

1 **Original Article**

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4 **Reliability of a smartphone electrocardiogram device in dogs: comparison with standard**  
5 **6-lead electrocardiography**

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21

22 **Abstract**

23 A prospective, multicenter, single-blind study was conducted in 166 dogs to assess the  
24 reliability of a smartphone electrocardiogram (ECG) device in evaluating heart rhythm and  
25 ECG measurements. A standard 6-lead ECG was acquired for 1 min in each dog. A  
26 smartphone ECG tracing was simultaneously recorded using a single-lead bipolar ECG  
27 recorder. All ECGs were reviewed by one blinded operator, who judged if tracings were  
28 acceptable for interpretation and assigned an electrocardiographic diagnosis. Agreement  
29 between smartphone and standard ECG in the interpretation of tracings was evaluated.  
30 Sensitivity and specificity for the detection of arrhythmia were calculated with the  
31 smartphone ECG.

32 Smartphone ECG tracings were interpretable in 162/166 (97.6%) tracings. A perfect  
33 agreement between the smartphone and standard ECG was found in detecting bradycardia,  
34 tachycardia, ectopic beats and atrioventricular blocks. A very good agreement was found in  
35 detecting sinus rhythm versus non sinus rhythm, with a 100% sensitivity and 97.9%  
36 specificity. The smartphone ECG provides tracings that are adequate for analysis in most dogs  
37 with a reliable assessment of heart rate, heart rhythm, atrioventricular blocks or ectopic beats.  
38 The smartphone device represents an additional tool in the management of dogs with  
39 arrhythmias, but does not substitute a 6-lead ECG. Arrhythmias identified by the smartphone  
40 device should be followed up with a standard ECG before taking clinical decisions.

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42

43 *Keywords:* Alivecor; Canine; Electrocardiography; iPhone; Mobile

44 **Introduction**

45 Many arrhythmias have paroxysmal presentation, while others require frequent  
46 monitoring due to the risk of progression. In these settings, serial electrocardiograms (ECG)  
47 are crucial for correct diagnosis and management. Clinical electrocardiography has thus  
48 undergone a continuous technological evolution since its invention by Willem Einthoven in  
49 1903, leading to the development of Holter monitoring, telemetry systems and loop recorders  
50 (Kennedy, 2013).

51  
52 One of the latest innovations is the 1-lead ECG recorded by smartphone devices using  
53 specific applications (Bruining et al., 2014; Walsh et al., 2014; Baquero et al., 2015). In  
54 human medicine there are many studies highlighting the accuracy of smartphone ECG  
55 tracings to measure the heart rate (HR) and evaluate heart rhythm (Lau et al.,2013; Ho et al.,  
56 2014; Haberman et al., 2015). Other studies have shown the good performances of  
57 smartphone ECG devices for diagnosing supraventricular tachycardia in pediatrics (Wackel et  
58 al., 2014; Ferdman et al., 2015; Nguyen et al., 2015), for detecting atrial fibrillation (Lau et  
59 al., 2013; Lee et al., 2013; McManus et al., 2013; Saxon, 2013; Orchard et al., 2014; Lowres  
60 et al., 2015a) and for identifying signs associated with myocardial ischemia (Wong, 2013). To  
61 the best of our knowledge only one preliminary study has been performed comparing a  
62 smartphone ECG device to standardized ECG tracings in dogs (Kraus et al., 2013). The aim  
63 of the present study was therefore to assess the use and reliability of a smartphone ECG to  
64 evaluate heart rhythm and ECG measurements in dogs.

65

66 **Materials and methods**

67 *Animals*

68           The study group included client-owned dogs that were referred to the Department of  
69   Veterinary Science of the University of Pisa or the Department of Cardiology of the Istituto  
70   Veterinario di Novara for a cardiologic consultation or assessment prior to anesthesia. The  
71   study was prospective, multicenter and single-blind. Dogs were recruited over a one-year  
72   period (December 2014-December 2015). Each case underwent a cardiac evaluation,  
73   including physical examination, standard 6-lead ECG and echocardiogram. The study  
74   protocol was reviewed and approved by the Institutional Welfare and Ethics Committee of the  
75   University of Pisa (permission number 39/2015).

76

#### 77   *ECG acquisition and analysis*

78           A standard 6-lead ECG (Elan 1100 ECG system, Cardioline; MAC 800 ECG system,  
79   GE Healthcare) was acquired for 1 min in conscious, unsedated dogs positioned in right  
80   lateral recumbency. Surface electrodes made of flattened alligator clips were attached to the  
81   skin at the level of the olecranon on the caudal aspect of the forelimb, and over the patellar  
82   ligaments on the cranial aspect of the hindlimbs, as previously described (Tilley, 1992).  
83   Rubbing alcohol was applied to maintain electrical contact with the skin. A sampling  
84   frequency of 1000Hz for standard ECG acquisition was used, with a 100 Hz low-pass filter  
85   and a 0.5 Hz high-pass filter to decrease respiratory artifact, as previously published  
86   (Hinchcliff, et al. 1997).

87           A smartphone ECG tracing was simultaneously recorded, starting and ending at the  
88   same time of the 6-lead ECG, using a single-lead bipolar ECG recorder (AliveCor Veterinary  
89   Heart Monitor, AliveCor, Inc.) and its application (AliveECG Vet, AliveCor, Inc.). Three  
90   operators (TV, CB, FM) recorded the smartphone ECGs with an iPhone 4S (Apple Inc.). The  
91   smartphone ECG was recorded placing it on the left precordial area of all the dogs. A cranio-  
92   caudal orientation of the smartphone case was used in each dog, with the camera side of the

93 smartphone located caudally (Fig. 1). In the short-haired dogs, a small amount of alcohol was  
94 placed on the left precordial area in order to obtain a good quality smartphone ECG signal. In  
95 the long-haired dogs, a small amount of alcohol was placed after shaving the left precordial  
96 area in order to acquire the same high-quality signal. Smartphone ECG recordings were  
97 automatically digitized by the device, sent via email and stored as a PDF file. For each dog,  
98 ECG tracings obtained with the two methods were printed at a paper speed of 50 mm/s with a  
99 gain of 10 mm/mV. The last 30 s of each ECG tracing were analyzed. Dogs with a  
100 smartphone ECG trace lasting < 30 s were excluded from the study.

101  
102 All ECG tracings were reviewed by a board-certified veterinary cardiologist (OD), in a  
103 blinded fashion, who judged whether the tracings were acceptable for interpretation. For all  
104 ECG tracings, the same operator evaluated the rhythm and performed ECG measurements.  
105 Measurements were achieved using the lead II of the standard ECG and using the only  
106 available lead of the smartphone ECG.

107  
108 In each case, the following measurements were performed: mean HR (beats per min,  
109 bpm); P wave amplitude (mV) and duration (ms); PQ interval duration (ms); R wave  
110 amplitude (mV); QRS complex duration (ms) and QRS polarity. The mean HR was calculated  
111 as the number of QRS complexes recorded in the 1 min ECG tracings (bpm). Other  
112 measurements were achieved as previously described (Kittleson and Kienle, 1998). The QRS  
113 polarity of the smartphone ECG traces was compared with lead II of the standard ECG.  
114 Finally, the mean HR calculated automatically by the smartphone application (App HR) was  
115 noted. Heart rate was classified as normal if between 70 and 160 bpm, bradycardia if < 70  
116 bpm and tachycardia if > 160 bpm, as previously described (Kittleson and Kienle, 1998).

117

118 *Statistical analysis*

119           The analysis was performed only with paired ECG tracings that were considered  
120 acceptable for interpretation, as defined by the operator, and the standard ECG was set as the  
121 reference method. Cohen's kappa ( $\kappa$ ) test was used to calculate the agreement between the  
122 smartphone ECG and standard ECG for HR classification (normal, bradycardia or  
123 tachycardia), heart rhythm (sinus rhythm, atrial fibrillation, ventricular rhythm,  
124 supraventricular rhythm), atrioventricular blocks (AVB) (absent, first-degree AVB, second-  
125 degree AVB, third-degree AVB), premature complexes (absent, ventricular, supraventricular),  
126 polarity of QRS complex (positive, negative). The kappa coefficient was interpreted as  
127 follows: values  $\leq 0.20$  as no agreement, 0.21–0.40 as fair, 0.41–0.60 as moderate, 0.61–0.80  
128 as good, 0.81–0.99 as very good, and 1 as perfect agreement. The sensitivity, specificity,  
129 positive and negative predictive values of the smartphone ECG to detect arrhythmia were  
130 calculated. In addition, the median and range of differences between the standard ECG and  
131 smartphone ECG were calculated for HR, amplitude of the P and R waves, for the duration of  
132 the P wave, PQ interval, QRS complex. Bland-Altman plots were used to show the  
133 differences between smartphone and standard ECG for numerical data. The Pearson or  
134 Spearman correlation coefficients were used to study correlations between HRs measured  
135 with the standard ECG and the smartphone ECG using values calculated either by the  
136 operator or automatically. The Shapiro-Wilk test was used to determine the normality of the  
137 datasets..

138           Statistical analysis was performed with a commercial software (GraphPad Prism 5). P  
139  $< 0.05$  was considered as significant.

140

141 **Results**

142 *Animals and feasibility*

143 A total of 166 dogs were enrolled in the study, of which 84 were males and 82 were  
144 females. The median age was nine years (ranging between 0.3 to 17 years) and median body  
145 weight was 25 kg (55.1 lb) with a range of 2.1 to 75 kg (4.6 to 165.3 lb). The majority of dogs  
146 (71/166, 43%) had cardiac diseases, both congenital or acquired; 32/166 (19%) dogs had  
147 neoplastic diseases, 30/166 (18%) were in the intensive care unit because of renal, respiratory,  
148 gastro-intestinal or neurologic diseases, and 33/166 (20%) were healthy dogs evaluated for  
149 pre-anesthesia assessment prior to elective surgeries.

150 The blinded operator judged 162/166 (97.6%) of the smartphone ECG tracings as  
151 acceptable for interpretation (Fig. 2, 3 and 4). In 4/166 (2.4%) cases the traces were deemed  
152 as non-interpretable; all 4 of the traces deemed uninterpretable were obtained from patients  
153 weighing <10kgs.

154

#### 155 *Heart rate*

156 According to the standard 6-lead ECG, 133/162 (82%) dogs had a normal HR, 20/162  
157 (12%) had tachycardia, and 9/162 (6%) had bradycardia. A perfect agreement ( $\kappa=1$ ) between  
158 the smartphone and standard ECG was found in the classification of HR when it was  
159 manually measured on digitized tracings (Table 1). A strong positive correlation was found  
160 between the HR values manually measured on standard ECGs and smartphone ECG tracings  
161 ( $r^2 = 0.99$ ;  $P < 0.0001$ ; Fig. 5). Median paired differences between the HR manually measured  
162 on standard ECG and smartphone ECG was 0 bpm (-10, +20 bpm; Table 2 and Fig. 6).

163 A strong positive correlation was also found between the App HR values and those  
164 manually measured on standard ECG tracings ( $r^2 = 0.923$ ;  $P < 0.0001$ ; Fig. 7). However, the  
165 App HR was less accurate than the manually measured HR on digitized standard ECG  
166 tracings ( $\kappa=0.91$ ). In 103/162 (63.6%) cases, the App HR underestimated the actual HR, with  
167 a median difference of -3 bpm; (-31, +20 bpm; Fig. 8). However in only 4/162 (2.5%) cases,

168 was there a misclassification of HR with the smartphone application. According to App HR,  
169 two dogs with tachycardia were classified as normal HR, one dog with normal HR was  
170 classified as bradycardia, and one dog with bradycardia was classified as normal HR. The  
171 greatest disagreement was found in a dog with severe bradycardia (40 bpm) secondary to a  
172 third-degree AVB in which the App HR read the P waves as QRS complexes, thus  
173 erroneously yielding an HR of 140 bpm.

174

### 175 *Heart rhythm*

176 The majority of dogs (141/162, 87%) had sinus rhythm or sinus arrhythmia; 14/162  
177 (9%) dogs had supraventricular arrhythmias; 7/162 (4%) dogs had ventricular rhythm or  
178 ventricular arrhythmias; 6/162 (4%) dogs had different types of AVBs.

179 Very good agreement ( $\kappa=0.94$ ) was found in the evaluation of the heart rhythm.  
180 Disagreement was found in only 3/162 (1.9%) cases, in which the sinus rhythm was  
181 erroneously classified as atrial rhythm due to the negative polarity of the P waves (one case)  
182 or as a slow atrial fibrillation due to non observable P waves (two cases) on the smartphone  
183 ECG trace (Table 3). In 128/141 (90.7%) cases of sinus rhythm, the smartphone ECG  
184 underestimated the amplitude of the P wave, with a median difference of -0.1 mV (-0.4; +0.1  
185 mV). The analysis of the P wave duration showed a median difference between the two  
186 methods of 0 ms (-20, +0 ms).

187 Considering all the arrhythmias taken together, the smartphone ECG had 100%  
188 sensitivity and 97.9% specificity in differentiating between sinus rhythm versus non sinus  
189 rhythm, with a positive predictive value of 87.5%, and a negative predictive value of 100%.

190



191 *QRS complex analysis*

192 A good agreement ( $\kappa=0.65$ ) was found in the polarity of the QRS complexes between  
193 the smartphone ECG and lead II of the standard 6-lead ECG (Fig. 2, 3, 4). The same QRS  
194 polarity was found in 158/162 (97.5%) cases. In three cases with positive polarity of the QRS  
195 complex in lead II, the smartphone tracing showed a negative QRS. In one case with negative  
196 polarity of the QRS complex in lead II, the smartphone tracing showed a positive QRS. The  
197 evaluation of the QRS duration showed a median difference of 0 ms (-20, +10 ms). Lastly, the  
198 smartphone ECG underestimated the amplitude of R wave in 121/162 (74.7%), with a median  
199 difference of -0.5 mV (-2.1; +1 mV), compared to the standard ECG.

200

201 *Ectopic beats*

202 A perfect agreement ( $\kappa=1$ ) between the smartphone ECG and standard ECG was found  
203 in the identification and classification of ectopic beats, including 16 cases with ventricular  
204 premature complexes, three cases with supraventricular premature complexes and four cases  
205 with both supraventricular and ventricular ectopic beats. In addition a perfect agreement was  
206 found regarding the polarity of ventricular premature complexes on the smartphone ECG  
207 tracings compared with lead II of the standard 6-lead ECG.

208

209 *Atrioventricular blocks*

210 A perfect agreement ( $\kappa=1$ ) between the smartphone ECG and standard ECG was found  
211 in the AVB diagnosis, including two cases with first-degree AVB, one with second-degree  
212 AVB and three cases with third-degree AVB. The PQ interval analysis using smartphone  
213 tracings was reliable in comparison to the standard ECG, with a median difference of 0 ms  
214 (range -20, +20 ms).

215

216 **Discussion**

217 In our investigation the smartphone ECG was easily performed in all dogs and 96.7%  
218 of tracings were deemed as interpretable. These results are in line with findings in human  
219 medicine where smartphone ECG tracings were interpretable in 87-99% of patients (Saxon,  
220 2013; Nguyen et al., 2015; Tarakji et al., 2015). The few tracings judged as non-interpretable  
221 were all recorded in small breed dogs, where motion artifacts are common, which likely  
222 accounted for the fact that the tracings were not readable.

223

224 In our study, the smartphone ECG was excellent in the HR evaluation in dogs. This is  
225 in accordance with a preliminary study in dogs, where a good agreement was found between  
226 smartphone ECG and reference ECG in the evaluation of instantaneous and mean HR (Kraus  
227 et al., 2013). In our investigation, the greatest reliability was found when the HR was  
228 manually measured on digitized tracings. Conversely the App HR was less reliable, since  
229 lower agreement was found between the HR values obtained by the smartphone device and  
230 the standard ECG. As the QRS complexes on smartphone ECG tracings had a low amplitude  
231 in most dogs, we hypothesize that the App HR may underestimate the HR due to the fact that  
232 some QRS complexes are not correctly interpreted by the instrument. In a few dogs, the App  
233 HR was totally unreliable. However, in only one case the disagreement was of a real clinical  
234 value: in a dog with severe bradycardia secondary to third-degree AVB, the App HR read the  
235 P waves as QRS complexes, thus erroneously resulting in a normal HR.

236

237 The smartphone ECG was very reliable in evaluating heart rhythm in dogs, as it  
238 showed 100% sensitivity and 97.9% specificity in the detection of arrhythmias. All cases of  
239 atrial fibrillation were correctly diagnosed, without false negatives. This result is similar to  
240 findings in humans where the sensitivity and specificity of the smartphone ECG in detecting

241 atrial fibrillation were 94-100% and 90-97%, respectively (Lau et al., 2013; Haberman et al.,  
242 2015; Tarakji et al., 2015). In humans, most false diagnoses of atrial fibrillation are due to  
243 small voltage P waves. Our results showed that the smartphone ECG underestimates the  
244 amplitude of the P wave. Despite this, the P waves remained clearly visible in the majority of  
245 dogs with sinus rhythm. In a few cases, however, the P waves were difficult to recognize and  
246 it was hard to determine between a sinus arrhythmia and atrial fibrillation. Consequently, two  
247 out of 141 cases of sinus rhythm were incorrectly classified as atrial fibrillation. In a small  
248 breed dog, the P waves had negative polarity on the smartphone ECG leading to the incorrect  
249 diagnosis of an atrial ectopic rhythm. A preliminary study in cats recommended positioning  
250 the smartphone case parallel to the long axis of the heart, with a more base-apex orientation in  
251 comparison to our cranio-caudal orientation (Stromberg and Kwart, 2015). It might be that in  
252 some small breed dogs, the orientation of the smartphone case should be individually adjusted  
253 to correctly visualize the P waves.

254

255         Atrial fibrillation is common in dogs with severe cardiac disease and increases the risk  
256 of cardiac-related death in those with myxomatous mitral valve degeneration and dilated  
257 cardiomyopathy (Calvert et al., 1997; Jung et al., 2016). Likewise, in humans, atrial  
258 fibrillation increases the chance of morbidity or mortality, and recent studies have highlighted  
259 the utility of the smartphone ECG in screening for this arrhythmia (Lau et al., 2013; Lee et al.,  
260 2013; McManus et al., 2013; Saxon, 2013; Orchard et al., 2014; Lowres et al., 2015b  
261 Haberman et al., 2015; Peritz et al. 2015;). Early diagnosis of atrial fibrillation is difficult in  
262 dogs. Our results show that the smartphone ECG may become a promising tool for frequent  
263 monitoring of dogs predisposed to atrial fibrillation. It could also be beneficial for dogs with  
264 atrial fibrillation that receive drugs to control HR. Holter monitoring is an essential tool for  
265 evaluating HR and in treating atrial fibrillation in dogs. However, 24-hour Holter monitoring

266 is expensive and necessitates the owner's compliance, hence its use may not always be  
267 practical. In the light of its ease and cost effectiveness, the smartphone ECG could represent a  
268 complementary tool for HR evaluation at home in dogs with atrial fibrillation.

269

270 The smartphone ECG showed a good reliability in the analysis of the QRS complex, in  
271 assessing both duration and polarity. In most dogs, QRS complexes displayed the same  
272 polarity on smartphone tracings and lead II of the 6-lead ECG, with a similar polarity in all  
273 cases of ventricular ectopic beats. In comparison to the standard ECG however, the  
274 smartphone device underestimated the R wave amplitude. In fact, wave amplitudes are  
275 extremely dependent on the electrocardiographic derivation method. Further studies are  
276 needed to establish reference values of wave amplitudes on smartphone ECG tracings.

277 In our opinion, smartphone tracings should not be used to assess the amplitude of ECG  
278 waves as a substitute for standard electrocardiograms and as a diagnostic method in the  
279 detection of chamber enlargement.

280

281 The smartphone ECG was highly reliable in the identification of ectopic beats.  
282 Ventricular premature complexes, accelerated idioventricular rhythms and ventricular  
283 tachycardias were easily identified in all cases with the smartphone ECG. One recent  
284 investigation used it as the sole electrocardiographic method in the identification of  
285 ventricular premature complexes in the screening of Doberman Pinschers for occult dilated  
286 cardiomyopathy (Gordon et al., 2015). It could thus be useful in screening or monitoring dogs  
287 with cardiomyopathies associated with ventricular arrhythmias.

288

289 With regard to the reliability of the smartphone ECG for AVBs, a good agreement  
290 with the standard ECG was found both in the evaluation of the PQ interval and in the

291 identification of the type of block. One study in humans described a higher percentage of false  
292 positives and negatives during the evaluation of AVBs compared to our results (Haberman et  
293 al., 2015). The authors reported motion artifacts (arm movement, muscle tension and tremor)  
294 as the main difficulties in AVB evaluation. None of the smartphone ECG tracings recorded  
295 motion artifacts that led to misdiagnosing AVBs. Thus, the agreement between devices was  
296 perfect, suggesting that the smartphone ECG can be helpful in the interpretation of AVBs in  
297 dogs.

298

299         Our investigation has some limitations. First, the study group was large but the  
300 number of dogs with arrhythmias was relatively low. A larger number of rhythm disturbances  
301 might have revealed a lower reliability of the smartphone ECG. However, most common  
302 types of canine arrhythmias were included in our study and in all these cases the smartphone  
303 ECG was consistent in diagnosing the arrhythmia. Second, the smartphone tracings were  
304 acquired by three operators but inter-operator variability in the quality of ECG recording was  
305 not evaluated.

306

### 307 **Conclusion**

308         In conclusion, the smartphone ECG can rapidly and simply record a single-lead ECG  
309 of good diagnostic quality in dogs. Tracing analysis performed by cardiologists reliably  
310 evaluated HR, heart rhythm, AVBs and ectopic beats.

311         The smartphone device does not substitute the 6-lead ECG or Holter monitoring but  
312 does represent an additional tool in the management of dogs with arrhythmias or in  
313 monitoring dogs at risk for heart rhythm disturbances. Therefore, any arrhythmia identified by  
314 the smartphone device should be followed by a standard 6-lead ECG and treatment decisions  
315 based upon smartphone ECG only are not recommended.

316 Further studies are needed to assess the diagnostic value of the smartphone ECG  
317 recorded by owners in a home setting.

318

### 319 **Conflict of interest statement**

320 None of the authors has any financial or personal relationships that could  
321 inappropriately influence or bias the content of the paper.

322

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440



441 **Table 1**

442 Agreement ( $\kappa$ ) between smartphone ECG and standard 6-lead ECG.

443

Type of analysis	$\kappa$	(95% CI)	Agreement
Manual HR	1		Perfect
App HR	0.91	(0.81-0.99)	Very good
Heart rhythm	0.94	(0.86-1)	Very good
AVBs	1		Perfect
Ectopic beats	1		Perfect
QRS polarity	0.65	(0.34 to 0.97)	Good

444

445 App HR, heart rate automatically measured by smartphone application; AVBs,

446 atrioventricular blocks; CI, confidence interval; Manual HR, heart rate manually measured on

447 printed ECG tracings.

448

449 **Table 2**

450 Differences between smartphone ECG and standard ECG in the evaluation of  
451 electrocardiographic parameters.

452

<b>Parameter</b>	<b>Difference</b>	<b>Range</b>
<b>Manual HR (bpm)</b>	0	-10; +20
<b>App HR (bpm)</b>	-3	-31; +20
<b>P (ms)</b>	0	-20; +0
<b>P (mV)</b>	-0,1	-0,4; +0,1
<b>PQ (ms)</b>	0	-20; +20
<b>QRS (ms)</b>	0	-20; +10
<b>R (mV)</b>	-0,5	-2,1; +1

453

454 Median difference and range are reported.

455 App HR, heart rate automatically measured by smartphone application; Manual HR, heart rate  
456 manually measured on printed ECG tracings.

457

458 **Table 3**

459 Agreement between smartphone ECG and standard 6-lead ECG in heart rhythm identification

460 in 162 dogs.

461

<b>Standard ECG</b>	<b>Smartphone ECG</b>				<b>Total</b>
	<b>S</b>	<b>AF</b>	<b>SV</b>	<b>V</b>	
<b>S</b>	138	2	1	0	141
<b>AF</b>	0	12	0	0	12
<b>SV</b>	0	0	2	0	2
<b>V</b>	0	0	0	7	7
<b>Total</b>	138	14	3	7	162

462

463 AF, atrial fibrillation; S, sinus rhythm; SV, supraventricular rhythm; V, ventricular rhythm.

464

465 **Figure legends**

466

467 Fig. 1. Cranio-caudal orientation of the smartphone in a dog. The camera side of the  
468 smartphone was located caudally.

469

470 Fig. 2. Sinus rhythm with standard ECG (A) and with smartphone ECG (B) in the same dog.

471 Paper speed = 50 mm/s; 10 mm = 1 mV.

472

473 Fig. 3. Atrial fibrillation with standard ECG (A) and with smartphone ECG (B) in the same

474 dog. Paper speed = 50 mm/s; 10 mm = 1 mV.

475

476 Fig. 4. Third-degree AVB with standard ECG (A) and with smartphone ECG (B) in the same

477 dog. Paper speed = 50 mm/s; 5 mm = 1 mV.

478

479 Fig. 5. Pearson test showing a strong positive correlation between the HR values manually

480 measured on standard ECGs and smartphone ECG tracings ( $r^2 = 0.99$ ;  $P < 0.0001$ ).

481

482 Fig. 6. Bland-Altman plot showing differences between HR values manually measured on

483 standard ECG and smartphone ECG tracings.

484

485 Fig. 7. Pearson test showing a strong positive correlation between the HR values manually

486 measured on standard ECGs and HR values produced by the smartphone application ( $r^2 =$

487  $0.92$ ;  $P < 0.0001$ ).

488

489 Fig. 8. Bland-Altman plot showing differences between the HR values manually measured on  
490 standard ECGs and HR values produced by the smartphone application.  
491