesis were rejected at level 0.05 for all
166 variable, and for the second group (forbidden particles), the normality hypothesis were
167 rejected at level 0.05 for 291 variables. Therefore, we decided to use a non-parametric method
168 to discriminate between groups. An artificial neural network (multilayer perceptron network
169 with back propropagation based on the partial least squares scores) was used to discriminate
170 between groups encoded as -1 for allowed particles and 1 for forbidden particles. Predicted
171 values below 0 were assigned to the first group and values above 0 were assigned to the
172 second group. Previously, data were processed using standard normal variate and detrend
173 SNVD along with a first derivative math treatment 1,4,4,1.

174

175 **Results**

176 *Spectral features*

177 The spectral features of the mean spectra of particles of the most characteristic raw materials
178 are shown in Figure 1. Characteristic bands of water are observable at 1452 nm (OH first

179 overtone) for plant raw materials and 1940 nm for all spectra. Soya and animal raw materials
180 show bands at 2312 and 2356 nm (CH combinations) due to fat and bands at 2064 and 2184
181 nm (NH combinations) due to protein whereas corn and wheat show a band at 2100 nm (OH
182 combinations) due to starch.

183 There is high similarities between spectra of MBM, poultry, feather and fish which are not
184 easily visually differentiable and which highlight the need to use chemometrics to distinguish
185 between them.

186 *Qualitative analysis*

187 A canonical discriminant analysis conducted on the calibration set allows the separation
188 between the two groups (allowed and forbidden raw materials) (Figure 2).

189 A predictive discriminant analysis, using an artificial neural network (ANN), is used to
190 classify particles into two groups on the basis of their absorbances from 1112 nm to 2500 nm,
191 the first group is made of forbidden particles and the second group gathers allowed particles.

192 A prediction rule is established with particles for which we know the group of origin
193 (calibration or construction set). More than 3000 particles were analysed to construct and to
194 validate the discrimination rule.

195 Table 2 shows the repartition of the particles analysed during the construction and the
196 validation of the discrimination rule and Table 3 shows the results of the validation of this
197 rule. Results of the ANN are shown in Figure 3 where the output of the ANN for each sample
198 in the test set has been plotted against its arbitrary sample number.

199

200 The overall error rate, estimated with the independent validation set of particles, is given by :

$$201 \quad \textbf{Overall error rate} = (0.63 + 0.66)/2 = \textbf{0.64\%}.$$

202

203

204 *Quantitative analysis*

205 1) Calibration step

206 To estimate the proportion of meat and bone meal in feed, compound feedstuffs with known
207 concentrations in meat and bone meals were used to construct the calibration equation. The
208 training set consist in a basic, non-adulterated compound feedstuff and different meat and
209 bone meals thoroughly mixed in different weight proportions (0 to 10% in 2% intervals).
210 Image analysis was used to measure the area proportion of the meat and bone particles in the
211 compound feed. Results are given in Table 4 and Figure 4 shows the relation between the
212 proportion of meat and bone meal in feed and the area proportion of the meat and bone
213 particles.

214

215 2) Validation step

216 The validation step consists in the analysis of four independent compound feedstuffs which
217 range from 2 to 6% MBM. Results are given in Table 5.

218

219 **Conclusions**

220 *Qualitative analysis*

221 Results of the discriminant analysis between particles of raw materials allowed or forbidden
222 for feeding ruminants indicate that it seems possible to detect, with NIR microscopy, MBM
223 particles in a compound feedstuff with a success rate greater than 99%. These good results
224 must be tempered for two reasons.

225 First, presently, it is difficult to differentiate between bovine meat meal particles and feather
226 or poultry meat meal particles.

227 Secondly, if the MBM proportion in a compound feedstuff is low, we need to analyse a large
228 set of particles if we want to observe at least one MBM particle with a high probability. For

229 example, if there is 2% MBM in a compound feedstuff and if we want to observe at least one
230 MBM particle with a probability of 95%, about 250 particles should be analysed. If there is
231 0.5% MBM in a compound feedstuff, which is the maximum acceptable level according to the
232 opinion of EC Scientific Steering Committee ⁸, and if we want to observe at least one MBM
233 particle with a probability of 95%, about 1000 particles should be analysed.

234

235 *Quantitative analysis*

236 The results of the analysis of compound feedstuffs with known concentrations in MBM
237 (Table 5), even if their are not sufficient to allow definitive conclusions, are promising. The
238 accuracy obtained, is not sufficient to allow a quantitative control of compound feedstuffs.
239 However, it seems certainly as good as the currently adopted microscopic method. Further
240 analysis are required to improve this accuracy.

241

242 *Perspectives*

243 The differentiation between bovine meat meal particles and feather or poultry meat meal
244 particles could be achieved by using a larger spectral range (780-2500 nm or 400-2500 nm).^{6,9}
245 The quantitative analysis is performed by a regression model related the proportion of meat
246 and bone meal in feed and the area proportion of the meat and bone particles. This model
247 could be improved by discriminating between meat particles and bones particles. The
248 discrimination between meat and bone particles could take into account density differences
249 between these particles and therefore get a better quantitative model.

250

251 **Acknowledgement**

252 This project was financed by the Ministry of Agriculture, DG4.

253

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281 Table 1
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283 Feed authentication by near infrared microscopy
284

285 Table 1. Composition of the
286 basic, non-adulterated, compound
287 feedstuff 1A.

Feedstuffs
Palmist
Wheat
Flax
Soy bean
Citrus
Coconut
Glutenfeed
Sugar beet roots
Bran
Minerals
Vitamins
Binder
Molasses

288

289 Table 2
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291 Feed authentication by near infrared microscopy

292
293 Table 2. Repartition of the particles analysed during the construction and the validation of the
294 discrimination rule.

	Number of particles (number of samples)		Total
	Forbidden particles	Allowed particles whose fish	
Construction of the discrimination rule	379 (13)	780- 210 (26-7)	1159
Validation of the discrimination rule	912 (31)	960- 360 (32-12)	1872
Total	1291 (44)	1740 (58)	3031

295

296 Table 3
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300

301 Table 3. Validation of the discrimination rule.

	Number of particles classified into each group (percentage)		Total
	"allowed"	"forbidden"	
Particles from the group "allowed"	954 (99.37%)	6 (0.63%)	960
Particles from the group "forbidden"	6 (0.66%)	906 (99.34%)	912

302

303 Table 4
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306
 307

308 Table 4. Quantitative analysis. Results of the calibration step.

309

Sample number	% MBM (w/w) in sample	Particles analysed		Particles identified as MBM			
		nb	area	nb	% nb	area	%area
0a	0	601		0	0	0	0
0b	0	600		0	0	0	0
2c	2	667	549327	12	1.80	9383	1.71
2d	2	627	599213	8	1.28	8102	1.35
4c	4	599	626755	13	2.17	11980	1.91
4d	4	604	649961	18	2.98	14011	2.16
6c	6	620	585088	19	3.06	22535	3.85
6d	6	634	549133	26	4.10	23081	4.20
8c	8	617	631854	36	5.83	36003	5.70
8d	8	618	542522	20	3.24	14764	2.72
10c	10	623	530410	45	7.22	36396	6.86

310

311 Table 5
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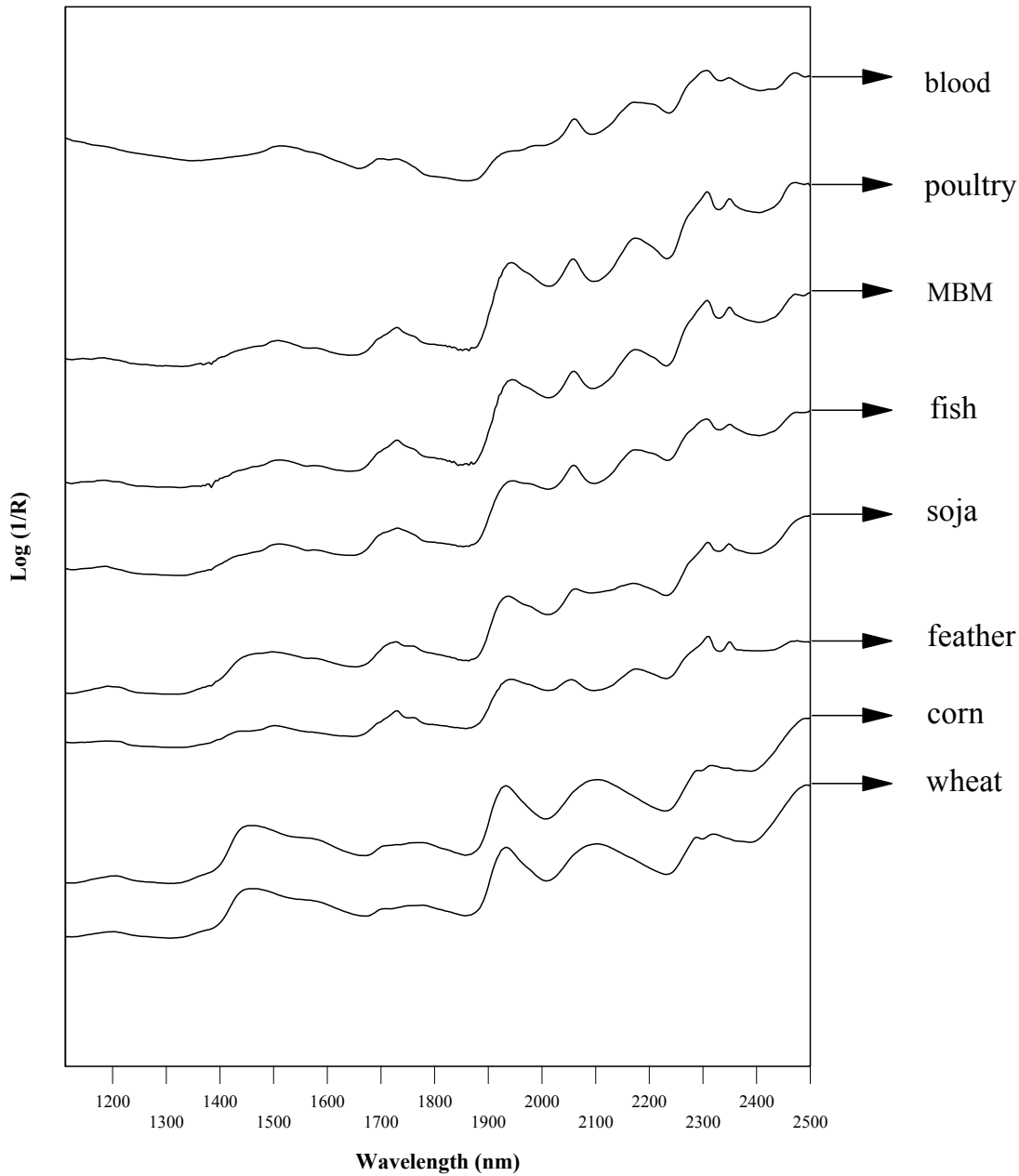
316 Table 5. Quantitative analysis. Results of the validation step.

317

Sample number	Sample description	% mbm (weighth)	% mbm estimated by NIR microscopy
2a	pasture supplement	2	2.02
3a	sow feed	6	4.57
4a	pheasant feed	6	3.36

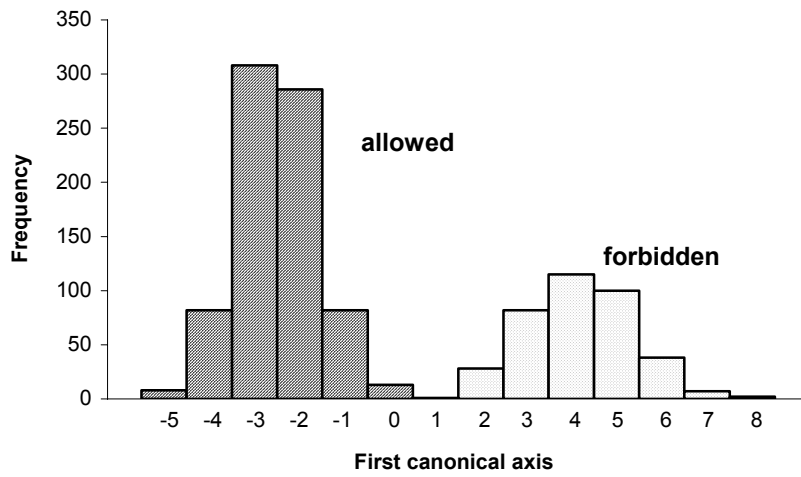
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319 Figure 1
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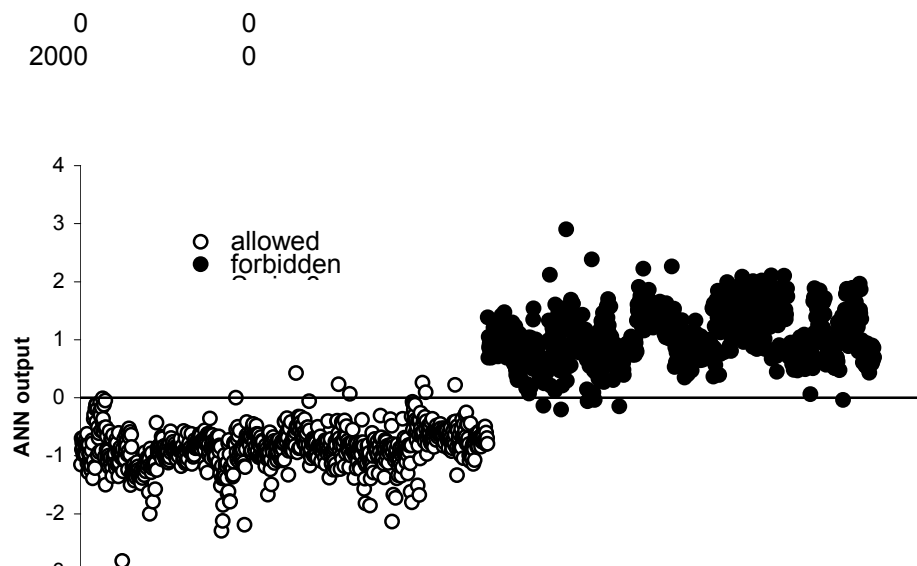
323 Figure 1. Mean spectra of particles of characteristic raw materials.

324 Figure 2
325 F. Piraux and P. Dardenne
326 Feed authentication by near infrared microscopy
327



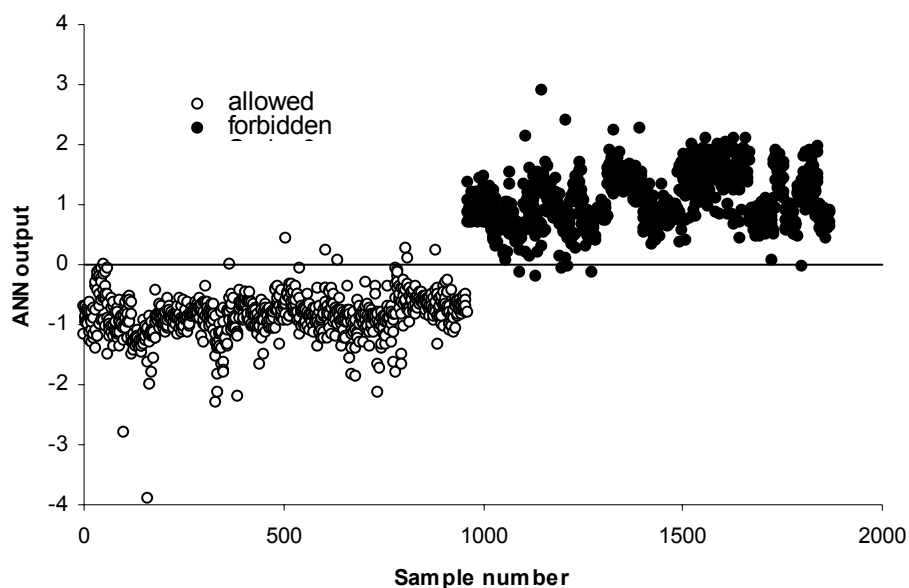
328 Figure 2. Separation between groups along first canonical axis for
329 calibration set.

330 Figure 3
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332 Feed authentication by near infrared microscopy
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335 Figure 3. classification of allowed and forbidden raw materials particles
336 using an artificial neural network for the validation set. The horizontal line
337 marks the threshold used to separate groups encoded as -1 for allowed
338 particles and 1 for forbidden particles.
339 Legend : O allowed, Δ forbidden.

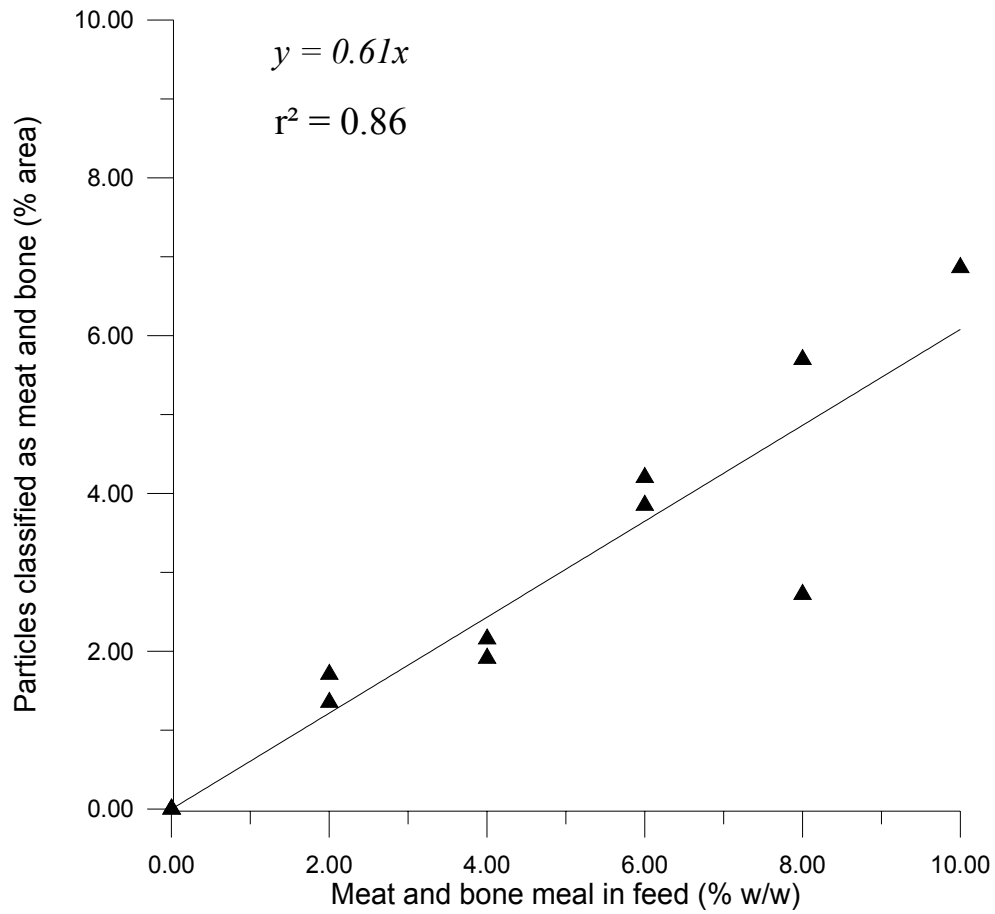
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346 Figure 4
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348 Feed authentication by near infrared microscopy

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353
354 Figure 4. Proportion of meat and bone meal in feed estimated by the area
355 proportion of meat and bone particles.

356