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Cardiac rhythm analysis during ongoing cardiopulmonary resuscitation using the “Analysis During Compressions with Fast Reconfirmation” (ADC-FR) technology

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1 **Cardiac rhythm analysis during ongoing cardiopulmonary resuscitation using the**
2 **“Analysis During Compressions with Fast Reconfirmation” (ADC-FR) technology**

3
4 **Short title:** Rhythm analysis during compressions

5
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11
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24 **Abstract**

25 **Background.** Pauses in chest compressions (CCs) have a negative association with survival
26 from cardiac arrest (CA). ECG rhythm analysis and defibrillator charging are significant
27 contributors to CC pauses.

28 **Objective.** Accuracy of the “Analysis During Compressions with Fast Reconfirmation” (ADC-
29 FR) algorithm, which features automated rhythm analysis and charging during CCs to reduce CC
30 pauses, was retrospectively determined in a large database of ECGs from 2,701 out-of-hospital
31 CAs.

32 **Methods.** The ADC-FR algorithm generated a total of 7,264 advisories, of which 3,575 were
33 randomly assigned to a development dataset and 3,689 to a test one. With ADC-FR, a high-pass
34 digital filter is used to remove CC artifacts, while the underlying ECG rhythm is automatically
35 interpreted. When CCs are paused at the end of the 2-min CPR interval, a 3 sec reconfirmation
36 analysis is performed using the artifact-free ECG to confirm the shock/no-shock advisory.
37 Sensitivity and specificity of the ADC-FR algorithm in correctly identifying shockable/non-
38 shockable rhythms during CCs were calculated.

39 **Results.** In both the datasets, the accuracy of the ADC-FR algorithm for each ECG rhythm
40 exceeded the recommended performance goals, which apply to a standard artifact-free ECG
41 analysis. Sensitivity and specificity were 97% and 99%, respectively, for the development
42 dataset, and 95% and 99% for the test dataset.

43 **Conclusion.** The ADC-FR algorithm is highly accurate in discriminating shockable and non-
44 shockable rhythms and can be used to reduce CC pauses.

45

46 **Key words.** Cardiac arrest; rhythm analysis; automated external defibrillation; pre-shock pauses;
47 chest compression; defibrillation.

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49 **Introduction**

50 Cardiopulmonary resuscitation (CPR) in conjunction with prompt electrical defibrillation can re-
51 establish spontaneous circulation (ROSC) after cardiac arrest (CA) from ventricular fibrillation
52 (VF) and pulseless ventricular tachycardia (VT).¹ Nevertheless, resuscitative efforts are often
53 unsuccessful and poor outcomes may result from ineffective and/or frequently interrupted chest
54 compressions (CCs).²⁻⁶

55

56 Among the different causes for interrupting CCs during CPR, are the pre-shock pauses mandated
57 by automated external defibrillators (AEDs). CCs create artifacts on the electrocardiographic
58 (ECG) signal, such that interruptions are mandatory for rhythm analysis prior to a defibrillation
59 attempt.⁷⁻⁹ Limiting the frequency and the duration of such CC interruptions may improve
60 outcomes of CA.¹⁰⁻¹⁴

61

62 A novel technology was developed to limit CC interruptions required for both rhythm analysis
63 and defibrillator charging. This technology, called “Advisory During CPR with Fast
64 Reconfirmation” (ADC-FR), features automated ECG analysis and defibrillator charging during
65 ongoing CCs, with a 3 sec ECG rhythm reconfirmation analysis.¹⁵ The purpose of the present
66 study was to investigate the sensitivity and specificity of the ADC-FR analysis algorithm on a
67 large dataset of ECG traces with CC artifacts obtained from prehospital CAs.

68

69 **Methods**

70 A database of defibrillator records (AED Pro[®], AED Plus[®], E Series) collected during
71 prehospital CPR was used to develop and test the ADC-FR algorithm. The database, managed by

72 ZOLL Medical Corporation (Chelmsford, MA), included field case submissions from multiple
73 emergency medical services (EMS) agencies between 2004-2014. The electronic data did not
74 contain any patient's identifiable information, in compliance with the Health Insurance
75 Portability and Accountability Act (HIPAA) regulations.

76
77 ECG was recorded at a sample rate of 250 Hz. CCs were detected using an accelerometer and
78 acceleration data were sampled at 125 Hz. The acceleration records associated with the ECG
79 traces were manually inspected to identify CC intervals (continuous CCs ≥ 15 sec) and
80 subsequent pauses (pauses in CCs ≥ 11 sec, regardless of the reason for the pause). All ECG
81 segments matching the above criteria were included. The included ECG traces were blindly and
82 randomly partitioned into a development dataset and a test dataset, and subsequently processed
83 by the ADC-FR algorithm, which generated a shock/no shock decision for each ECG trace.
84 Sensitivity and specificity of the ADC-FR algorithm were calculated based on comparison of the
85 automated analysis results with a corresponding expert reviewers' rhythm annotation (QT, NZ,
86 GR). ECG rhythms were evaluated and coded according to the recommendations for specifying
87 and reporting arrhythmia analysis algorithm performance from the American Heart Association
88 (AHA).¹⁶ Since the aim of the ADC-FR algorithm was to discriminate between shockable and
89 non-shockable rhythms, a simplified rhythm categorization was used, as detailed in Table 1.
90 Methods details are reported in the Supplemental Methods.

91
92 *ADC-FR Technology*

93 The ADC-FR technology uses the signal from the accelerometer embedded in the defibrillation
94 pads (CPR-D padz® or CPR-stat padz®), to identify the presence of CCs. When CCs are

95 detected, a high-pass digital filter is used to minimize CC artifacts from the ECG signal. A
96 previously validated algorithm for ECG analysis during ongoing CCs is then applied to the
97 filtered trace in order to determine whether or not the patient's rhythm is shockable.¹⁷
98 Subsequently, upon interruption of CCs and settling of the ECG, the ADC-FR algorithm
99 performs a 3 sec analysis using the compression-free ECG trace to reconfirm the decision
100 determined during CCs. This reconfirmation analysis is compared against the previous analysis
101 during CCs and the shock/no-shock decision is immediately made if both match. For the clinical
102 implementation of the feature, the ADC-FR algorithm is applied only at the end of the pre-
103 configured 2-min CPR interval, as detailed in Figure 1. The capacitor of the defibrillator is
104 automatically charged 4 sec before the end of the 2-min CPR interval, allowing for immediate
105 defibrillation after the reconfirmation pause, if a shockable rhythm is confirmed. The same 3 sec
106 analysis during CC pause occurs in the instance of a non-shockable rhythm. In this case, the
107 AED issues a "no shock advised" order and CCs can restart promptly (Figure 1).

108 If a shock/no shock decision cannot be made using the combination of analysis during CCs and
109 the 3 sec reconfirmation analysis, an additional segment of ECG is analyzed. In the instance that
110 a shock/no-shock decision cannot be made after two ECG segments, a final ECG segment is
111 analyzed. In summary, the ADC-FR algorithm makes the shock/no-shock determination during
112 CCs based on three 3 sec segments (9 sec total); then, when CCs are paused, the algorithm
113 performs the reconfirmation analysis, again based on a 3 sec segment (requiring from a
114 minimum of one up to three, i.e. 3-9 sec). Supplemental Figure 1 provides more details on the
115 ADC-FR algorithm, while Supplemental Figure 2 describes the logical decision algorithm of the
116 ADC-FR technology. Samples of raw ECG traces for different rhythms, correctly interpreted by
117 the ADC-FR algorithm, are reported in Figure 2.

118

119 *Statistical analysis*

120 The performance of the ADC-FR algorithm was evaluated in terms of accuracy of the shockable
121 or non-shockable decision. Accuracy was defined as the number of correct advisories (shockable
122 or non-shockable) divided by the total number of advisories for each ECG rhythm. Sensitivity
123 was defined as the number of ECG rhythms correctly classified as shockable divided by the total
124 number of shockable rhythms. Specificity was defined as the number of ECG rhythms correctly
125 classified as non-shockable (Table 1) divided by the total number of non-shockable rhythms.
126 Accuracy of the ADC-FR algorithm for each rhythm was compared to the AHA
127 recommendations for arrhythmia algorithm performance.¹⁶ Calculation of confidence intervals
128 (CI) is reported in the Supplemental Methods.

129

130 **Results**

131 A total of 7,264 CC intervals with one of the ECG rhythms listed in Table 1, from 2,701 CA
132 patients were included in the analysis (3.8 ± 2.9 segments/patient). Of these, 3,575 were randomly
133 assigned to the development dataset, while the remaining 3,689 to the test one. The different
134 ECG rhythms included in the development and test datasets are reported in Table 2. For both
135 datasets, the number of intervals analyzed for each ECG rhythm exceeded the minimum required
136 sample size from the AHA recommendations.¹⁶

137

138 The ADC-FR algorithm accuracy for each rhythm exceeded the recommended arrhythmia
139 algorithm performance goals, even though these were set for artifact-free ECGs.¹⁶ The accuracy
140 of the ADC-FR in identifying non-shockable rhythms (i.e. normal sinus rhythm, asystole, and

141 other non-shockable rhythms) ranged between 99% and 100% (Table 2). Similarly, the accuracy
142 for identification of shockable rhythms (i.e. coarse VF or rapid VT) was between 99-100%. For
143 the rhythms without specific performance goal recommendations, i.e. fine VF and other VTs, the
144 accuracy ranged between 91-100%. Even the 90% lower CI for each rhythm exceeded the
145 performance goals.¹⁶

146 The overall performance of the ADC-FR algorithm yielded a sensitivity of 97% and 95% in the
147 development and test dataset respectively, while the specificity was 99% in both the datasets.
148 Considering only coarse VF and rapid VT and excluding fine VF, the sensitivity increased to
149 100%, in both datasets.

150
151 The algorithm accuracy was re-evaluated considering only one interval of any ECG rhythm from
152 each patient (n=4,544), as suggested in the AHA recommendations¹⁶ and the results are reported
153 in the Supplemental Table. Again, accuracy exceeded performance goals,¹⁶ ranging between 98-
154 100% for coarse VF, rapid VT, NSR, asystole, and other non-shockable rhythms, and between
155 89-100% for fine VF and other VT, in the development and test datasets. Specificity remained at
156 99% in both databases and sensitivity was 96% and 94% in the development and test datasets,
157 respectively.

158
159 The majority of ECG rhythms (81% in the development dataset and 83% in the test one), were
160 correctly identified using a combination of analysis during CCs and one 3 sec reconfirmation
161 analysis (median=3 sec, IQR=3, 3 sec; Table 3). In 16% of the instances in both datasets, an
162 additional analysis was needed. A third analysis was necessary in 3% of cases in the

163 development dataset and in 1% of the test dataset. Overall, fine VF, other VT, and asystole
164 represented the rhythms that required more re-analyses.

165

166 **Discussion**

167 This study demonstrated that the newly developed ADC-FR algorithm, which features ECG
168 rhythm analysis during CCs with the need for a brief pre-shock CPR pause for rhythm
169 reconfirmation, is highly accurate in discriminating shockable and non-shockable rhythms. The
170 ADC-FR algorithm yielded a sensitivity greater than 95% and a specificity greater than 99% for
171 identification of a shockable/non-shockable rhythm. The accuracy of the ADC-FR algorithm for
172 each ECG rhythm exceeded the arrhythmia analysis performance goals recommended by the
173 AHA, which apply to a standard artifact-free ECG analysis.¹⁶ Moreover, in 83% of the instances,
174 the ECG rhythm was correctly identified and a shock/no shock decision was made with one 3 sec
175 reconfirmation analysis.

176

177 Although some randomized clinical trials comparing 2000 vs. 2005 CPR Guidelines showed that
178 shortening pre-shock and post-shock CC pauses and increasing the CC fraction (CCF) did not
179 improve survival,^{18, 19} more recent evidence suggests that when controlling for the effects of
180 other resuscitation interventions, higher CCF was predictive of survival.²⁰ Moreover, in several
181 other studies, it has been demonstrated that during CPR, greater CCFs were associated with
182 higher likelihood of ROSC and survival after out-of-hospital cardiac arrests.^{4, 6, 10-14, 21} However,
183 CC pauses as long as 32 sec have been recently described during pre-hospital CPR. Among
184 these, peri-shock pauses accounted for the longest CC interruptions, with more than 23 sec.⁶ In a
185 study of more than 800 CA patients with a shockable rhythm, the odds of survival were

186 significantly lower for patients with pre-shock pauses > 20 sec, while pre-shock pauses < 10 sec
187 and CCF $> 60\%$ were associated with improved survival.¹¹

188
189 Implementing the proposed ADC-FR algorithm in a defibrillator can significantly reduce pre-
190 shock pauses. With a standard AED algorithm, CCs have to be interrupted after each 2-min CPR
191 cycle for rhythm analysis, charging of the defibrillator capacitor, warning the rescuer to stand
192 clear from the patient, and delivering of the shock.⁷ The ADC-FR technology, instead, allows for
193 accurate automated rhythm analysis during ongoing CCs and for automatic charging of the
194 defibrillator at 4 sec before the end of the timed CC interval, with a 3 sec reconfirmation analysis
195 once the ECG is free of CC artifacts. Compared to the earlier algorithm we published in 2008,¹⁷
196 the ADC-FR one includes the capability to quickly detect the end of the CC interval, to perform
197 the reconfirmation analysis, and to compare its result against the previous analysis during CCs to
198 achieve the shock/no-shock decision. These adjuncts contribute to the higher accuracy of the
199 ADC-FR technology compared to the earlier one,¹⁷ as shown from this study in a large and
200 complex (for the variety of rhythms) ECG dataset.

201
202 An earlier study investigated the duration of CC pauses in simulated CPR on manikins with the
203 use of a defibrillator set either to a standard AED mode or to the ADC-FR mode.¹⁵ Although the
204 rescuers received no specific information or training on ADC-FR apart from the instruction to
205 perform CPR following the defibrillator prompts, overall CC interruptions at the end of each
206 CPR interval were significantly reduced by almost 5 sec, for both shockable and non-shockable
207 rhythms, when the new technology was employed. In our database, more than 80% of ECG
208 rhythms were correctly identified during CCs, and thus required one 3 sec reanalysis period prior

209 to the shock/non-shock prompt. A rescuer trained in the use of a defibrillator equipped with
210 ADC-FR technology can use this brief pause to assure nobody is touching the patient and to
211 prepare for immediate shock delivery. In the other 20% of instances, the ECG rhythm identified
212 during CCs was not confirmed, suggesting that either the rhythm changed between CCs and the
213 subsequent pause or the analysis during CCs was wrong due to excessive artifacts. Nevertheless,
214 only in 1-3% of instances, 2 additional reconfirmations were necessary, supporting the
215 hypothesis that overall this algorithm would greatly reduce the pre-shock pauses.

216 The duration of CC interruptions to interpret ECGs might be particularly severe since some AED
217 can require more than 20 sec of “hands off” in order to perform a reliable rhythm analysis and
218 charging the capacitor.^{9, 22} The purpose of filtering compression artifacts from the ECG signal is
219 to enable rhythm analysis during uninterrupted CCs, shortening the pre-shock pause and
220 ultimately improving resuscitation outcome.^{6, 10-14} Human ECG and CPR artifacts, however,
221 show a large spectral overlap that makes the filtering approach difficult.⁷ Over the years,
222 considerable effort has been dedicated to developing more sophisticated methods of rhythm
223 analyses during CCs. The numerous signal processing techniques can be summarized into two
224 major approaches, either based on adaptive filters for the suppression of artifacts or on
225 algorithms for analyses performed directly on the corrupted ECG.²³⁻²⁹ However, overall the
226 proposed algorithms achieved a sensitivity > 90%, but a specificity ranging from 79.9% to 93%.
227 Although the sensitivity of the shock advisory algorithms greatly improved over time, the
228 specificity for identification of non-shockable rhythms remained below the recommended level.
229 Indeed, achieving a high specificity is a major determinant for the clinical use of a new algorithm
230 for rhythm analysis during CPR, because an insufficient specificity may erroneously cause
231 inappropriate shock delivery to patients with non-shockable rhythms.²³ The performance goals

232 recommended by the AHA task force on AEDs require a sensitivity > 90% for VF and a
233 specificity > 95% for non-shockable rhythms.¹⁶ Our study validated the ADC-FR algorithm in
234 CC-corrupted ECG traces from 2,701 CA patients. The ADC-FR algorithm, with rhythm
235 analysis during CCs and a quick reconfirmation analysis during CC pause, demonstrated a very
236 high accuracy in rhythm identification, with a sensitivity greater than 95% for identification of a
237 shockable rhythm and a specificity greater than 99% for identification of a non-shockable one.
238 Only in the instance of fine VF was the algorithm accuracy lower (i.e. 91-93%); however, fine
239 VF is considered to be a rhythm for which the benefits of defibrillation are limited or uncertain.¹⁶
240 For this reason, when the rhythm did not meet any shock or no shock criteria, for safety reasons
241 the ADC-FR algorithm considered it as non-shockable. Thus, the accuracy of the ADC-FR
242 technology for each rhythm exceeded the recommended performance goals.¹⁶ The 3 sec pre-
243 shock analysis for rhythm reconfirmation allows to discriminate those cases misinterpreted
244 during CCs and those in which a change in the rhythm might occur during CCs,³⁰ while being
245 short enough to not negatively affect outcomes. In a small retrospective analysis of data on pre-
246 and post-shock CC pauses from 36 VF patients, a pre-shock pause as brief as 3 sec was
247 associated with a 6-fold increase in the likelihood of ROSC, compared to longer pauses.¹⁴
248 Similarly, with all the limitations related to animal protocols, in a swine model of cardiac arrest,
249 a 100% ROSC was documented with pre-shock intervals of 3 sec, while a 0% ROSC occurred
250 when the pre-shock pauses increased to 15 sec.¹⁰

251

252 Several approaches have been employed to reduce the pre-shock pause. The use of the
253 defibrillator in manual mode, as opposed to AED mode, has been tested because hands-off time
254 for rhythm analysis and defibrillator charging can be shorter when a defibrillator is operated in

255 manual mode.^{31, 32} However, analysis accuracy has been reported to be lower with manual
256 compared to automatic analysis.³² Moreover, only advanced life support providers are trained
257 and allowed to perform manual rhythm analysis, while for basic life support providers, the
258 hands-off time for cardiac rhythm analysis associated with an AED operation is unavoidable.
259 Pre-shock pauses have been also reduced by 42% in the prehospital setting using a technology
260 that featured automated charging during compressions.³³ To reduce the impact of long charging
261 times, resumption of CPR during the charging process has been also suggested.⁴ Finally, rhythm
262 analysis during pauses for ventilations has been proposed, but the accuracy of this approach was
263 lower than standard AED analysis.³⁴ The ADC-FR technology enables rhythm analysis during
264 ongoing CCs while maintaining high accuracy; CC interruptions are further reduced by pre-
265 charging of the defibrillator capacitor. A combination with algorithms to predict defibrillation
266 success,³⁵ reducing the number of futile defibrillation attempts, may further improve outcome of
267 CPR.

268
269 We acknowledge several limitations of this study. It was a retrospective data analysis; however,
270 the randomized bifurcation of the database into a development and a test one, as recommended
271 by the AHA,¹⁶ eliminates some possibility of bias. No assessment was performed on how the
272 effect of CPR quality may have influenced algorithm performance. Nevertheless, the high
273 accuracy of the proposed technology has been demonstrated in a large database of ECG traces
274 from multiple EMS, which should guarantee the representativeness of the data and the normal
275 distribution of CPR quality usually encountered in the field. Finally, the application of the ADC-
276 FR algorithm depends on the availability of a CC-sensor for detecting of compressions, and
277 currently this may limit its use in both manual defibrillators and AEDs.

278

279 Conclusions

280 The ADC-FR is a highly accurate shock decision algorithm, which may be incorporated and used
281 in defibrillators to greatly reduce pre-shock pauses in CCs during CPR. Clinical studies are
282 required to investigate the impact of the ADC-FR algorithm on interruptions in CPR and
283 outcome.

284

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287

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290

291 **References**

- 292 1. Perkins GD, Handley AJ, Koster RW, Castrén M, Smyth MA, Olasveengen T, Monsieurs
293 KG, Raffay V, Gräsner JT, Wenzel V, Ristagno G, Soar J. European Resuscitation Council
294 Guidelines for Resuscitation 2015: Section 2. Adult basic life support and automated
295 external defibrillation. *Resuscitation* 2015;95:81-99.
- 296 2. Berdowski J, Berg RA, Tijssen JGP, Koster RW. Global incidences of out-of-hospital
297 cardiac arrest and survival rates: Systematic review of 67 prospective studies.
298 *Resuscitation* 2010;81:1479–1487.
- 299 3. Wik L, Kramer-Johansen J, Myklebust H, Myklebust H, Sørebo H, Svensson L, Fellows B,
300 Steen PA. Quality of cardiopulmonary resuscitation during out-of-hospital cardiac arrest.
301 *JAMA* 2005;293:299–304.
- 302 4. Soar J, Nolan JP, Böttiger BW, Perkins GD, Lott C, Carli P, Pellis T, Sandroni C, Skrifvars
303 MB, Smith GB, Sunde K, Deakin CD. European Resuscitation Council Guidelines for
304 Resuscitation 2015: Section 3. Adult advanced life support. *Resuscitation* 2015;95:100-
305 147.
- 306 5. Wissenberg M, Lippert FK, Folke F, et al. Association of national initiatives to improve
307 cardiac arrest management with rates of bystander intervention and patient survival after
308 out-of-hospital cardiac arrest. *JAMA* 2013;310:1377-1384.
- 309 6. Brouwer TF, Walker RG, Chapman FW, Koster RW. Association between chest
310 compression interruptions and clinical outcomes of ventricular fibrillation out-of-hospital
311 cardiac arrest. *Circulation* 2015;132:1030-1037.

- 312 7. Affatato R, Li Y, Ristagno G. See through ECG technology during cardiopulmonary
313 resuscitation to analyze rhythm and predict defibrillation outcome. *Curr Opin Crit Care*
314 2016;22:199-205.
- 315 8. Eftestøl T, Sunde K, Steen PA. Effects of interrupting precordial compressions on the
316 calculated probability of defibrillation success during out-of-hospital cardiac arrest.
317 *Circulation* 2002;105:2270-2273.
- 318 9. Snyder D, Morgan C. Wide variation in cardiopulmonary resuscitation interruption
319 intervals among commercially available automated external defibrillators may affect
320 survival despite high defibrillation efficacy. *Crit Care Med* 2004;32:S421-424.
- 321 10. Yu T, Weil MH, Tang W, Sun S, Klouche K, Povoas H, Bisera J. Adverse outcomes of
322 interrupted precordial compression during automated defibrillation. *Circulation*
323 2002;106:368-372.
- 324 11. Cheskes S, Schmicker RH, Christenson J, et al. Perishock pause: an independent predictor
325 of survival from out-of-hospital shockable cardiac arrest. *Circulation* 2011;124:58–66.
- 326 12. Cheskes S, Schmicker RH, Verbeek PR, et al. The impact of peri-shock pause on survival
327 from out-of-hospital shockable cardiac arrest during the Resuscitation Outcomes
328 Consortium PRIMED trial. *Resuscitation* 2014;85:336–342.
- 329 13. Vaillancourt C, Everson-Stewart S, Christenson J, Andrusiek D, Powell J, Nichol G,
330 Cheskes S, Aufderheide TP, Berg R, Stiell IG. The impact of increased chest compression
331 fraction on return of spontaneous circulation for out-of-hospital cardiac arrest patients not
332 in ventricular fibrillation. *Resuscitation* 2011;82:1501–1507.

- 333 14. Sell RE, Sarno R, Lawrence B, Castillo EM, Fisher R, Brainard C, Dunford JV, Davis DP.
334 Minimizing pre- and post-defibrillation pauses increases the likelihood of return of
335 spontaneous circulation (ROSC). *Resuscitation* 2010;81:822–825.
- 336 15. Partridge R, Tan Q, Silver A, Riley M, Geheb F, Raymond R. Rhythm analysis and
337 charging during chest compressions reduces compression pause time. *Resuscitation*
338 2015;90:133-137.
- 339 16. Kerber RE, Becker LB, Bourland JD, Cummins RO, Hallstrom AP, Michos MB, Nichol G,
340 Ornato JP, Thies WH, White RD, Zuckerman BD. Automatic external defibrillators for
341 public access defibrillation: Recommendations for specifying and reporting arrhythmia
342 analysis algorithm performance, incorporating new waveforms, and enhancing safety. A
343 statement for health professionals from the American Heart Association task force on
344 automatic external defibrillation, subcommittee on AED safety and efficacy. *Circulation*
345 1997;95:1677-1682.
- 346 17. Tan Q, Freeman G, Geheb F, Bisera J. Electrocardiographic analysis during uninterrupted
347 cardiopulmonary resuscitation. *Crit Care Med* 2008;36:S409-412.
- 348 18. Jost D, Degrange H, Verret C, Hersan O, Banville IL, Chapman FW, Lank P, Petit JL,
349 Fuilla C, Migliani R, Carpentier JP. DEFI 2005: a randomized controlled trial of the effect
350 of automated external defibrillator cardiopulmonary resuscitation protocol on outcome
351 from out-of-hospital cardiac arrest. *Circulation* 2010;121:1614-1622.
- 352 19. Beesems SG, Berdowski J, Hulleman M, Blom MT, Tijssen JG, Koster RW. Minimizing
353 pre- and post-shock pauses during the use of an automatic external defibrillator by two
354 different voice prompt protocols. A randomized controlled trial of a bundle of measures.
355 *Resuscitation* 2016;106:1-6.

- 356 20. Wik L, Olsen JA, Persse D, Sterz F, Lozano M, Brouwer MA, Westfall M, Souders CM,
357 Travis DT, Herken UR, Lerner EB. Why do some studies find that CPR fraction is not a
358 predictor of survival? *Resuscitation* 2016;104:59-62.
- 359 21. Christenson J, Andrusiek D, Everson-Stewart S, et al. Chest compression fraction
360 determines survival in patients with out-of-hospital ventricular fibrillation. *Circulation*
361 2009;120:1241-1247.
- 362 22. Savastano S, Vanni V, Burkart R, et al. Comparative performance assessment of
363 commercially available automatic external defibrillators: A simulation and real-life
364 measurement study of hands-off time. *Resuscitation* 2017;110:12–17.
- 365 23. Gong Y, Chen B, Li Y. A review of the performance of artifact filtering algorithms for
366 cardiopulmonary resuscitation. *J Healthc Eng* 2013;4:185-202.
- 367 24. Ruiz J, Irusta U, Ruiz De Gauna S, Eftestøl T. Cardiopulmonary resuscitation artefact
368 suppression using a Kalman filter and the frequency of chest compressions as the reference
369 signal. *Resuscitation* 2010;81:1087-1094.
- 370 25. Werther T, Klotz A, Granegger M, Alonso E, Eftestøl T, Kramer-Johansen J. Suppression
371 of the cardiopulmonary resuscitation artefacts using the instantaneous chest compression
372 rate extracted from the thoracic impedance. *Resuscitation* 2012;83:692-698.
- 373 26. Eilevstjønn J, Eftestøl T, Aase SO, Myklebust H, Husøy JH, Steen PA. Feasibility of shock
374 advice analysis during CPR through removal of CPR artefacts from human ECG.
375 *Resuscitation* 2004;61:131–141.
- 376 27. Aramendi E, Ayala U, Irusta U, Alonso E, Eftestøl T, Kramer-Johansen J. Suppression of
377 the cardiopulmonary resuscitation artefacts using the instantaneous chest compression rate
378 extracted from the thoracic impedance. *Resuscitation* 2012;83:692–698.

- 379 28. Rad AB, Engan K, Katsaggelos AK, Kvaløy JT, Wik L, Kramer-Johansen J, Irusta U,
380 Eftestøl T. Automatic cardiac rhythm interpretation during resuscitation. *Resuscitation*
381 2016;102:44-50.
- 382 29. Li Y, Bisera J, Geheb F, Tang W, Weil MH. Identifying potentially shockable rhythms
383 without interrupting cardiopulmonary resuscitation. *Crit Care Med* 2008;36:198–203.
- 384 30. Steinberg MT, Olsen JA, Brunborg C, et al. Minimizing pre-shock chest compression
385 pauses in a cardiopulmonary resuscitation cycle by performing an earlier rhythm analysis.
386 *Resuscitation* 2015;87:33-37.
- 387 31. Garza AG, Gratton MC, Salomone JA, Lindholm D, McElroy J, Archer R. Improved
388 patient survival using a modified resuscitation protocol for out-of-hospital cardiac arrest.
389 *Circulation* 2009;119:2597–2605.
- 390 32. Pytte M, Pedersen TE, Ottem J, Rokvam AS, Sunde K. Comparison of hands-off time
391 during CPR with manual and semi-automatic defibrillation in a manikin model.
392 *Resuscitation* 2007;73:131–136.
- 393 33. Bobrow BJ, Vadeboncoeur TF, Stolz U, Silver AE, Tobin JM, Crawford SA, Mason TK,
394 Schirmer J, Smith GA, Spaite DW. The influence of scenario-based training and real-time
395 audiovisual feedback on out-of-hospital cardiopulmonary resuscitation quality and survival
396 from out-of-hospital cardiac arrest. *Ann Emerg Med* 2013;62:47–56.
- 397 34. Ayala U, Irusta U, Ruiz J, Ruiz de Gauna S, González-Otero D, Alonso E, Kramer-
398 Johansen J, Naas H, Eftestøl T. Fully automatic rhythm analysis during chest compression
399 pauses. *Resuscitation* 2015;89:25-30.
- 400 35. Ristagno G, Li Y, Fumagalli F, Finzi A, Quan W. Amplitude spectrum area to guide
401 resuscitation-a retrospective analysis during out-of-hospital cardiopulmonary resuscitation

402 in 609 patients with ventricular fibrillation cardiac arrest. Resuscitation 2013;84:1697-
403 1703.

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404 **Table 1.** ECG rhythm categories identified by the “Analysis During Compressions with Fast
 405 Reconfirmation” (ADC-FR) algorithm

ECG rhythm categories
<i>Shockable</i>
Coarse ventricular fibrillation (Coarse VF)
Ventricular tachycardia with rate ≥ 150 beats/min (Rapid VT)
<i>Non-shockable</i>
Normal sinus rhythm (NSR)
Asystole
Any other non-shockable rhythms: atrial fibrillation, atrial flutter, supraventricular tachycardia, sinus bradycardia, premature ventricular contractions, second- or third-degree heart block, idioventricular rhythm
<i>Intermediate rhythms</i>
Fine ventricular fibrillation (Fine VF)
Ventricular tachycardia with rate < 150 beats/min (Other VT)

406 **Table 2.** “Analysis During Compressions with Fast Reconfirmation” (ADC-FR) algorithm accuracy in the development and in the test
 407 datasets

Rhythms	Recommendations ¹⁶		ADC-FR Development Dataset				ADC-FR Test Dataset			
	Minimum test sample size	Performance Goal (%)	Total Segments	Correctly Analyzed	Incorrectly Analyzed	Observed Accuracy (%) [90% low CI]	Total Segments	Correctly Analyzed	Incorrectly Analyzed	Observed Accuracy (%) [90% low CI]
Coarse VF	200	> 90	276	275	1	99 [97]	342	338	4	99 [96]
Rapid VT	50	> 75	58	58	0	100 [91]	58	58	0	100 [91]
NSR	100	> 99	341	341	0	100 [98]	419	419	0	100 [99]
Asystole	100	> 95	926	920	6	99 [98]	841	839	2	100 [99]
Other non-shockable	30	> 95	1590	1569	21	99 [98]	1631	1618	13	99 [98]
Fine VF	25	Report only	346	324	22	94 [89]	347	316	31	91 [86]
Other VT	25	Report only	38	38	0	100 [87]	51	49	2	96 [83]
Overall Performance										
			ADC-FR Development Dataset				ADC-FR Test Dataset			
			No Shock Advised	Shock Advised	(%) [90% low CI]		No Shock Advised	Shock Advised	(%) [90% low CI]	

Non-Shockable	2868	27	99% [99] (Specificity)	2925	17	99% [99] (Specificity)
Shockable	23	657	97% [94] (Sensitivity)	35	712	95% [93] (Sensitivity)

408 CI, 90% lower confidence interval; VF, ventricular fibrillation; VT, ventricular tachycardia; NSR, normal sinus rhythm.

409 **Table 3.** Overall reconfirmation analysis duration by the “Analysis During Compressions with
 410 Fast Reconfirmation” algorithm for each ECG rhythm

	Development Dataset			Test Dataset		
	1 ECG segment	2 ECG segments	3 ECG segments	1 ECG segment	2 ECG segments	3 ECG segments
Coarse VF	240 (87)	35 (13)	1 (0)	298 (87)	42 (12)	2 (1)
Rapid VT	46 (79)	11 (19)	1 (2)	44 (76)	10 (17)	4 (7)
NSR	323 (95)	18 (5)	0 (0)	391 (93)	26 (6)	2 (1)
PEA	1357 (85)	211 (13)	22 (1)	1393 (85)	221 (14)	17 (1)
Asystole	728 (79)	185 (20)	13 (1)	656 (78)	171 (20)	14 (2)
Fine VF	228 (58)	106 (27)	58 (15)	229 (66)	104 (30)	14 (4)
Other VT	29 (76)	8 (21)	1 (3)	35 (69)	14 (27)	2 (4)
Total	2951 (81)	574 (16)	96 (3)	3046 (83)	588 (16)	55 (1)

411 Data presented as n (%); VF, ventricular fibrillation; VT, ventricular tachycardia; NSR, normal
 412 sinus rhythm; PEA, pulseless electrical activity.

413 **Legends to figure**

414

415 **Figure 1.** The “Analysis During Compressions with Fast Reconfirmation” (ADC-FR)
416 technology can shorten chest compression pauses.

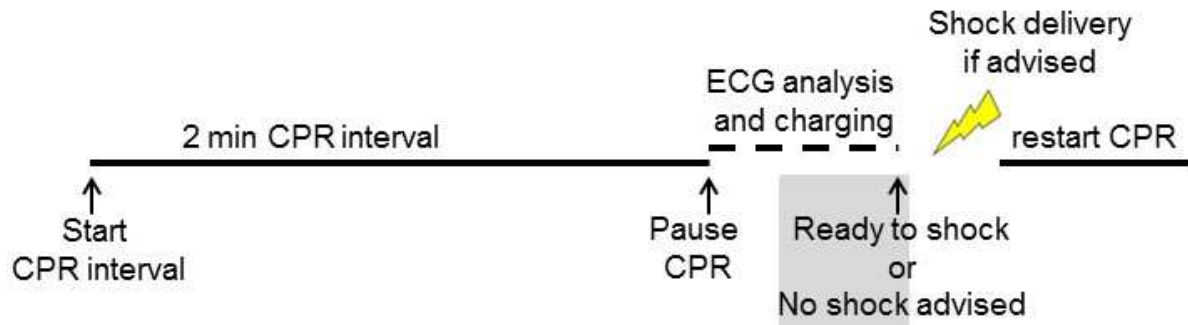
417 The standard automated external defibrillator (AED) algorithm requires interruption of
418 cardiopulmonary resuscitation (CPR) for ECG rhythm analysis and defibrillator charging (dash
419 line). In the ADC-FR algorithm, ECG rhythm analysis and defibrillator charging occur during
420 the 2-min CPR interval, requiring only a 3 sec pause for reconfirmation analysis prior to the
421 shock delivery (dotted line). The gray area represents the pause shortening when the ADC-FR is
422 used compared to the standard AED protocol. The ADC-FR algorithm does not apply to the first
423 AED application at the arrival to the patient, in which an immediate rhythm analysis is
424 performed and a shock delivery prompted if necessary, as recommended by guidelines.

425

426 **Figure 2.** Samples of raw ECG tracings for different rhythms generated by the automated
427 external defibrillator and correctly interpreted by the “analysis during chest compression with
428 fast reconfirmation ADC-FR” algorithm. The algorithm analyzes the ECG trace during chest
429 compression (CC) and then requires up to three 3-sec segments (seg.) of artefact-free ECG for
430 reconfirmation.

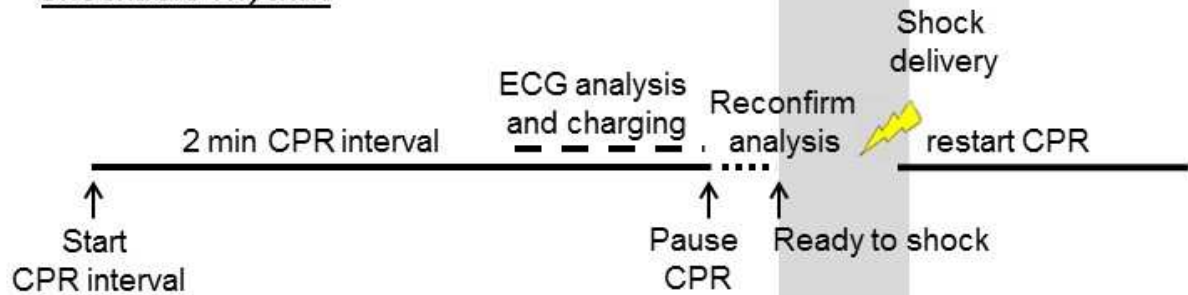
431 NSR, normal sinus rhythm; VF, ventricular fibrillation; VT, ventricular tachycardia.

Standard AED protocol

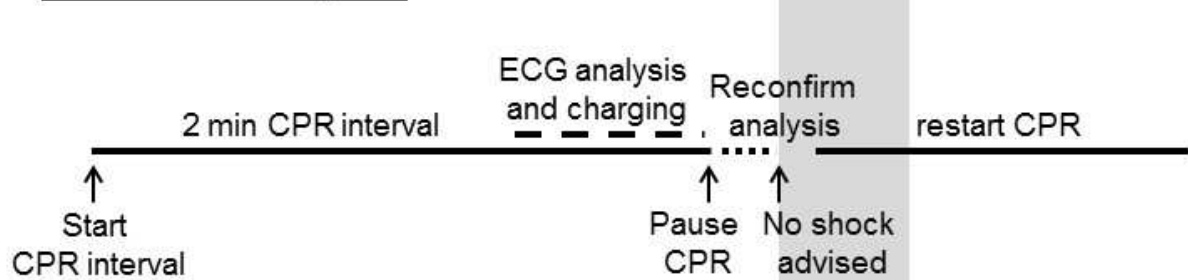


ADC-FR protocol

Shockable rhythm



Non-shockable rhythm



00:00

02:00

02:10

Time

