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17 Differential Associations of Specific Selective Serotonin Reuptake Inhibitors with Resting-

18 State Heart Rate and Heart Rate Variability: Implications for Health and Wellbeing

19

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56

57 Authors take responsibility for all aspects of the reliability and freedom from bias of the data  
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59

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61 wellbeing

62

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83

84

85 **Abstract:**

86 **Objective:** Debate has focused on the effects of the selective serotonin reuptake inhibitor  
87 (SSRI) antidepressants on heart rate (HR) and heart rate variability (HRV), both of which are  
88 predictors of adverse cardiovascular events. Here we examine the associations between  
89 specific SSRI antidepressants and resting state HR (and HRV) after accounting for a host of  
90 potential confounding factors.

91  
92 **Methods:** Participants included 10,466 not taking antidepressants, 46 participants taking  
93 escitalopram, 86 taking citalopram, 66 taking fluoxetine, 103 taking paroxetine, and 139  
94 taking sertraline. HR and HRV (RMSSD, HF) were extracted from 10-minute resting-state  
95 electrocardiograms. Analyses including propensity score weighting and matching were  
96 conducted using R-statistics to control for potentially confounding variables.

97  
98 **Results:** Major findings indicated that users of all SSRI medications – except fluoxetine –  
99 displayed lower HRV relative to non-users. Users of paroxetine also displayed significantly  
100 lower HRV relative to users of citalopram (Cohen's  $d = 0.42$ ), fluoxetine (Cohen's  $d = 0.54$ ) and  
101 sertraline (Cohen's  $d = 0.35$ ), but not escitalopram. While associations were also observed for  
102 HR these were less robust than those for HRV.

103  
104 **Conclusions:** While paroxetine is associated with decreases in HRV relative to non-users, as  
105 well as users of other SSRI medications, fluoxetine was the only medication not to display  
106 significant alterations in HR or HRV. These conclusions are limited by the cross-sectional  
107 design and non-randomized nature of medication prescriptions. Findings highlight the  
108 importance of focusing on specific medications, rather than more heterogeneous groupings  
109 according to antidepressant action, and may have implications for health and wellbeing over  
110 the longer term.

111  
112 **Abbreviations:** analysis of covariance (ANCOVA); analysis of variance (ANOVA); Anatomical  
113 Therapeutic Chemical Classification code (ATC); body mass index (BMI); Brazilian  
114 longitudinal study of adult health (ELSA-Brasil); Clinical Interview Schedule-Revised (CIS-R);  
115 coronary heart disease (CHD); electrocardiogram (ECG); heart rate (HR); heart rate variability  
116 (HRV); high frequency power (HF-HRV); Minnesota code (MC); propensity score matching  
117 (PSM); propensity score weighting (PSW); root mean square of successive squared

118 differences (RMSSD); serotonin and noradrenaline reuptake inhibitors (SNRIs); selective  
119 serotonin reuptake inhibitor (SSRI); tricyclic antidepressants (TCA's)  
120

## 121 **Introduction**

122 Antidepressant medications are a first-line treatment option for moderate to severe mood  
123 and anxiety disorders, yet some studies suggest that long-term use may be associated with an  
124 increased risk for cardiovascular disease [1-3]. We recently reported that use of tricyclic  
125 antidepressants (TCA's) is associated with a two-fold higher prevalence in coronary heart  
126 disease (CHD), relative to non-use in a cross-sectional analysis on the Brazilian longitudinal  
127 study of adult health (ELSA-Brasil) [4]. Although no associations were observed for the SSRI  
128 class, antidepressant use in Brazil is lower than in high-income countries. With the exception  
129 of sertraline and fluoxetine, SSRIs are not freely dispensed in public health pharmacies, as are  
130 tricyclics [5]. While TCA's are generally not recommended for depressed patients who have  
131 CHD [6], the effects of the selective serotonin reuptake inhibitor (SSRI) class of  
132 antidepressants remain unclear. In the present study, we examined the associations of  
133 specific medications in the SSRI class with resting state heart rate (HR) and heart rate  
134 variability (HRV), two psychophysiological indicators of health and wellbeing shown to  
135 predict future mortality [7]. The heart is under tonic inhibitory control by the  
136 parasympathetic (vagal) nervous system when at rest [8], and both HR and HRV under resting  
137 conditions may reflect vagally mediated cardiac activity. It is noted however, that HRV is a  
138 more specific measure of vagal activity [9, 10], while HR may also include sympathetic input.

139  
140 While the SSRIs are considered to be the safest class of antidepressant medications for use in  
141 cardiac patients [e.g. 11], they have also been reported to reduce HRV in depressed patients,  
142 compared to those not receiving an antidepressant, and to normal controls [12]. A variety of  
143 mechanisms have been proposed to contribute to the development of cardiovascular disease  
144 in users of antidepressants including SSRIs. These include increased HR, orthostatic  
145 hypotension, slowing of ventricular cardiac conduction, and antiarrhythmic activity [13].  
146 Another strong candidate for increased risk of cardiovascular disease is impairment in vagal



147 function [7]. Vagal function plays an important regulatory role over a variety of allostatic  
148 systems [14] and investigation of the associations between SSRIs, HR and HRV have  
149 important implications for the physical wellbeing of patients who use these medications over  
150 the long-term.

151

152 Use of antidepressants is associated with impairment in vagally mediated cardiac activity [15]  
153 [see also 16], associations that are most pronounced for the tricyclic antidepressants,  
154 followed by the serotonin and noradrenaline reuptake inhibitors (SNRIs) and the SSRIs. We  
155 recently observed that the SSRIs are associated with a small decrease in heart rate *and* HRV  
156 [15]. Consistent effects had been reported in a prior study [16], with decreases in HR  
157 interpreted as a decrease in sympathetic activity and decreases in HRV reflecting parallel  
158 decreases in cardiac vagal effects. SSRIs may interfere with the activation of fast Na<sup>+</sup> channels  
159 consistent with class I anti-arrhythmic agents, and calcium current, which reflects a negative  
160 inotropic effect reducing contractility [17, 18]. The HRV reductions associated with SSRI use  
161 have also been shown to be at least partly reversible, suggesting a possible causal effect [16].  
162 Although adverse effects of SSRIs have been reported [15, 16], these findings contradict other  
163 reports of increases [19] and no impact [20] on HRV. We have suggested previously [21] that  
164 one of the factors underpinning these contradictory findings may be the practice of grouping  
165 together heterogeneous medications within the SSRI class, leading to variable findings that  
166 depend on what SSRI medications are combined in a particular study. For instance, paroxetine  
167 displays six times more antimuscarinic (anticholinergic) potency than sertraline [22],  
168 highlighting the heterogeneity of these medications. This limitation of prior studies highlights  
169 the importance of comparing specific medications within the SSRI class. Other explanations  
170 for the reported contradictory findings are that studies have often not controlled for various  
171 confounding factors, which may impact on measures of vagal function. When studies *have*  
172 controlled for these factors, statistical analyses such as ANCOVA have often been employed,

173 which may lead to a phenomenon known as the “reversal paradox” – such that the  
174 relationship between two variables is reversed, diminished or enhanced when attempting to  
175 statistically control for a third variable – when studies do not randomly allocate participants  
176 to group [see 21 for discussion]. This makes it difficult to draw conclusions from prior studies  
177 that have employed this statistical approach.

178

179 For the first time, we compare multiple medications within the SSRI class, to determine and  
180 compare the impact of specific SSRI antidepressants on HR and HRV. Some of the limitations  
181 in prior studies were addressed using robust analytical techniques for controlling potential  
182 confounding factors. This approach has several advantages over traditional regression-based  
183 approaches, including improved control of confounding by not conflating propensity score  
184 methods with the modelling approach, and application of flexible machine learning methods  
185 to capture complex and nonlinear relationships between participant grouping and potential  
186 confounding variables without over-fitting the data [23].

187

## 188 **Methods**

### 189 *Participants*

190 ELSA-Brasil is a cohort of 15,105 civil servants aged 35-74 years enrolled between August  
191 2008 and December 2010 at 6 different sites in Brazil (Belo Horizonte, Porto Alegre, Rio de  
192 Janeiro, Salvador, Sao Paulo and Vitoria). The study design and sampling procedures of ELSA-  
193 Brasil have been reported previously [24, 25]. Briefly, eligible participants included males and  
194 females aged between 35 and 74 years who were active or retired employees of the six  
195 institutions. Exclusion criteria included severe cognitive or communication impairment,  
196 intention to quit working at the institution, and, if retired, residence outside the  
197 corresponding metropolitan area. Women with current or recent pregnancy were  
198 rescheduled so that the first interview could take place 4 months after delivery of their child

199 [24]. The ethics committees of the participating universities approved the research protocol.  
200 All participants provided written informed consent after a complete description of the study.  
201  
202 Here we report on a total of 10,906 participants after dropping participants on  
203 antidepressants other than an SSRI (n=382 including 113 on SNRIs, 174 on TCA's, and 96 on  
204 other antidepressants), participants on whom no HRV exam was available (n=1813, including  
205 504 participants with ectopic beats), participants on whom ECGs were not available for  
206 scoring major Q wave abnormalities (n=1740), and participants missing data on other  
207 variables used in analysis (n=563). Included participants comprised non-users of  
208 antidepressant medications (controls, n=10,466), those taking escitalopram (n=46),  
209 citalopram (n=86), fluoxetine (n=66), paroxetine (n=103) and sertraline (n=139). Participants  
210 on fluvoxamine were not included in the present study due to small numbers of participants  
211 taking this medication (n=3).

212

### 213 *Procedures*

214 Participants were asked to abstain from caffeine, alcohol and physical activity for at least 12  
215 hours before assessments. Participants were asked to bring all of the prescription and over-  
216 the-counter pill bottles to an interview for review by the interviewer. Individuals taking one  
217 selective serotonin reuptake inhibitor (SSRI) medication continuously over the past two  
218 weeks were classified as users, and grouped according to the specific antidepressant they  
219 were taking. Selective serotonin reuptake inhibitors were defined using the Anatomical  
220 Therapeutic Chemical (ATC) Classification code: N06AB. A continuous, 10-minute, resting-  
221 state ECG was also obtained from participants while in the supine position from which HR and  
222 HRV were extracted using standardised methods. [See also: 26, 27]. The electrocardiograms  
223 (ECGs) were always collected in the morning (8:00 to 12:00h) in a temperature-controlled  
224 room (21-24°C) and were sampled at 250 Hz with a digital electrocardiograph (Micromed,

225 Brazil) consistent with Task Force recommendations. ECGs were processed blindly at a  
226 Central ECG Reading Center, where they were visually inspected for technical errors and  
227 inadequate quality, and then stored for subsequent analysis in a Pyramis ECG management  
228 system (version 6.2.b, Cardiac Science Corporation, Bothel, WA, USA). ECGs were codified  
229 electronically using the Minnesota code manual of electrocardiographic findings by validated  
230 software, with manual over-reading by trained cardiologists to ensure quality control. Major  
231 Q wave abnormalities were determined from a 12-lead ECG as defined by the Minnesota code  
232 (MC) scheme (MC 1-1-X through to 1-2-X). Dedicated software (Micromed Wincardio 4.4a,  
233 Brazil) automatically generated the R-R interval series from the selected ECG lead with the  
234 highest R-wave amplitude (usually D2). Data were then processed to obtain measures of HR  
235 and HRV including the root mean square of successive squared differences (RMSSD) and high  
236 frequency power (HF-HRV). RMSSD and HF-HRV both reflect vagal parasympathetic activity  
237 and are usually highly correlated. HF-HRV (0.15–0.40 Hz) was estimated and expressed in  
238 absolute units. Both RMSSD and HF-HRV were then log-transformed as a normalisation  
239 strategy.

240

#### 241 *Covariates*

242 Covariates included sociodemographic factors (age; sex; level of education; race),  
243 cardiovascular risk factors (smoking; body mass index; hypertension; diabetes; and  
244 dyslipidemia), established heart disease and associated medications, physical inactivity and  
245 psychiatric morbidity. Level of education was entered as two dummy coded variables (less  
246 than high school: yes versus no; completed high school: yes versus no), while race was  
247 entered as a categorical variable indicating whether participants were non-White (yes versus  
248 no). Smoking status was indicated if participants were current smokers (current versus  
249 past/never) and body mass index (BMI) was determined as follows: weight in kilograms  
250 divided by height in meters squared. Hypertension was defined as a systolic blood pressure

251  $\geq 140$  mmHg, a diastolic blood pressure  $\geq 90$  mmHg, or use of antihypertensive medications.  
252 Diabetes was defined as self-reported or fasting blood glucose level  $\geq 126$  mg/dL, a 2-hour  
253 oral glucose tolerance test glucose level  $\geq 200$  mg/dL, or a glycated hemoglobin level  $\geq 6.5\%$ .  
254 Dyslipidemia was defined as an LDL cholesterol level  $\geq 130$  mg/dL or use of lipid-lowering  
255 medication. Blood samples were collected after a 12-hour overnight fast and medication use  
256 was determined on the basis of pill bottle review. Established heart disease was determined  
257 through a prior history of a physician-diagnosed myocardial infarction, a prior percutaneous  
258 coronary intervention including balloon angioplasty with or without stent placement, a prior  
259 surgical revascularization consisting of either arterial or venous grafts. Complementing this  
260 self-report information, major Q wave abnormalities (yes versus no) on the 12-lead ECG were  
261 also entered into analyses as a covariate. Physical activity was measured using the  
262 International Physical Activity Questionnaire [28] and categorized according to low activity  
263 versus moderate or high activity, as determined using scoring guidelines. Psychiatric  
264 morbidity was determined by trained interviewers using the Portuguese version [29] of the  
265 Clinical Interview Schedule-Revised (CIS-R) [30]. The CIS-R version was applied and severity  
266 scores were obtained ranging from zero to 57.

267

### 268 *Statistical Analysis*

269 Statistical analysis was conducted using IBM SPSS Statistics Version 21 and the R-statistical  
270 environment (version 3.0.1). Participant characteristics were examined using one-way  
271 analyses of variance (ANOVA) for contrasts involving continuous dependent measures, and  $\chi^2$   
272 statistics for categorical variables (Table 1). Tukey's HSD is reported for ANOVAs, correcting  
273 for multiple comparisons, while standardised residuals (z-scores) were used to help interpret  
274  $\chi^2$  statistics. Our main analyses involved comparison of SSRI antidepressant users and non-  
275 users on HR, RMSSD and HF-HRV, before and after application of propensity score techniques  
276 including propensity score weighting (PSW) [23] and propensity score matching (PSM) [31]

277 to adjust findings for the above covariates. These techniques involve calculating a single  
278 propensity score on the basis of entered covariates for each participant that relates to the  
279 probability that the participant belongs to the same distribution (i.e. antidepressant  
280 grouping). Two propensity analytic methods were employed: PSW and PSM. While PSW  
281 involves entering the propensity score into regression models, PSM involves selecting  
282 comparison participants (non-users of antidepressants) to match other groups on propensity  
283 scores. PSW was carried out using the 'twang' and 'survey' packages, while PSM was  
284 conducted using the 'MatchIt' package in the R statistical environment. Details on how to  
285 implement these procedures have been described previously [23, 31]. PSM was conducted as  
286 a sensitivity analysis, allowing the *effective sample size* of medication groupings to be  
287 increased and potential Type 1 error associated with discrepant sample sizes (i.e. between  
288 users and non-users of antidepressants) in PSW to be avoided. Effective sample sizes reflect  
289 the adverse impact of increased variance on precision and power [23], providing an estimate  
290 of the number of comparable participants in each group after introducing propensity score  
291 weights or dropping cases when matching. Additional PSW analyses were conducted after  
292 dropping control participants from analysis. This analysis allowed the effective sample size of  
293 medication groupings to be increased, achieving a higher-powered, head-to-head comparison  
294 between SSRI medications. Cohen's d effect size statistics were calculated for each pair-wise  
295 comparison with values of 0.2, 0.5, and 0.8 interpreted as small, medium, and large effects,  
296 respectively according Cohen's guidelines [32, 33]. Effect sizes were calculated using an  
297 online calculator (available here at The Campbell Collaboration: <http://goo.gl/zeLyuH>) [based  
298 on: 34].

299

## 300 **Results**

### 301 *Participant Characteristics*

302 Descriptive statistics for participants are reported in Table 1. Participants differed on age, sex,  
303 education, ethnicity, LDL cholesterol, and psychiatric morbidity, highlighting the importance  
304 of propensity score techniques to better control for the associations between these variables  
305 and HR, and HRV. It is possible for instance, that the differences in confounding variables may  
306 account for differences between groups on HR and HRV. The unadjusted results for HR and  
307 HRV are also reported in Table 1. Findings indicate that HR is reduced in users of citalopram,  
308 and that HRV is reduced in all users of SSRIs with the exception of those on fluoxetine. In the  
309 following sections, we report results after adjusting for potentially confounding variables on  
310 the basis of propensity scores. It is relevant to note here that the findings for fluoxetine did  
311 not change after controlling for confounding variables.

312

### 313 *Impact of SSRIs on Heart Rate and HRV: Propensity Score Analyses*

314 The differential impact of SSRIs was determined following PSW and PSM. Effective sample  
315 sizes after PSW were as follows: controls, n=10,451; escitalopram, n=28; citalopram=28;  
316 fluoxetine=48; paroxetine=62; sertraline=61. PSW analyses revealed significant alterations in  
317 RMSSD ( $F(5,10900)=10.66$ ,  $p<0.001$ ) and HF-HRV ( $F(5,10900)=8.06$ ,  $p<0.001$ ), while  
318 alterations were observed for HR at trend levels ( $F(5,10900)=1.91$ ,  $p=0.089$ ). Effective sample  
319 sizes after PSM were as follows: escitalopram, n=46; citalopram=86; fluoxetine=66;  
320 paroxetine=103; sertraline=139. Descriptive data, statistical tests and Cohen's  $d'$  effect size  
321 measures are summarised in Table 2. The major finding was that alterations in HR and HRV  
322 were observed for all users of SSRIs except for fluoxetine (light grey shaded cells). In addition,  
323 users of paroxetine displayed robust reductions in both measures of HRV (RMSSD, HF-HRV)  
324 in both PSW and PSM, relative to non-users, findings associated with small to moderate effect  
325 size (dark grey shaded cells in Table 2).

326

### 327 *Specificity Analyses*

328 Additional PSW analysis was conducted to compare each SSRI medication after dropping  
329 controls from the analyses. This allowed for the effective sample size of medication groupings  
330 to be increased and for a higher-powered, head-to-head comparison between SSRI  
331 medications to be conducted. After application of PSW, effective sample sizes were as follows:  
332 escitalopram,  $n=36$ ; citalopram= $73$ ; fluoxetine= $57$ ; paroxetine= $84$ ; sertraline= $122$ . Analyses  
333 revealed significant differences on HR ( $F(4,435)=2.52$ ,  $p=0.041$ ) and RMSSD ( $F(4,435)=2.99$ ,  
334  $p=0.019$ ), but not HF-HRV ( $F(4,435)=1.48$ ,  $p=0.21$ ). Post hoc tests indicated that users of  
335 paroxetine ( $M_{HR}=67.48$ ,  $SE=1.12$ ) and sertraline ( $M_{HR}=66.89$ ,  $SE=0.83$ ) displayed significantly  
336 higher HR than users of citalopram ( $M_{HR}=63.83$ ,  $SE=0.82$ ) ( $p=0.009$ , Cohen's  $d= 0.43$ ;  $p=0.009$ ,  
337 Cohen's  $d= 0.36$ ). Users of paroxetine ( $M_{RMSSD}=2.88$ ,  $SE=0.05$ ) also displayed significantly  
338 lower RMSSD than users of citalopram ( $M_{RMSSD}=3.08$ ,  $SE=0.06$ ;  $p=0.019$ , Cohen's  $d= 0.42$ ),  
339 fluoxetine ( $M_{RMSSD}=3.14$ ,  $SE=0.07$ ;  $p=0.003$ , Cohen's  $d = 0.54$ ) and sertraline ( $M_{RMSSD}=3.06$ ,  
340  $SE=0.05$ ;  $p=0.011$ , Cohen's  $d= 0.35$ ).

341

## 342 **Discussion**

343 This study examined and compared the impact of specific antidepressants within the SSRI  
344 class on resting-state HR and HRV. This is an important issue as chronic alterations of HR and  
345 HRV by SSRI antidepressants may lead to morbidity from a host of conditions and diseases,  
346 and mortality [7, 35]. Major findings from this study suggest that: 1) all users of SSRIs – except  
347 fluoxetine – display alterations in HR or HRV relative to non-users; findings for HRV appeared  
348 to be more robust and consistent for HRV, than those for HR, 2) users of citalopram display a  
349 mild bradycardia, characterised by reductions in HR by up to 4 beats per minute, findings  
350 associated with a small to moderate effect size, 3) only users of paroxetine display robust  
351 reductions in both measures of HRV, findings again associated with a small to moderate effect  
352 size, and 3) users of paroxetine also display small to moderate reductions on HRV relative to  
353 users of citalopram, fluoxetine and sertraline, but not escitalopram.



354

355 These associations may be produced through a variety of mechanisms including serotonergic  
356 receptors in brainstem regions involved in cardiovagal control, including the nucleus tractus  
357 solitarius at which cardiorespiratory afferent fibres terminate, and the cardiac vagal  
358 preganglionic neurones and rostral ventrolateral medulla (the location of sympathetic  
359 premotor neurones) [36, 37]. While different receptors appear to have variable effects, 5-HT<sub>1A</sub>  
360 and 5-HT<sub>7</sub> contribute to mild bradycardia, a finding that was observed here for users of  
361 citalopram, who displayed a reduction in resting state HR by approximately ~4 beats per  
362 minute. SSRIs may also inhibit cardiac and vascular Ca<sup>2+</sup>, Na<sup>+</sup> and K<sup>+</sup> channels further  
363 contributing to mild bradycardia [17, 18]. In addition, the effect of paroxetine may also be  
364 associated with anticholinergic effects [22], including inhibition of vagal efferent activity  
365 through blockade of muscarinic acetylcholine receptors at the sinoatrial node.

366

367 Large cohort studies and meta-analyses on the impact of SSRI medications have reported  
368 contradictory findings including increases [19], decreases [15, 16] and no alterations [20] in  
369 HRV, leading to much discussion in the literature [21, 38, 39]. These reports highlight the  
370 need for comparisons between different antidepressant medications from the SSRI class, an  
371 important contribution of the present study. Our findings extend our recent study on the  
372 impact of *antidepressant class* [15] to *specific antidepressants* within the SSRI class,  
373 demonstrating that paroxetine displays the most robust reductions in HRV after controlling  
374 for a number of confounding factors relative to controls, as well as users of other SSRI  
375 medications. These findings provide important new evidence for individual medications  
376 within the SSRI class, and highlight the need for further study in this area including  
377 investigation of the long-term effects of specific antidepressants within the SSRI class – and  
378 paroxetine in particular – on physical health and illness. To our knowledge, this is the first  
379 comparison of multiple medications within the SSRI class on measures of HR and HRV.

380

381 Paroxetine displayed the most pronounced reductions in HRV, relative to both controls as  
382 well as users of other medications, with the exception of escitalopram. Interestingly, a prior  
383 study on 28 inpatients with a DSM-IV diagnosed depressive episode also reported HRV  
384 reductions when paroxetine was prescribed at 40mg per day over a 35 day period [40]. The  
385 authors of this study [40] suggested that the higher dosage of paroxetine may impact on HRV  
386 in a similar way to tricyclic antidepressants. At higher concentrations (40 mg / day and  
387 higher), paroxetine is known to act as a dual serotonin/noradrenaline reuptake inhibitor [41,  
388 42] and is characterised by appreciable antimuscarinic (anticholinergic) potency [41]. It is  
389 interesting to note that our earlier study [15] reported that tricyclic antidepressants and  
390 SNRIs were associated with moderate to large increases in HR and decreases in HRV. By  
391 contrast, paroxetine in the present study is associated with small to moderate reductions in  
392 HRV. These effect sizes are presumably smaller than those we observed for tricyclic and SNRI  
393 medications, as some participants on paroxetine may have been prescribed dosages less than  
394 40mg/day. It is also notable that while SNRIs and tricyclic antidepressants may also lead to  
395 tachycardia in addition to reductions in HRV, users of paroxetine in the present study only  
396 exhibited decreases in HRV, not increases in HR (relative to non-users).

397

398 In contrast to paroxetine, fluoxetine was the only antidepressant that was not associated with  
399 significant alterations in cardiac activity. Fluoxetine is generally considered a safe medication  
400 for patients with cardiovascular disease. An early study [43] on depressed elderly patients  
401 with pre-existing cardiovascular disease reported that fluoxetine decreased HR by 6% (n=27),  
402 while nortriptyline, a tricyclic antidepressant, was associated with a 9% increase (n=52). This  
403 study highlighted the contrasting effects of fluoxetine versus nortriptyline on HR, with the  
404 authors concluding that increases in HR may reflect an increase in cardiac work, which over  
405 time may have clinically adverse effects. The only medication associated with robust

406 decreases in HR in the present study was citalopram (PSW Cohen's *d*: 0.35; PSM Cohen's *d*:  
407 0.49).

408

409 Although fluoxetine has been a popular pharmacological treatment for mood and anxiety  
410 disorders, recent systematic reviews indicate that other medications (e.g. sertraline,  
411 escitalopram) may be more efficacious [44]. In fact, sertraline is the most commonly studied  
412 SSRI medication in depressed patients with cardiovascular disease and is considered to be the  
413 first line drug of choice in this patient population [45]. In the present study, sertraline  
414 displayed some alterations of cardiac activity relative to non-users, however PSW findings  
415 were not confirmed using PSM. We have recently reported [46] that chronic treatment with  
416 sertraline (50-mg/d) does not impact on HRV over a period of 6-weeks in an independent  
417 cohort of patients with major depressive disorder.

418

419 Escitalopram also displayed alterations of cardiac activity, including moderate reductions in  
420 RMSSD in both PSW and PSM. Escitalopram was also the only SSRI medication that did not  
421 significantly differ from paroxetine in a direct comparison across multiple antidepressants  
422 highlighting the potentially adverse chronic effects of this medication on HRV. Recall that  
423 these findings were observed in users taking this medication continuously over the past two  
424 weeks. In contrast to these results, we have previously reported that a *single dose* of  
425 escitalopram (20mg), relative to placebo, is associated with *increases in HF-HRV*, findings  
426 associated with a moderate to large effect size [47]. It is possible therefore that acute versus  
427 chronic administration of escitalopram leads to different effects on HRV. Others have noted  
428 that chronic administration of SSRIs may lead to significant inhibition of various  
429 cardiovascular ion channels leading to certain pro- or arrhythmic effects [17, 18].

430

431 It is important to acknowledge here a variety of limitations associated with our study  
432 including a lack of additional information on participant's use of antidepressants including  
433 dose and length of use. It is possible for instance that the findings observed here are less than  
434 what might be observed for participants on higher dosage (e.g. paroxetine at dosages higher  
435 than 40mg / day) and extended use (e.g. years). While sensitivity analyses confirmed findings  
436 for paroxetine (robust reductions in HRV) and fluoxetine (no significant associations for HR  
437 or HRV), it is possible that factors including dosage and length of use may have contributed to  
438 the findings that could not be confirmed using sensitivity analyses especially HRV reductions  
439 for other medications including escitalopram, citalopram and sertraline. It is important  
440 however, to place these limitations in the context of various strengths of our study, including  
441 a comparison of multiple medications from the SSRI class on HR and HRV for the first time  
442 and use of propensity score techniques to better control for potentially confounding factors.

443

444 In summary, users of all SSRI medications – with the exception of fluoxetine – display  
445 alterations in cardiac activity, relative to non-users. Critically, users of paroxetine even  
446 display reductions in HRV relative to users of other SSRI medications, with the exception of  
447 escitalopram. These findings may have important clinical implications. First, HRV in particular  
448 reflects the functioning of the vagus nerve [48] and its impairment may lead to impaired  
449 regulation over various allostatic systems [14, 49], which may have adverse impacts on  
450 physical health in those patients who use these medications over the long-term. Second,  
451 patients with mood and anxiety disorders already display alterations in HR and HRV [15, 46,  
452 50, 51] and further reductions may have further consequences for patient health. Third,  
453 clinicians should be particularly mindful of physical health in patients treated with  
454 paroxetine, and also, possibly escitalopram, which are associated with the greatest impacts on  
455 HRV. Future research on the long-term effects of SSRI antidepressants, and the possibility that  
456 simple changes in health behaviours may ameliorate these associations, is needed.

458 **References**

- 459 1. Hamer M, David Batty G, Seldenrijk A, Kivimaki M (2011) Antidepressant medication  
460 use and future risk of cardiovascular disease: the Scottish Health Survey. *Eur Heart J*  
461 32:437–442. doi: 10.1093/eurheartj/ehq438
- 462 2. Smoller JW, Allison M, Cochrane BB, Curb JD, Perlis RH, Robinson JG, Rosal MC, Wenger  
463 NK, Wassertheil-Smoller S (2009) Antidepressant use and risk of incident  
464 cardiovascular morbidity and mortality among postmenopausal women in the women's  
465 health initiative study: Antidepressants and CVD risk after menopause. *Arch Intern Med*  
466 169:2128–2139. doi: 10.1001/archinternmed.2009.436
- 467 3. Whang W, Kubzansky LD, Kawachi I, Rexrode KM, Kroenke CH, Glynn RJ, Garan H,  
468 Albert CM (2009) Depression and risk of sudden cardiac death and coronary heart  
469 disease in women. *J Am Coll Cardiol* 53:950–958. doi: 10.1016/j.jacc.2008.10.060
- 470 4. Kemp AH, Brunoni AR, Bittencourt MS, Nunes MA, Benseñor IM, Lotufo PA (2015) The  
471 association between antidepressant medications and coronary heart disease in Brazil: A  
472 cross-sectional analysis on the Brazilian Longitudinal Study of Adult Health (ELSA-  
473 Brazil). *Front Public Health* 3:9. doi: 10.3389/fpubh.2015.00009
- 474 5. Brunoni AR, Nunes MA, Figueiredo R, Barreto SM, da Fonseca M de JM, Lotufo PA,  
475 Benseñor IM (2013) Patterns of benzodiazepine and antidepressant use among middle-  
476 aged adults. the Brazilian longitudinal study of adult health (ELSA-Brasil). *J Affect*  
477 *Disord* 151:71–77. doi: 10.1016/j.jad.2013.05.054
- 478 6. Colquhoun DM, Bunker SJ, Clarke DM, Glozier N, Hare DL, Hickie IB, Tatoulis J,  
479 Thompson DR, Tofler GH, Wilson A, Branagan MG (2013) Screening, referral and  
480 treatment for depression in patients with coronary heart disease. *Med J Aust* 198:483–  
481 484. doi: 10.5694/mja13.10153
- 482 7. Kemp AH, Quintana DS (2013) The relationship between mental and physical health:  
483 insights from the study of heart rate variability. *Int J Psychophysiol* 89:288–296. doi:  
484 10.1016/j.ijpsycho.2013.06.018
- 485 8. Thayer J, Hansen AL, Saus-Rose E, Johnsen BH (2009) Heart rate variability, prefrontal  
486 neural function, and cognitive performance: the neurovisceral integration perspective  
487 on self-regulation, adaptation, and health. *Ann Behav Med* 37:141–153. doi:  
488 10.1007/s12160-009-9101-z
- 489 9. Saul JP (1990) Beat-To-Beat Variations of Heart Rate Reflect Modulation of Cardiac  
490 Autonomic Outflow. *Physiology* 5:32–37.
- 491 10. Reyes Del Paso GA, Langewitz W, Mulder LJM, Roon A, Duschek S (2013) The utility of  
492 low frequency heart rate variability as an index of sympathetic cardiac tone: A review  
493 with emphasis on a reanalysis of previous studies. *Psychophysiology* 50:477–487. doi:  
494 10.1111/psyp.12027
- 495 11. Glassman AH, O'Connor CM, Califf RM, Swedberg K, Schwartz P, Bigger JT, Krishnan  
496 KRR, van Zyl LT, Swenson JR, Finkel MS, Landau C, Shapiro PA, Pepine CJ, Mardekian J,  
497 Harrison WM, Barton D, McIvor M, Sertraline Antidepressant Heart Attack Randomized

- 498 Trial SADHEART Group (2002) Sertraline treatment of major depression in patients  
499 with acute MI or unstable angina. *JAMA* 288:701–709.
- 500 12. Licht CMM, de Geus EJC, Zitman FG, Hoogendijk WJG, van Dyck R, Penninx BWJH (2008)  
501 Association between major depressive disorder and heart rate variability in the  
502 Netherlands Study of Depression and Anxiety (NESDA). *Arch Gen Psychiatry* 65:1358–  
503 1367. doi: 10.1001/archpsyc.65.12.1358
- 504 13. Paraskevaidis I, Palios J, Parissis J, Filippatos G, Anastasiou-Nana M (2012) Treating  
505 depression in coronary artery disease and chronic heart failure: what's new in using  
506 selective serotonin re-uptake inhibitors? *Cardiovasc Hematol Agents Med Chem*  
507 10:109–115.
- 508 14. Thayer J, Sternberg E (2006) Beyond heart rate variability: Vagal regulation of allostatic  
509 systems. *Ann N Y Acad Sci* 1088:361–372. doi: 10.1196/annals.1366.014
- 510 15. Kemp AH, Brunoni AR, Santos IS, Nunes MA, Dantas EM, Carvalho de Figueiredo R,  
511 Pereira AC, Ribeiro ALP, Mill JG, Andreão RV, Thayer J, Benseñor IM, Lotufo PA (2014)  
512 Effects of depression, anxiety, comorbidity, and antidepressants on resting-state heart  
513 rate and its variability: an ELSA-Brasil cohort baseline study. *American Journal of*  
514 *Psychiatry* 171:1328–1334. doi: 10.1176/appi.ajp.2014.13121605
- 515 16. Licht CMM, de Geus EJC, van Dyck R, Penninx BWJH (2010) Longitudinal evidence for  
516 unfavorable effects of antidepressants on heart rate variability. *Biol Psychiatry* 68:861–  
517 868. doi: 10.1016/j.biopsych.2010.06.032
- 518 17. Pacher P, Kecskemeti V (2004) Cardiovascular side effects of new antidepressants and  
519 antipsychotics: new drugs, old concerns? *Curr Pharm Des* 10:2463–2475.
- 520 18. Pacher P, Ungvari Z, Nanasi PP, Furst S (1999) Speculations on difference between  
521 tricyclic and selective serotonin reuptake inhibitor antidepressants on their cardiac  
522 effects. Is there any? *Current medicinal chemistry*
- 523 19. van Zyl LT, Hasegawa T, Nagata K (2008) Effects of antidepressant treatment on heart  
524 rate variability in major depression: a quantitative review. *BioPsychoSocial Med* 2:12.  
525 doi: 10.1186/1751-0759-2-12
- 526 20. Kemp AH, Quintana DS, Gray MA, Felmingham KL, Brown K, Gatt JM (2010) Impact of  
527 depression and antidepressant treatment on heart rate variability: A review and meta-  
528 analysis. *Biol Psychiatry* 67:1067–1074. doi: 10.1016/j.biopsych.2009.12.012
- 529 21. Kemp AH, Quintana DS, Malhi GS (2011) Effects of serotonin reuptake inhibitors on  
530 heart rate variability: methodological issues, medical comorbidity, and clinical  
531 relevance. *Biol Psychiatry* 69:e25–6– author reply e27–8. doi:  
532 10.1016/j.biopsych.2010.10.035
- 533 22. Richelson E (1996) Synaptic effects of antidepressants. *J Clin Psychopharmacol* 16:1S–  
534 7S.
- 535 23. McCaffrey D, Griffin BA, Almirall D, Slaughter ME, Ramchand R, Burgette LF (2013) A  
536 tutorial on propensity score estimation for multiple treatments using generalized  
537 boosted models. *Stat Med* 32:3388–3414. doi: 10.1002/sim.5753

- 538 24. Schmidt MI, Duncan BB, Mill JG, Lotufo PA, Chor D, Barreto SM, Aquino EM, Passos VMA,  
539 Matos SM, Molina MDCB, Carvalho MS, Benseñor IM (2014) Cohort profile: Longitudinal  
540 Study of Adult Health (ELSA-Brasil). *International journal of epidemiology*. doi:  
541 10.1093/ije/dyu027
- 542 25. Aquino EML, Barreto SM, Benseñor IM, Carvalho MS, Chor D, Duncan BB, Lotufo PA, Mill  
543 JG, Molina MDC, Mota ELA, Azeredo Passos VM, Schmidt MI, Szklo M (2012) Brazilian  
544 Longitudinal Study of Adult Health (ELSA-Brasil): Objectives and design. *Am J*  
545 *Epidemiol* 175:315–324. doi: 10.1093/aje/kwr294
- 546 26. Mill JG, Pinto K, Griep RH, Goulart A (2013) Medical assessments and measurements in  
547 ELSA-Brasil. *Rev Saúde Pública* 47:54–62. doi: 10.1590/S0034-8910.2013047003851
- 548 27. Ribeiro AL, Lotufo PA, Pereira SV, Bergmann K, Ladeira RM, Oliveira RA, Mill JG, Barreto  
549 SM (2013) Challenges to implementation of the ECG reading center in ELSA-Brazil. *Rev*  
550 *Saude Pública*
- 551 28. Craig CL, Marshall AL, Sjöström M, Bauman AE, Booth ML, Ainsworth BE, Pratt M,  
552 Ekelund U, Yngve A, Sallis JF, Oja P (2003) International physical activity questionnaire:  
553 12-country reliability and validity. *Medicine and science in sports and exercise*  
554 35:1381–1395. doi: 10.1249/01.MSS.0000078924.61453.FB
- 555 29. Nunes MA, de Mello Alves MG, Chor D, Schmidt MI, Duncan BB (2012) Adaptação  
556 transcultural do CIS-R (Clinical Interview Schedule- Revised Version) para o português  
557 no Estudo Longitudinal De Saúde Do Adulto (ELSA). *Revista HCPA* 31:487–490.
- 558 30. Lewis G, Pelosi AJ, Araya R, Dunn G (1992) Measuring psychiatric disorder in the  
559 community: a standardized assessment for use by lay interviewers. *Psychol Med*  
560 22:465–486.
- 561 31. Ho DE, Imai K, King G, Stuart EA (2011) MatchIt: Nonparametric preprocessing for  
562 parametric causal inference. *Journal of Statistical Software* 42:1–28.
- 563 32. Cohen J (1988) *Statistical power analysis for the behavioral sciences*. Lawrence  
564 Erlbaum Associates, Hillsdale, New Jersey
- 565 33. Cohen J (1992) A power primer. *Psychol Bull* 112:155–159.
- 566 34. Lipsey MW, Wilson DB (2001) *Practical meta-analysis*. Sage Publications, Inc
- 567 35. Thayer J, Yamamoto SS, Brosschot JF (2010) The relationship of autonomic imbalance,  
568 heart rate variability and cardiovascular disease risk factors. *Int J Cardiol* 141:122–131.  
569 doi: 10.1016/j.ijcard.2009.09.543
- 570 36. Jordan D (2005) Vagal control of the heart: central serotonergic (5-HT) mechanisms.  
571 *Experimental Physiology* 90:175–181. doi: 10.1113/expphysiol.2004.029058
- 572 37. Chang JS, Ha K, Yoon I-Y, Yoo CS, Yi SH, Her JY, Ha TH, Park T (2012) Patterns of  
573 cardiorespiratory coordination in young women with recurrent major depressive  
574 disorder treated with escitalopram or venlafaxine. *Prog Neuropsychopharmacol Biol*  
575 *Psychiatry* 1–7. doi: 10.1016/j.pnpbp.2012.06.002
- 576 38. Licht CMM, Penninx BWJH, Geus EJC de (2011) Reply to: Effects of serotonin reuptake

- 577 inhibitors on heart rate variability: Methodological issues, medical comorbidity, and  
578 clinical relevance. BPS 1–2. doi: 10.1016/j.biopsycho.2010.12.039
- 579 39. Brunoni AR, Lotufo PA, Benseñor IM (2012) Are antidepressants good for the soul but  
580 bad for the matter? Using noninvasive brain stimulation to detangle  
581 depression/antidepressants effects on heart rate variability and cardiovascular risk.  
582 *Biol Psychiatry* 71:e27–8– author reply e29–30. doi: 10.1016/j.biopsycho.2011.08.026
- 583 40. Lederbogen F, Gernoth C, Weber B, Colla M, Kniest A, Heuser I, Deuschle M (2001)  
584 Antidepressive treatment with amitriptyline and paroxetine: comparable effects on  
585 heart rate variability. *Journal of Clinical Psychopharmacology* 21:238–239.
- 586 41. Richelson E (2003) Interactions of antidepressants with neurotransmitter transporters  
587 and receptors and their clinical relevance. *J Clin Psychiatry* 64 Suppl 13:5–12.
- 588 42. Sanchez C, Reines EH, Montgomery SA (2014) A comparative review of escitalopram,  
589 paroxetine, and sertraline: Are they all alike? *International clinical*  
590 *psychopharmacology* 29:185–196. doi: 10.1097/YIC.000000000000023
- 591 43. Roose SP, Glassman AH, Attia E, Woodring S, Giardina E-GV, Bigger JT Jr. (1998)  
592 Cardiovascular Effects of Fluoxetine in Depressed Patients With Heart Disease. *Am J*  
593 *Psychiatry* 155:660–665. doi: 10.1176/ajp.155.5.660
- 594 44. Magni LR, Purgato M, Gastaldon C, Papola D, Furukawa TA, Cipriani A, Barbui C (2013)  
595 Fluoxetine versus other types of pharmacotherapy for depression. *Cochrane Database*  
596 *Syst Rev* 7:CD004185.
- 597 45. Mavrides N, Nemeroff C (2013) Treatment of depression in cardiovascular disease. *Depress*  
598 *Anxiety* 30:328–341. doi: 10.1002/da.22051
- 599 46. Brunoni AR, Kemp AH, Dantas EM, Goulart AC, Nunes MA, Boggio PS, Mill JG, Lotufo PA,  
600 Fregni F, Benseñor IM (2013) Heart rate variability is a trait marker of major  
601 depressive disorder: evidence from the sertraline vs. electric current therapy to treat  
602 depression clinical study. *Int J Neuropsychopharm* 16:1937–1949. doi:  
603 10.1017/S1461145713000497
- 604 47. Kemp AH, Outhred T, Saunders S, Brunoni AR, Nathan PJ, Malhi GS (2014) Impact of  
605 escitalopram on vagally mediated cardiovascular function in healthy participants:  
606 implications for understanding differential age-related, treatment emergent effects.  
607 *Psychopharmacology* 231:2281–2290. doi: 10.1007/s00213-013-3374-4
- 608 48. Cacioppo JT, Tassinary LG, Berntson G (2007) *Handbook of psychophysiology*.  
609 Cambridge University Press
- 610 49. Tracey KJ (2002) The inflammatory reflex. *Nature* 420:853–859.
- 611 50. Chalmers JA, Quintana DS, Abbott MJ-A, Kemp AH (2014) Anxiety disorders are  
612 associated with reduced heart rate variability: A meta-analysis. *Front Psychiatry* 5:80.  
613 doi: 10.3389/fpsy.2014.00080
- 614 51. Kemp AH, Quintana DS, Quinn CR, Hopkinson P, Harris AWF (2014) Major depressive  
615 disorder with melancholia displays robust alterations in resting state heart rate and its  
616 variability: implications for future morbidity and mortality. *Frontiers in Psychology*



617 5:1387. doi: 10.3389/fpsyg.2014.01387

618

Table 1: Participant characteristics (unadjusted) including means (M) and standard errors (SE) for controls and treatment participants (n) and percentage of participants relative to sample size of sub-group (%).

	<b>Controls (n=10,466)</b>	<b>Escitalopram (n=46)</b>	<b>Citalopram (n=86)</b>	<b>Fluoxetine (n=66)</b>	<b>Paroxetine (n=103)</b>
<b>Age, M (SE)</b>	52.10 (0.09)	54.48 (1.37)	55.95 (1.02)*	53.86 (1.18)	53.63 (0.92)
<b>Females, n (%)</b>	5,515 (52.7)	32 (69.6)	67 (77.9)*	55 (83.3)*	78 (75.7)*
<b>Education, n (%)</b>					
<b>Less than High School</b>	1386 (13.2)	0*	3 (3.5)*	8 (12.1)	5 (4.9)*
<b>High School</b>	3620 (34.6)	4 (8.7)*	31 (36.0)	25 (37.9)	26 (25.2)
<b>Ethnicity, n (%)</b>					
<b>Non-White</b>	5,154 (49.2)	15 (32.6)	28 (32.6)*	22 (33.3)	37 (35.9)
<b>Current Smokers, n (%)</b>	1,330 (12.7)	6 (13.0)	9 (10.5)	12 (18.2)	16 (15.5)
<b>Body Mass Index, M (SE)</b>	27.03 (0.05)	27.27 (0.78)	26.60 (0.51)	27.31 (0.55)	25.98 (0.42)
<b>Hypertension, n (%)</b>	3,773 (36.1)	16 (34.8)	32 (37.2)	21 (31.8)	34 (33.0)
<b>Diabetes, n (%)</b>	2,063 (19.7)	7 (15.2)	13 (15.1)	9 (13.6)	12 (11.7)
<b>Dyslipidemia, n (%)</b>	5,985 (57.2)	33 (71.7)	60 (69.8)	44 (66.7)	73 (70.9)
<b>Hard CHD, n (%)</b>	474 (4.5)	2 (4.3)	1 (1.2)	1 (1.5)	9 (8.7)
<b>Major Q-Waves, n (%)</b>	263 (2.5)	1 (2.2)	1 (1.2)	0	2 (1.9)

<b>Physical Inactivity, n (%)</b>	8,037 (76.8)	40 (87.0)	61 (70.9)	52 (78.8)	80 (77.7)
<b>CIS-R Total Score<sup>2</sup>, M (SE)</b>	7.87 (0.08)	9.15 (1.08)	12.26 (1.03)*	10.12 (1.10)	11.29 (1.04)*
<b>Heart Rate, M (SE)</b>	66.78 (0.09)	65.55 (1.19)	63.72 (0.80)*	65.64 (0.94)	66.53 (0.95)
<b>RMSSD, M (SE)</b>	3.23 (0.01)	2.98 (0.08)*	3.08 (0.06)	3.17 (0.07)	2.94 (0.05)*
<b>HF-HRV, M (SE)</b>	5.36 (0.01)	4.83 (0.18)*	4.97 (0.12)*	5.19 (0.13)	4.84 (0.11)*

<sup>1</sup> CIS-R Total Score: Severity of psychiatric morbidity determined using Clinical Interview Schedule-Revised

\*Refers to one-way ANOVA in which each group is compared to controls (Tukey's HSD, p<0.05) or from  $\chi^2$  statistics lying outside  $\pm 1.96$  reflecting a significance value of p < 0.05 using Fisher's exact test where necessary

**Table 2.** Effects of multiple selective serotonin reuptake inhibitors (SSRIs) on heart rate and heart rate variability using propensity score weighting (PSW) and propensity score matching (PSM).

Propensity Score Weighting (using ‘twang’ and ‘survey’ packages)																		
	Non-users (n=10,451)		Escitalopram (n=28)			Citalopram (n=28)			Fluoxetine (n=48)			Paroxetine (n=62)						
	M	SE	M	SE	Cohen’s <i>d</i>	PSW <i>p</i>	M	SE	Cohen’s <i>d</i>	PSW <i>p</i>	M	SE	Cohen’s <i>d</i>	PSW <i>p</i> <sup>3</sup>	M	SE	Cohen’s <i>d</i>	PSW <i>p</i>
<b>HR</b>	66.81	0.09	66.50	1.77	-0.03	0.005	63.55	1.58	-0.35	<0.001	64.75	0.93	-0.22	0.68	67.25	1.16	0.05	<0.001
<b>RMSSD</b>	3.23	0.01	2.98	0.09	-0.24	0.001	2.98	0.08	-0.24	0.001	3.20	0.07	-0.03	0.60	2.89	0.06	-0.33	<0.001
<b>HF-HRV</b>	5.36	0.01	4.91	0.17	-0.44	0.009	4.78	0.17	-0.57	0.001	5.28	0.15	-0.08	0.60	4.78	0.18	-0.57	<0.001
Propensity Score Matching (using MatchIt package) <sup>1</sup>																		
	Non-users (matched) <sup>2</sup>		Escitalopram (n=46)			Citalopram (n=86)			Fluoxetine (n=66)			Paroxetine (n=103)						
	M	SE	M	SE	Cohen’s <i>d</i>	PSM <i>p</i>	M	SE	Cohen’s <i>d</i>	PSM <i>p</i>	M	SE	Cohen’s <i>d</i>	PSM <i>p</i> <sup>3</sup>	M	SE	Cohen’s <i>d</i>	PSM <i>p</i>
<b>HR</b>	67.75	0.99	65.55	1.19	-0.07	0.74	63.72	0.80	-0.49	0.002	65.64	0.94	-0.30	0.09	66.53	0.95	-0.03	0.8
<b>RMSSD</b>	3.15	0.05	2.98	0.08	-0.46	0.038	3.08	0.06	-0.14	0.42	3.17	0.07	0.11	0.55	2.94	0.05	-0.50	<0.001
<b>HF-HRV</b>	5.16	0.12	4.82	0.18	-0.34	0.11	4.97	0.12	-0.17	0.28	5.19	0.14	0.04	0.82	4.84	0.11	-0.34	0.0

Notes: <sup>1</sup> PSM was conducted to confirm findings from PSW to enhance effective sample size and combat potential for Type 1 error resulting from unequal sample sizes (e.g., non-users vs. antidepressant groups). <sup>2</sup> Values for non-users from bipartite matching was used, and control participants were matched to each antidepressant group, values differ for each comparison. For brevity the values for non-users are not shown here. Note however, that the Cohen’s *d* effect size measure for each comparison was calculated using the values from controls that were

antidepressant being compared.<sup>3</sup> Light grey shaded cells reflect correspond to p-values for fluoxetine, the only antidepressant not associated with alterations in HRV. Findings for users of paroxetine in which robust reductions in both measures of HRV (RMSSD, HF-HRV) on both PSW and PSM were observed, relative to non-users, are of moderate effect size.