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1 **An innovative approach to predict the growth in intensive poultry farming**

2

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10 **Abstract**

11 Chicken weight provides information about growth and feed conversion of the flock in order to identify
12 deviations from the expected homogeneous growth trend of the birds. This paper proposes a novel method
13 to automatically measure the growth rate of broiler chickens by sound analysis.

14 Through the application of process engineering, Precision Livestock Farming (PLF) can combine audio and
15 video information into on-line automated tools that can be used to control, monitor and model the
16 behaviour, health and production of animals and their biological response.

17 The aim of this study was to record and analyse broiler vocalisations under normal farm conditions, to
18 identify the relation between animal sounds and growth trend. Recordings were made at regular intervals,
19 during the entire life of birds, in order to evaluate the variation of frequency and bandwidth of the sounds
20 emitted by the animals.

21 Two experimental trials were carried out in an indoor reared broiler farm; the audio recording procedures
22 lasted for 38 days. The recordings were made, in an automated, non-invasive and non-intrusive way and
23 without disturbing the animals in to the broiler house. Once a week, 50 birds were selected at random and
24 their weight recorded in order to follow the growth trend in the birds.

25 Sound recordings were manually analysed and labelled using the Adobe Audition CS6 software.

26 Analysing the sounds recorded, it was possible to find a significant correlation ($P < 0.001$) between the
27 frequencies of the vocalisations recorded and the weight of the broilers.

28 The results explained how the frequency of the sounds emitted by the animals was inversely proportional
29 to the age and to the weight of the broilers; the more they grow, the lower the frequency of the sounds
30 emitted by the animals.

31 This preliminary study shows how this method based on the identification of specific frequencies of the
32 sounds, in an indoor reared broiler farm, linked to the age and to the weight of the birds, could be used as
33 an early warning method/system to evaluate the health and welfare status of the animals at farm level,
34 developing also an automated growth monitoring tool.

35 **Keywords:** broiler, vocalization, PLF, grow trend, frequency analysis

36 **Introduction**

37 The demand for meat is rapidly growing all over the world (Tullo et al., 2013) and poultry is one of the
38 cheapest sources of animal protein. Currently, more than 40 billion chickens are produced every year by
39 specialised industries.

40 Broilers are the fastest-growing farmed species and their performance is influenced by adequate
41 environmental conditions such as ambient temperature, relative humidity, air and litter quality, and
42 ventilation speed. Thank to the progress in farming technologies, broiler chickens now mature at a higher
43 rate than in the past, have higher feed conversion efficiency, a reduced slaughter age and a higher final
44 weight (Rauw et al., 1998).

45 Chicken weight provides information about the growth and the feed conversion efficiency of the flock
46 useful to identify deviations from the expected homogeneous growth trend of the birds (Mollah et al.,
47 2010), having also details about the health and welfare status of the animals.

48 Since the animal health strongly depends on good welfare, during the last years many progresses have
49 been made in developing new indices/indexes and procedures to assess animal's health and welfare status.

50 Nevertheless, these monitoring procedures are time consuming and require trained manpower (Aydin et

51 al., 2014). For this reason, one possible way to make animal welfare assessment easier and faster could be
52 the application of audio and video data analysis. (Tefera, 2012) (Ferrari et al., 2013; Tullo et al., 2013).
53 Image analysis, in particular, was successfully used to estimate the body weight of the animals (Mollah et
54 al., 2010) while audio analysis have been widely used to better identify specific behaviours and vocalisation
55 patterns in different animals' species (Chan et al., 2011; Vandermeulen et al., 2013).
56 Animals use vocalisation to express different inner states provoked either by internal or external events,
57 and also to reveal some of their behavioural needs (Aydin et al., 2014). For instance, chicken broiler
58 vocalisations have been studied (Marx et al., 2001) to better understand the vocal pattern of this species in
59 relation to environmental temperatures and stress situations (e.g. high/low temperatures). Moreover,
60 information technologies have been used to monitor feed intake, body weight and growth trend (Aydin et
61 al., 2014).
62 The non-invasive nature of the audio and video equipment allows its use in long term monitoring of
63 animals, without disturbing them (Aydin et al., 2013).
64 The combination of audio and video information into automated tools could be used in early warning
65 systems to detect health or welfare problems (Precision Livestock Farming-PLF) (Costa et al., 2013). One of
66 the objectives of PLF is to develop on-line tools for monitoring farm animals continuously and automatically
67 (Viazzi et al., 2011) during their life without imposing additional stress. The PLF approach can be applied to
68 different aspects of management, with a focus on the animals and/or on the environment, and at different
69 scales, from the individual to the entire flock/herd (Wathes, 2009). Moreover, PLF may also be used to aid
70 the management of some complex biological production processes, to measure the growth rate and to
71 monitor the animal activity (Halachmi et al., 2002; Ismayilova et al., 2013; Tullo et al., 2013).
72 The aim of this study was to record and analyse broiler vocalisations under normal farm conditions, to
73 identify the relation between animal sounds and growth trends. The relation between Peak Frequency (PF)
74 of sounds emitted by broiler chickens during the production cycle and their weights (both measured with
75 an automated and a manual scale) were investigated. This study proves that audio and video data

76 monitoring is a promising technique for the development of an automated growth-monitoring tool for the
77 farming of broiler chickens.

78 **Material and methods**

79 Two experimental trials were carried out in an indoor reared broiler farm; the first one took place in June
80 and July 2013 and the second one in August and September 2013.

81 The farm where the experimental trials took place was an indoor broiler farm rearing birds to the RTFA
82 (ACP) standard. The house dimensions were 61m x 21m and the total floor area available to the birds was
83 1,130m². Inside the house there were 2,340 nipples drinkers, and 385 feed pans available to birds. 27,940
84 day old chicks were placed inside the house at day 1 in both trials.

85 Sound recordings were collected using a professional handheld solid state recorder (Marantz PMD 661 MK
86 II) which was connected to two different directional microphones placed at an intermediate height of
87 between 0.4m and 0.8m (depending on the height of the animals in order to keep the same distance
88 among animals and microphones during the entire data-collecting procedure).

89 The supercardioid/lobe microphone (Mic. 1) was a Sennheiser K6 / ME66" (frequency response: 40-
90 20,000Hz ± 2,5 dB) and it was held by a short tripod microphone stand (Quiklok A341) above the feeder.

91 The (cardioid) microphone (Mic. 2) was a Sennheiser K6 / ME64" (frequency response: 40-20,000Hz ± 2,5
92 dB) and it was placed on a long tripod (Quiklok A492 Heavy-Duty Boom Mic Stand) directly above the
93 drinkers.

94 Both the microphones were slightly inclined toward the floor in order to capture preferentially the sounds
95 coming from the birds walking exactly in front of the microphone axis.

96 The recordings provided a sound image of background noise, and gave a better idea of the overall condition
97 inside the broiler house.

98

99 The Marantz PMD 661 MK II recording machine had a large range of potential recording settings. The
100 settings found to give the most sensitivity to bird sounds in the poultry house environment were:

101 Rec. Format: PCM-16, Stereo Sample Rate: 44.1k

102 Level Control.: Manual Low Cut: Off High Cut: Off

103

104 Animal sounds were recorded for 1 continuous hour using 2 different microphones during each
105 experimental session from day 1 to day 38. Recordings were made at regular intervals every Monday,
106 Wednesday and Friday, with the same position of the equipment along the trial procedures.

107 Once a week, 50 birds were selected at random and their weight recorded in order to follow the growth
108 changes in the birds. Throughout the production period from day 1 to day 38 house temperature and
109 humidity levels were recorded.

110 The entire data collection consisted in 16 days of sound recordings for trial 1, 15 days of sound recordings
111 for trial 2, and 6 weekly weight collections for both trials.

112 In total 55 h 20 min of recordings were collected and 600 birds were weighted during trial 1 and trial 2; only
113 the audio files recorded in conjunction with the weight collection of the birds were included in the data
114 analysis.

115 In total 600 sounds (50 sounds per day), chosen at random and selected from 12 days of recordings were
116 manually labelled and analysed in this study.

117 **Sound analysis**

118 Sound recordings were manually analysed and labelled using sound analysis software: Adobe® Audition™
119 CS6. Every hour-long duration recorded digital file was cut into shorter files of 10 minutes each in order to
120 simplify the sound analysis.

121 Sound labelling involved the extraction and classification of both individual animal sounds and general
122 sounds coming from the whole flock on the basis of the amplitude and frequency of the sound signal in
123 audio files recorded at farm level (Tullo et al., 2013).

124 Labelling is a manual procedure based on acoustic analysis combined with visual spectral analysis, which is
125 used to extract fragments of sounds from the entire recording. The labelling procedure was done offline by
126 extrapolating those sounds that the operator classified as significant vocalisation sounds *via* auditive
127 analysis and visual observation of the spectrogram (Ferrari et al., 2008).

128 Through Adobe® Audition™ CS6 each sounds were identified and analysed using time (x-axis) and
129 frequency (y-axis).

130 The Fast Fourier Transform (FFT) was used to perform the frequency analyses using a Hamming window
131 with a FFT dimension of 256 sampled points (Figure 1).

132 For each sound the peak frequency (PF= representing the frequency of maximum power) was manually
133 extracted. The frequency range was band pass filtered between 1,000 Hz to 13,000 Hz. The lower frequency
134 limit was set at 1,000 Hz to remove the low frequency background noise and the upper limit was set at
135 13,000 Hz to cut off the high frequency noise and also because broilers are sensitive to a frequency range
136 of about 60 to 11,950 (Appleby et al., 1992; Tefera, 2012).

137 Figure 1.

138 **Statistical analysis**

139 Differences among PF extracted from the 600 sounds recorded in the two trials were tested with the PROC
140 TTEST of SAS 9.3. A paired t-test was performed to compare PF of sounds recorded at different ages of
141 birds within the same trial. The relation between weight and PF of sounds recorded at different ages was
142 also investigated with PROC CORR in SAS 9.3. The PROC REG. was used to predict variation in the PF
143 according to the change of age of the birds (in weeks) with the following model:

$$144 \quad \text{PF} = \text{week}$$

145 The estimation of effects influencing the PF was performed with the GLM procedure in SAS 9.3. The model
146 used was the following:

$$147 \quad \text{PF} = \text{weight} * \text{age}$$

148 Table 1.

149 The fixed effect (weight*age) was divided in 12 classes, as the result of the interaction (pairing) of the age
150 with the average weight of the birds (Table 1). The division in classes allowed avoiding the nesting effect.

151 **Results and discussion**

152 Table 2.

153 For each sound the frequency analysis was carried out, in order to extract the peak frequency of each
154 vocalisation. The mean weights collected during both trials agree with the growth trend of this breed found
155 in literature (Aviagen, 2012).

156 Table 3 shows the means and standard deviations of the peak frequency (PF) of sounds recorded in trial 1
157 and trial 2.

158 The comparison shown in Figure 2 shows how there is no difference (P value= 0.4508) between PF means
159 of the sounds recorded in the two trials.

160 Figure 2.

161 Furthermore, the comparison between PF of sounds collected on the same week of age of birds during the
162 experimental trials (Figure 3) confirmed that the two trials could be considered as the equivalent. This
163 could be related to the use in poultry farming of fast-growing hybrid broilers with typical and homogeneous
164 growth rate across production cycles.

165 Indeed all the P values reported in Fig 3 reveal the non-significant difference between PF means of the
166 sounds emitted by the animals during specific days of both trials.

167 Figure 3.

168 In Table 3 the paired T-test between days of the same trial were tested to verify the difference between the
169 PF means of the vocalisation during the life of the broiler chickens; the difference is resulted significant in
170 both trials

171 As it is possible to see in Table 4 and Table 5 and in Figure 3 each age is characterised by its own typical
172 peak frequency that decreases with the growth of the birds.

173 Considering the difference between week 1 and week 6 it is possible to see how the peak frequency
174 decreases of about 2,000 Hz.

175 In both trials the average frequency reduction was around 350 Hz per week.

176 Furthermore analysing the PF related to the weight of birds, it was possible to confirm a significant negative
177 correlation (-0.80; $P < 0.0001$) between the frequencies of the vocalisations recorded and the weight of the
178 broilers, during the different experimental trials.

179 Table 3.

180 As it is shown in Figure 4 the peak frequency of the vocalisations of the broiler chickens is strictly
181 dependent on the age and on the weight of birds.

182 The regression model is significant ($F=251.52$, $P < 0.0001$), indicating that the model accounts for a
183 significant portion of variation in the data. The R^2 indicates that the model accounts for 98% of the variation
184 in peak frequency.

185 The confidence interval (CI_obs_95) of the observed values shows a 95% probability that the true linear
186 regression line of the population will lie within the confidence interval of the regression line calculated
187 from the sample data.

188 The confidence interval (CI_exp_95) that includes the expected values of the regression model with a
189 probability of 95% (grey area in Fig 4) indicates the goodness of fit of the regression model.

190 Figure 4.

191 The results of the GLM were useful to verify the high impact of the weight and the age of the birds on the
192 PF of the vocalisation emitted by the animals during their life. In Figure 5 are reported the LSMEANS(\pm SEM)
193 of the PF of vocalisations according to the increase of the age and weight of the animals.

194 There is a decrease of peak frequency in vocalisations according to the age of the broiler chickens.

195 As reported by Marx et al. (2001) the PF of the vocalisation emitted by one week old chicks ranged from
196 3,000 to 4,000 Hz, reinforcing the results of the present study that very young chicks vocalise at high
197 frequency under non-stress condition.

198 Figure 5.

199

200

201 **Conclusion**

202 The results indicate that the peak frequency of the sounds emitted by the animals, is inversely proportional
203 to the age and the weight of the broilers; specifically the more they grew, the lower the frequency of the
204 sounds emitted by the animals.

205 Usually, nowadays, the weight of the birds is automatically collected by a single solid scale placed on the
206 floor of the house. The high numbers of animals inside the flock and the insufficient funds of scales make
207 impossible to collect the weight of all the birds. Manually measure the weight of a significant number of
208 animals requires manpower and deprives the farmer of useful time. Due to this, it should be useful to
209 automatically collect simultaneously information about the growth trend of all the birds inside the flock to
210 identify deviations from the expected homogeneous growth trend of the birds, having also details about
211 the health and welfare status of the animals.

212 This preliminary study shows that the methodological approach based on the identification of specific
213 sound frequencies emitted by the animals in an indoor reared broiler farm linked to their age and weight,
214 could be used as an early warning method/system or a continuous monitoring system to evaluate the
215 general status of the animals at farm level. Furthermore, this strict correlation between weight of the birds
216 and peak frequency of the sounds emitted by the animals could open the scenario to an automated tool
217 based on vocalisation to predict the weight and the growth trend of the birds. This allow the farmer to
218 automatically monitor the growth trend of the birds,

219 Of course further studies, in different farms, with daily data collection are necessary to improve the
220 knowledge on the relationship between vocalisation and weight of birds in order to create an accurate
221 weight prediction algorithm based on sounds emitted by the animals.

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225

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267

268 Figure Headings:

269 Figure 1. Screenshot of the Adobe® Audition™ software showing the spectrograms and the frequency
270 analysis window relative to a specific vocalisation. In the main window the time- frequency vocalisation
271 graph is shown, while the inset represents the frequency analysis.

272 Figure 2. Comparison between PF means of the sounds recorded in trial 1 and in trial 2.

273 Figure 3. Comparison between PF means of sounds emitted during days of the same week of age recorded
274 in different trials.

275 Figure 4. Linear regression of PF in relation to the age of the animals expressed in weeks. Confidence
276 intervals of the mean are reported in dotted lines. Confidence intervals of the prediction are represented
277 by the grey area.

278 Figure 5. LSMEANS(\pm SEM) of the peak frequency of vocalisation according to the increase of age and
279 weight. $P < .0001$

280

281 Tables:

282 Table 1. Description of the fixed effect Weight*age used in the GLM model. The 12 classes, are the result of
283 the interaction (pairing) of the age with the average weight of the birds.

284 Table 2. 50 Chicken broilers randomly chosen were weighted during their entire life, both in trial 1 and trial
285 2. Means and standard deviations (SD) of the peak frequency (PF) of the sounds recorded in both trials.

286 Table 3. Paired T-test between different days to verify the difference between the PF means of the
287 vocalisations during the entire life of the broiler chickens in trial 1.

288

289

290 Table 1.

Weight (g)	Age (d)	Weight*age	Weight (g)	Age (d)	Weight*age
40.72	1	1	1,039.46	22	7
44.56	1	2	1,092.84	23	8
198.64	8	3	1,529.00	29	9
231.42	9	4	1,731.60	30	10
550.30	15	5	2,104.28	36	11
608.66	16	6	2,275.44	37	12

291

292

293

294 Table 2.

Week	Trial	Day	Mean weights (g) \pm SD	Mean PF (Hz) \pm SD
1	1	1	44.56 \pm 1.5	3,545 \pm 365
2	1	8	198.64 \pm 10.1	3,059 \pm 459
3	1	15	550.3 \pm 21.7	2,618 \pm 360
4	1	22	1,039.5 \pm 68.6	2,329 \pm 605
5	1	29	1,529 \pm 120.5	1,943 \pm 569
6	1	36	2,104.28 \pm 208.5	1,506 \pm 434
1	2	1	40.72 \pm 4.9	3,621 \pm 402
2	2	8	231.42 \pm 1.1	2,953 \pm 353
3	2	15	608.66 \pm 26.7	2,474 \pm 384
4	2	22	1,092.84 \pm 74.4	1,955 \pm 520
5	2	29	1,731.6 \pm 130.3	1,902 \pm 585
6	2	36	2,275.44 \pm 247.0	1,475 \pm 493

295

296 Table 3.

Trial 1			Trial 2		
Comparison	Difference Mean (SEM)	P-value	Comparison	Difference Mean (SEM)	P-value
Day 1 – Day 8	485.8 (76.7)	***	Day 1 – Day 9	668.4 (73.4)	***
Day 1 – Day 15	926.8 (66.9)	***	Day 1 – Day 16	1,174.3 (87.69)	***
Day 1 – Day 22	1,216.2 (103.8)	***	Day 1 – Day 23	1,674.1 (121.4)	***
Day 1 – Day 29	1,602.1 (93.3)	***	Day 1 – Day 30	1,740.3 (120.7)	***
Day 1 – Day 36	2,039.6 (94.3)	***	Day 1 – Day 37	2,146.4 (80.8)	***
Day 8 – Day 15	441.0 (72.2)	***	Day 9 – Day 16	478.9 (79.4)	***
Day 8 – Day 22	730.4 (106.8)	***	Day 9 – Day 23	949.7 (96.6)	***
Day 8 – Day 29	1,116.3 (108.4)	***	Day 9 – Day 30	1,015.9 (109.0)	***
Day 8 – Day 36	1,553.8 (85.5)	***	Day 9 – Day 37	1,478.0 (80.6)	***
Day 15 – Day 22	289.4 (91.5)	***	Day 16 – Day 23	485.9 (102.2)	***
Day 15 – Day 29	675.3 (100.7)	***	Day 16 – Day 30	552.1 (107.2)	***
Day 15 – Day 36	1,112.8 (81.8)	***	Day 16 – Day 37	999.1 (97.1)	***
Day 22 – Day 29	385.9 (124.8)	**	Day 23 – Day 30	366.3 (136.4)	*
Day 22 – Day 36	823.4 (101.5)	***	Day 23 – Day 37	428.5 (137.0)	**
Day 29 – Day 36	437.6 (101.7)	***	Day 30 – Day 37	362.2 (130.6)	**

297