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## **Exploring factors associated with patients' adherence to antihypertensive drugs among people with primary hypertension in the United Kingdom**

**Short title:** "Adherence and patients' characteristics"

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1 **Abstract (248 words)**

2 **Objective**

3 To explore factors associated with adherence to antihypertensive drugs overall  
4 (“therapy adherence”) and to particular classes (“class adherence”) in hypertensive  
5 patients.

6  
7 **Methods**

8 This retrospective cohort study included adults with primary hypertension identified in  
9 the UK Clinical Practice Research Datalink from April/2006 to March/2013. Individuals  
10 were followed from the date of first-ever antihypertensive drug class (class adherence)  
11 prescribed and from the date of the first-ever antihypertensive drug (therapy  
12 adherence) issued to the earliest of study end, patient leaving the database or death.  
13 Prescribing episodes (time from a drug class being first prescribed to the end of follow-  
14 up time) of six antihypertensive drug classes were recorded. Proportion of Days  
15 Covered (PDC) was used to estimate therapeutic adherence for a patient’s  
16 antihypertensive drugs therapy during follow-up period and class adherence of a  
17 specific antihypertensive class in each episode, respectively. Generalized linear  
18 modelling was used to examine factors associated with PDC.

19  
20 **Results**

21 Median therapy and class PDC were 93.9% and 98.3% in the 176,835 patients and  
22 371,605 prescribing episodes; 20% and 38.4% of the patients and episodes had  
23 PDC<80%, respectively. Higher therapy and class PDC was associated with increasing  
24 age, using renin angiotensin drugs and being pre-existing patient and user of  
25 antihypertensive drugs. Higher deprivation, multiple comorbidities and switching of  
26 antihypertensive drugs were associated with lower PDC.

27

## 28 **Conclusions**

29 Several patient factors were confirmed as determinant of adherence to  
 30 antihypertensive drug classes and therapy; hence they can assist in identifying patients  
 31 at risks of non-adherence; thus targeting them for adherence improving interventions.

32

## 33 **Keywords**

34 Adherence; Antihypertensive drugs; Clinical Practice Research Datalink; Generalized  
 35 Linear model

36

## 37 **List of Abbreviations**

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95%CI	95% Confidence Interval
ACEIs	Angiotensin-Converting Enzyme Inhibitors
ARBs	Angiotensin Receptor Blockers
BBs	Beta-Blockers
BP	Blood Pressure
CCBs	Calcium-Channel Blockers
CCI	Charlson Comorbidity Index
CPRD	Clinical Practice Research Datalink
CVDs	Cardiovascular Diseases
GLM	Generalised Linear Model
HT	Hypertension
IQR	Interquartile Range
MPR	Medication Possession Ratio
OLS	Ordinary Least Square
PDC	Proportion of Days Covered
SD	Standard Deviation
SES	Socio-Economic Status
UK	United Kingdom

---

## 38 **Introduction**

39 Hypertension (HT) is a highly prevalent condition in the United Kingdom (UK) with an  
40 estimated prevalence of 13.7% [1]. Antihypertensive drugs have been shown to reduce the  
41 risk of cardiovascular complications, premature mortality [2] and achieve cost-savings [3] in  
42 people with HT. Nevertheless, suboptimal control of BP has been consistently reported in  
43 population-based surveys of hypertension management worldwide [4, 5] .

44

45 Patients' poor adherence to antihypertensive drugs is considered one of the key contributing  
46 factors to suboptimal BP control [6]. Long-term adherence to antihypertensive therapy is  
47 crucial to achieve and maintain optimal BP control [7]. Reported adherence to  
48 antihypertensive drugs varies from 28% to 78% [8, 9], attributed mostly to differences in  
49 study populations, types of medications being considered, study designs, follow-up time, and  
50 definitions and measurement of adherence.

51

52 Poor adherence to antihypertensive drugs is associated with increased cardiovascular  
53 events and hospitalisations with subsequently high costs and healthcare resources utilisation  
54 [10, 11]. In England, the estimated potential cost of the health gains foregone as a result of  
55 non-adherence to antihypertensive drugs is about £390 million per annum [12]. It was also  
56 estimated that over £100 million per annum would be saved if 80% of people with  
57 hypertension were adherent to their medications [12].

58

59 Understanding factors associated with adherence is crucial for patients, and healthcare  
60 professionals and providers. Previous studies have found associations between adherence  
61 to antihypertensive drug therapy and factors such as patients' age, gender, comorbidity and  
62 type of antihypertensive drug class [8, 9], however, the joint impact of these factors have not  
63 been evaluated together in a single cohort.

64

65 Most studies assessed adherence to antihypertensive drug classes in patients with newly-  
66 diagnosed hypertension [13]. Patients with pre-existing hypertension are expected to have  
67 different medication-taking behaviours compared with newly-diagnosed patients [13]. The  
68 impact of switching from one antihypertensive drug class to another on a patient's  
69 adherence to overall antihypertensive drug therapy (therapy adherence) as well as to a  
70 particular antihypertensive drugs class (class adherence) has not been widely studied [8, 9].  
71 Assessing adherence to individual antihypertensive drug classes without considering  
72 adherence to overall antihypertensive drug therapy limits the applicability of research  
73 findings from most previous studies as the majority of hypertensive patients are prescribed  
74 more than one antihypertensive drug class for their BP control [2]. Many studies have  
75 transformed adherence into a binary variable, using a cut-off point of 80%. Furthermore, a  
76 simple binary measure for adherence [8, 14] assumes patients over a wide range of  
77 adherence values (PDC 0-80%) to have same medication-taking behaviour, and thus may  
78 potentially misclassify/misjudge a patient's adherence behaviour.

79

80 These factors limit the application and generalisability of previous study results to patient  
81 medicine-taking behaviour in real practice. To add to what is known about adherence in  
82 hypertension, this study assessed the association between patient characteristics and  
83 adherence to both overall antihypertensive therapy and individual drug classes by applying a  
84 robust analytical method to analyse adherence as a continuous variable in patients with both  
85 newly diagnosed and pre-existing primary hypertension as an approach to produce more  
86 accurate and generalisable findings.

87

## 88 **Methods**

### 89 **Study design and data source**

90 This retrospective cohort study used data from the UK Clinical Practice Research Datalink  
91 (CPRD) database [15] from April 2006 to March 2013, as it was the most updated date for  
92 the availability of CPRD data at the time of the study. CPRD is a primary care database

93 containing longitudinal electronic clinical data of more than 13.7 million patients including  
94 information about patients' demographics, medical conditions, diagnoses, prescribed  
95 medications, vaccination and laboratory tests. By March 2015, CPRD included 5.4 million  
96 active patients from 685 primary care practices across the UK [16]; it covers about 8.5% of  
97 the UK population and is considered to be broadly representative in terms of patient and  
98 practice characteristics [17]. This study protocol was approved by the Independent Scientific  
99 Advisory Committee of CPRD database (protocol number 13\_150).

100

### 101 **Study cohort**

102 Adults ( $\geq 18$  years old) with a diagnosis of primary hypertension and at least two  
103 antihypertensive drug prescriptions after the diagnosis date during the study period were  
104 included in this study. Included patient needed to have at least one year of CPRD records  
105 before and after the date of their first-ever antihypertensive drug prescription (index date)  
106 during the study period. Sporadic users who were prescribed only one antihypertensive  
107 prescription were excluded [13]. In order to ensure that treating hypertension is at least one  
108 of the potential indications of the prescribed antihypertensive drugs, participants were  
109 required to have their antihypertensive drugs prescribed on or after their hypertension  
110 diagnosis date.

111

112 Patients with history of cardiovascular diseases (CVDs) prior to the index date were  
113 excluded because the presence of CVDs may affect the choice of antihypertensive drugs  
114 (indication bias) and patients' medication-taking behaviours (i.e., higher adherence as they  
115 are more willing to follow medical instructions) [18, 19]. Patients who were initiated on  
116 multiple antihypertensive drugs (either as fixed-dose combination or multiple pills) on the  
117 index date were also excluded as it was not possible to assign patients into a particular  
118 antihypertensive drug class which in turn conflicted with the study's objective of measuring a  
119 patient's adherence to any antihypertensive drug therapy. Indeed, these patients have often  
120 been excluded from previous adherence studies as they were reported to be at high risk of



121 HT-related complications, having higher BP value and hence would have different  
122 medication-taking behaviours [2, 20, 21].

123

#### 124 **Measurement of adherence**

125 Individuals in the cohort were followed from the index date to the earliest of: study end date,  
126 patient transferred out of the dataset (e.g. left the practice), or patient's death; during this  
127 period, all antihypertensive prescriptions issued were retrieved and the duration of each  
128 prescription was calculated. Antihypertensive drugs were further divided into six classes:  
129 angiotensin-converting enzyme inhibitors (ACEIs), calcium-channel blockers (CCBs),  
130 diuretics, angiotensin receptor blockers (ARBs), beta-blockers (BBs), and "Others" (including  
131 vasodilators, centrally acting drugs, alpha-blockers).

132

133 A commonly used adherence measure [22], *Proportion of Days Covered* (PDC), was used  
134 as a 'proxy' for adherence in this study, and both antihypertensive 'therapy adherence' and  
135 'class adherence' were measured. Individual patients' adherence to any antihypertensive  
136 drug therapy (PDC for therapy adherence) during the study period was calculated by dividing  
137 the '*total number of days covered with any antihypertensive drug*' by the '*number of days in*  
138 *the follow-up period*' [22]. Likewise, adherence to a specific antihypertensive drug class  
139 (PDC for class adherence) in each prescribing episode of a class was calculated by dividing  
140 the '*total number of days covered with an antihypertensive class*' by the '*number of days in a*  
141 *prescribing episode of that class*'.

142

143 The prescribing episode for a class was the duration when a patient was consecutively  
144 prescribed with the same antihypertensive drug class, starting from the date of a patient's  
145 first-ever prescription of the class during study period to the final date covered by the  
146 antihypertensive class. Multiple episodes can be identified in one patient's follow-up period,  
147 as patients may discontinue or switch to other drug classes.

148

**149 Study covariates**

150 Baseline characteristics of patients, including patients' demographics (age, gender, and  
151 socioeconomic status), disease status (comorbidity, hypertension status) and their drug use  
152 status (type of antihypertensive drug class, antihypertensive drug use status [pre-existing or  
153 new users]) on the index date were included as covariates that may be associated with  
154 patients' adherence. Furthermore, whether patients have been switched from their  
155 antihypertensive drug class was also included as a study covariate. To account for the  
156 variations in patients' follow up time, which resulted from differences in patients' study entry  
157 and exit dates, individual patient's follow up time was included as a covariate in the  
158 regression model both as a continuous and as a categorical variable.

159

160 Townsend deprivation score [23] ranging from one to five (one being least deprived and five  
161 most deprived) was used a proxy for individual patients' socioeconomic status (SES).  
162 Individual's comorbidity status was measured by the Charlson comorbidity index (CCI) [24].  
163 Hypertension status, i.e. pre-existing (prevalent) or newly-diagnosed (incident) hypertension  
164 was judged by whether a patient had any hypertension-related diagnosis codes in the year  
165 prior to the first hypertension diagnosis code identified during the study period [25].

166

167 Similarly, antihypertensive drug use status, i.e. pre-existing (prevalent) or new (incident)  
168 users of a specific antihypertensive class was judged by whether any antihypertensive class  
169 was issued in the year prior to the index prescription date identified during the episode.  
170 Switching was defined as stopping the initial antihypertensive class and starting another  
171 class.

172

**173 Data analysis**

174 Descriptive statistics were used to describe patient-related factors at baseline. Mean with  
175 standard deviation (SD) and median with interquartile range (IQR) were used to present  
176 normally and non-normally distributed continuous variables, respectively; proportion was

177 used to present categorical variables. The association between individual patient  
178 characteristics and the non-normally distributed PDC was first tested in non-parametric  
179 univariate analyses, including Spearman's rank correlation test for continuous variables  
180 (age, follow-up time), Wilcoxon rank sum (Mann-Whitney) and Kruskal-Wallis tests for binary  
181 and categorical variables.

182

183 The influence of all study covariates on therapy and class adherence was assessed by using  
184 two generalised linear models (GLM) with gamma family and log link function, with the  
185 dependent variable as 'PDC for antihypertensive drug therapy' and 'PDC for each episode of  
186 antihypertensive classes', respectively. The results were presented as regression  
187 coefficients and 95% confidence interval (95%CI). The models' goodness of fit, in terms of  
188 the appropriateness of the chosen family and link function, was checked using the modified  
189 Park test [26] and Pregibon Link test [27].

190

## 191 **Results**

### 192 **Baseline characteristics**

193 Overall, 176,835 adults with primary hypertension were included in this study with 371,605  
194 prescribing episodes of the six antihypertensive classes identified during the follow-up  
195 period. The mean age of patients at baseline was  $60.8 \pm 13.6$  years, 55.6% (n=98,320) were  
196 female, 53.4% (n=94,430) were newly diagnosed hypertensive patients and 51.0%  
197 (n=90,186) were new users of antihypertensive drugs. The median follow-up duration was  
198 5.3 (IQR: 3.1, 6.5) years. Of the 371,605 prescribing episodes, the most commonly  
199 prescribed class episodes were ACEIs (29.7%) and CCBs (25.1%). Patients' characteristics  
200 and drug use status were significantly different across the episodes of six antihypertensive  
201 classes (Table 1).

202

### 203 **Proportion of days covered**

204 Both individual patients' PDCs for antihypertensive drug therapy overall (Figure 1) and PDCs  
205 for antihypertensive class in each episode (Figure 2) were not normally distributed. Although  
206 the median PDC was 93.3% (IQR: 47.3%, 100%) and 98.3% (IQR: 86.5%, 100%) for  
207 therapy and class adherence, respectively; 20.0% of patients' therapeutic adherence and  
208 38.4% of prescribing episodes' class adherence were suboptimal (PDC<80%). Mean  
209 therapy and class adherence was  $87\% \pm 22.2$  and  $73\% \pm 33.8$ , respectively.

210

### 211 **Univariate analyses of factors influencing adherence**

212 The univariate analyses demonstrated that all the covariates were significantly associated  
213 with PDC for therapy adherence in the study cohort and with PDC for class adherence in  
214 each episode (Table 2). Patients who were initiating antihypertensive therapy on CCBs had  
215 the highest PDC for therapy adherence (median: 98.6%, IQR: 86.5%, 100%). On the other  
216 hand, the median PDC for class adherence in the prescribing episodes of ARBs (median:  
217 97.4%, IQR: 74.2%, 100%) was the highest amongst all antihypertensive drug classes,  
218 followed by ACEIs (median: 95.7%, IQR: 51.3%, 100%).

219 Both higher therapy and class PDCs were associated with increasing age, lower deprivation,  
220 prevalent drug users, and higher comorbidity index ( $CCI \geq 2$ ). Male gender, being pre-existing  
221 hypertensive patient were associated with higher PDCs in the episodes of antihypertensive  
222 classes but lower PDCs for patients' overall therapy adherence. Switching between  
223 antihypertensive drug classes was also associated with lower PDC for therapy adherence.

224

### 225 **Multivariate analyses of factors influencing adherence**

226 The results from the GLM analysis indicated that all the patient characteristics were  
227 independent factors for both patients' adherence to antihypertensive therapy and to a  
228 specific drug class in each episode (Table 3).

229

230 Being female, having pre-existing hypertension, previous utilisation of antihypertensive  
231 medicines, and older age were associated with higher PDC for patients' antihypertensive  
232 therapy; on the other hand, higher deprivation index, high comorbidity scores ( $CCI \geq 2$ ), and  
233 switching of antihypertensive drug class were associated with lower PDC of patients'  
234 antihypertensive therapy. Patients who were initiated with ACEIs and ARB as the index drug  
235 class during the study period; their PDCs for therapy adherence significantly increased by  
236 4% and 3% ( $p < 0.001$ ), respectively.

237

238 Similarly, pre-existing hypertension, pre-existing antihypertensive drug user, and older age  
239 were also associated with a higher PDC for class adherence in each episode; on the other  
240 hand, being female, higher deprivation index, and high comorbidity scores ( $CCI \geq 2$ ) were  
241 associated with lower a PDC for class adherence. Comparing between different  
242 antihypertensive drug classes, the highest PDC was in the episodes started from ARBs  
243 (13%,  $p < 0.001$ ), followed by ACEIs (8%,  $p < 0.001$ ); and the PDC for class categories  
244 "Others" was the lowest (11%,  $p < 0.001$ ).

245

246 Both class and therapy PDCs significantly changed over patients' follow-up time. There was  
247 a significant declining trend in class PDC across follow-up time categories with an average  
248 decline of 1.4% for each year increase in follow up time. Whereas for therapy PDC, although  
249 there was an average increase of patients' adherence to any antihypertensive therapy by  
250 0.7% for each year increase in follow up time, the effect across follow up time categories  
251 were different.

252

253 The fitted multivariate GLM regression models can predict both the mean therapy and class  
254 PDCs for any patient with a particular set of characteristics included in the model. For  
255 instance, the predicted mean PDC of diuretics in the episodes for a 50-year old, female  
256 patient, with a deprivation index of 2 and comorbidity score  $\geq 2$ , having diuretics as the index  
257 antihypertensive drug class, being a new antihypertensive drug user, having pre-existing  
258 hypertension, and four years of follow-up time, is 67.7% (95%CI: 66.8%, 69.5%).

## 259 Discussion

### 260 Main findings

261 This study assessed adherence to both individual antihypertensive drug classes and overall  
262 antihypertensive therapy using longitudinal data over a seven-year period. To our  
263 knowledge, this study is the only study that has collectively analysed adherence, as a  
264 continuous variable, to both antihypertensive drug classes and overall therapy over a long  
265 period in a population of both new and existing hypertensive patient; thus providing  
266 generalisable findings by overcoming the aforementioned limitations of the previous studies.  
267 Although no similar studies were found for direct comparison of the study findings, the  
268 findings were compared with results from various studies. Overall findings are not dissimilar  
269 to these earlier studies, but now we can more confidently describe adherence behaviours in  
270 both incident and prevalent populations and better understand the relationship between  
271 individual drug class and overall therapy adherence.

272

273 The median PDC at first glance may appear generally high, but the other summary  
274 measures, despite their limitations, (such as mean and proportion with PDC<80%)  
275 demonstrated a sub-optimal PDC level that is comparable with other adherence studies [22].  
276 The overall mean PDC for antihypertensive drug class in each prescribing episode was 73%  
277 and about 40% episodes had PDC<80%. Although these results are comparable with the  
278 mean class adherence of 67% and PDC<80% of 36% reported by a systematic review of  
279 139 observational studies of adherence to antihypertensive drug classes [28], the follow-up  
280 time over which adherence was measured in the systematic review was only one year which  
281 provided limited insights into the dynamic nature of adherence beyond one year. However,  
282 this current study examined adherence over seven years and has provided deeper  
283 understanding of patients' behaviours in taking their antihypertensive medications. Class  
284 adherence declined steadily, unlike therapy adherence that showed a different pattern  
285 consisting of significant reduction in the early course of therapy (>2-3 years), followed by  
286 insignificant change (>3-5 years) then a significant increase afterward (>5 years).

287

288 Furthermore, a recent observational study, assessing association between patients'  
289 characteristics and adherence to overall antihypertensive drug therapy, also reported a  
290 similar high, overall median adherence to overall antihypertensive drug therapy of 96%, with  
291 more than 75% being considered adherent (PDC $\geq$ 80%) [29], however, again this study was  
292 limited by short follow-up of one year as well as analysing adherence as a binary variable.

293

## 294 **Factors associated with patients' adherence**

### 295 **Medications and clinically related factors**

296 Type of antihypertensive drug class was a significant predictor for adherence to both  
297 antihypertensive drug classes and therapy; ARBs followed by ACEIs were associated with  
298 the highest-class adherence, while diuretics and BBs were associated with the lowest. This  
299 confirms the historical findings from many other adherence studies [21, 30, 31], which has  
300 been attributed largely to the more favourable tolerability profile of ARBs and ACEIs  
301 compared with other antihypertensive drug classes. However, once switching was  
302 considered in measuring adherence to overall antihypertensive drug therapy, ACEIs rather  
303 than ARBs had the highest adherence, with BBs no longer having lower adherence  
304 compared with diuretics. This implies that all the previous historical findings were indeed  
305 biased by not considering switching in measuring adherence, especially given the better  
306 tolerability of ARBs compared with others [31], and hence less switching and better  
307 adherence profile of ARBs if switching was not considered.

308

309 Lower adherence to antihypertensive drug classes and therapy was observed in newly  
310 diagnosed hypertensive patients and new antihypertensive drug users. Differences in  
311 beliefs, perceptions and attitudes towards hypertension and antihypertensive drug therapy  
312 between incident and prevalent patients could explain the observed disparity in adherence  
313 behaviour between these two groups of patients since prevalent patients may have passed  
314 the stages of lack of belief in the necessity of treating hypertension [32]. Furthermore,



315 patients' concerns and fears about antihypertensive drugs' adverse effects in the early  
316 stages of treatment in the case of incident patients may act as a barrier of adherence to  
317 antihypertensive drugs, particularly when patients' hold the belief that a drug's side effects  
318 outweigh any potential future benefits [33].

319

320 This study found a negative association between adherence to antihypertensive drug  
321 classes/therapy and presence of comorbidities. It has been reported that patients with no  
322 comorbidity were 29% more likely to be adherent compared with those with a high  
323 comorbidity score [34]. The negative association between high comorbidity and adherence  
324 could be partly explained by comorbidity-related polypharmacy, as additional medications  
325 are needed in response to increasing comorbid conditions [35], which has been found to  
326 decrease adherence [36]. Importantly, it appears that this has to exceed a limit before  
327 comorbidities having any negative impact of adherence, as it is evident by the fact that both  
328 class and therapy adherence were decreasing only for patients with high comorbidity score  
329 (CCI $\geq$ 2).

330

331 In previous studies [7, 37], switching between antihypertensive drug classes was associated  
332 with lower adherence to any antihypertensive drug therapy. This association could be related  
333 to many switching-related concerns that would potentially decrease patients' adherence,  
334 such as changes in product packaging and tablet appearance [38] and taste [35], differences  
335 in adherence profiles of the various antihypertensive drug classes [8], and impairing patient's  
336 confidence in drug therapy [39]. Furthermore, it has been shown that patients' concerns  
337 about switching may produce a nocebo effect (i.e. patients' negative perceptions may cause  
338 negative outcomes) [40].

339

#### 340 **Demographic factors**

341 Patients' demographics, such as age, gender and SES, were also significant predictors for  
342 antihypertensive drug adherence. Poor SES has been recognised by the WHO as one of the

343 potential factors for patients' non-adherence to antihypertensive drugs [41]. An American  
344 cohort study has found that increasing in patients' income quintile, as a proxy for SES, was  
345 associated with a 10% increase in the proportion of adherent patients (OR: 1.10, 95%CI:  
346 1.08, 1.12) [34]. Furthermore, a recent retrospective cohort study, which included more than  
347 30,000 adult patients, assessed the association between patients' characteristics and  
348 medication adherence across eight diseases, including hypertension, and found a higher  
349 adherence level in those living in higher SES (lower deprivation) [29].

350

351 Females, in general, have been consistently shown to be less adherent to antihypertensive  
352 drug classes [29, 42, 43]. Although similar finding was observed in the current study,  
353 importantly this was not the case for adherence to antihypertensive drug therapy as females  
354 had higher adherence than males. This could be explained by not allowing/considering  
355 switching in measuring adherence to antihypertensive drug classes, especially giving the  
356 higher switching rates in females [44]; i.e., once patients have been switched to another  
357 antihypertensive drug class they were considered as non-adherent to the initial drug class by  
358 definition as they have stopped taking it, but obviously patients have been adherent to the  
359 antihypertensive drug therapy overall as they continued to take the new drug class while  
360 stopped the initial class. This demonstrates how insights into patients' medication taken  
361 behaviours could be biased by purely measuring adherence to antihypertensive drug classes  
362 without considering the overall antihypertensive drug therapy, which is more influential on  
363 controlling BP.

364

### 365 **Strengths and limitations**

366 One of the major strengths of the current study is analysing adherence as a continue  
367 measure by applying an advanced statistical technique (GLM) unlike most of the previous  
368 studies [8, 9] which measured and analysed adherence as a binary variable using a non-  
369 empirical, arbitrary cut-off point of 80% [8, 14]. Dichotomisation of adherence simplifies  
370 statistical analysis, presentation and interpretation of results [45] but incurs several

371 disadvantages. Dichotomisation of a continuous variable is often associated with loss of  
372 information [46] that can lead to loss of both estimation efficiency and power in hypothesis  
373 testing [45, 47] due to a reduction in the number of degrees of freedom [48]. Furthermore,  
374 although the 80% cut-off point for optimal adherence has been generally used and linked  
375 with clinical outcomes in previous studies, the optimal adherence cut-off point may be higher  
376 than 80%, as BP has found to continuously reduce with increases in adherence from 80% to  
377 100% [49].

378

379 Therefore, the International Society for Pharmaceutical and Outcomes Research [48] has  
380 recommended against converting continuous adherence data into binary data. On a related  
381 notes, previous studies [50, 51] that analysed adherence as a continuous measure have  
382 used inappropriate statistical methods to perform the analysis such as ordinary least square  
383 (OLS) regression. OLS is considered an inappropriate method because it requires a  
384 normally distributed outcome variable that is almost violated by the skewed distribution of the  
385 continuous adherence measure.

386

387 Another main strength of this study lies in measuring adherence to both antihypertensive  
388 drug classes and any antihypertensive drug therapy using a large population dataset of both  
389 incident and prevalent hypertensive patients over a long period. Furthermore, applying an  
390 advanced statistical technique (GLM) to analyse the association between adherence (as a  
391 continuous variable) with a wide range of patient related factors. This approach has not been  
392 observed in previous adherence studies and rendered the findings more generalisable to the  
393 wider hypertensive population. For instance, measuring adherence to both antihypertensive  
394 drug classes and any antihypertensive drug therapy has increased the applicability of the  
395 study findings to the real-world management of hypertension, given the increased proportion  
396 of hypertensive patients who are prescribed more than one antihypertensive drug classes to  
397 control their BP [2].

398

399 Furthermore, failure to allow for switching in measuring class adherence in previous studies,  
400 implies that the patient failed to take the drug as recommended [9], which, in fact, may not  
401 be the case because patient's switching is often recommended by physicians in response to  
402 treatment failure or side effects [13]. Therefore, measuring adherence to any  
403 antihypertensive drug therapy (therapy adherence), in this current study, helped to avoid  
404 misunderstanding of patients' medication-taking behaviours toward a particular  
405 antihypertensive drug class and provided more insights.

406

407 Additionally, the model generated from using GLM method in this study could potentially be  
408 applied as a predication tool for identifying patients at risks of poor adherence who could  
409 possibly then be targeted for adherence improving interventions; however, this requires  
410 further validation and evaluation research.

411

412 However, a number of limitations need to be acknowledged. Although a wide range of  
413 demographics and clinically related factors were considered in this study, bias due to  
414 unmeasured confounders, such as dosing history, cannot be ruled out due to the  
415 retrospective nature of the study design. Although some of the antihypertensive drugs could  
416 be used to treat other conditions alongside hypertension, the criterion of antihypertensive  
417 drugs' prescription date always being on or after the hypertension diagnosis date has  
418 ensured that treating hypertension was at least one of the drug's potential indications.

419

420 In addition, the CPRD contains only prescribed data, therefore adherence was measured  
421 indirectly by PDC as a proxy, which may lead to further overestimation of medication  
422 adherence. Furthermore, overestimation of adherence might have resulted also from  
423 excluding patients on multiple therapies at the index date as they might have higher risk of  
424 poor adherence.

425

426 Another limitation, which applies to any secondary database analysis, includes measuring  
427 adherence using secondary databases. This has been validated with other methods of  
428 adherence measurement such as electronic devices, patients' self-reports and pill counts  
429 [52, 53], and no substantial differences between dispensing and prescribing datasets were  
430 found [54]. Given the different methods to measure medication adherence using secondary  
431 databases, it could be argued that each method may produce different results. However,  
432 Hess *et al* (2006) [55] in their comparison of the various methods of measuring adherence  
433 using secondary databases found that all the methods provide comparable values.

434

435 Among the adherence measures, medication possession ratio (MPR) and PDC were the  
436 best predictors of patients' hospitalisations [56]. PDC is considered preferable than MPR as  
437 it provides more conservative estimates of adherence, especially in the presence of  
438 therapeutic switching or concurrent drug therapy [57, 58], even though adherence alone  
439 does not provide information on whether patients benefit from the increased use of  
440 medicines.

441

## 442 **Conclusions**

443 Overall, adherence to antihypertensive medications was suboptimal among patients with  
444 primary hypertension. A set of patient-level factors has been identified as potential  
445 determinants for patients' adherence to antihypertensive drugs that would potentially assist  
446 to identify patients at risk of poor adherence. Subsequently, those patients can be targeted  
447 for adherence improving interventions and/or more intensive follow-up by healthcare  
448 professionals to improve their adherence level.

449

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## Tables

**Table 1. Patient characteristics at first-ever antihypertensive drug class episodes**

Covariates	ACEIs	CCBs	Diuretics	BBs	ARBs	“Others”	Total
<b>Number of episodes (%)</b>	110,493 (29.7)	93,119 (25.1)	71,883 (19.3)	42,164 (11.4)	39,862 (10.7)	14,084 (3.8)	371,605
<b>Mean age (±SD) years</b>	57.9±12.4	64.8±12.0	67.4±12.1	61.5±13.1	62.2±12.6	64.7±14.1	62.2±12.9
<b>Gender (%)<sup>a</sup></b>							
Male	61,655 (55.8)	46,839 (50.3)	23,865 (33.2)	17,709 (42.0)	18,695 (46.9)	6,648 (47.2)	177,627 (47.8)
Female	48,838 (44.2)	46,280 (49.7)	48,018 (66.8)	24,455 (58.0)	21,167 (53.1)	7,436 (52.8)	193,978 (52.2)
<b>Townsend deprivation score (quintile) (%)<sup>b</sup></b>							
1 (Least deprived)	28,176 (25.5)	22,535 (24.2)	17,252 (24.0)	11,131 (26.4)	11,042 (24.5)	3,451 (24.5)	97,361 (26.2)
2	26,739 (24.2)	22,162 (23.8)	17,539 (24.4)	10,372 (24.6)	10,045 (25.2)	3,324 (23.6)	86,212 (23.2)
3	22,983 (20.8)	18,996 (20.4)	15,455 (21.5)	8,728 (20.7)	81,72 (20.5)	3,042 (21.6)	73,578 (19.8)
4	19,889 (18.0)	17,227 (18.5)	13,298 (18.5)	7,547 (17.9)	6,418 (16.1)	2,577 (18.3)	63,544 (17.1)
5 (Most deprived)	12,707 (11.5)	12,292 (13.2)	8,338 (11.6)	4,385 (10.4)	4,186 (10.5)	1,648 (11.7)	50,910 (13.7)
<b>Median follow up time (IQR, years)<sup>b</sup></b>	4.6 (2.9, 6.4)	4.4 (2.6, 6.4)	5.6 (3.6, 6.9)	6.8 (4.6, 7.0)	6.5 (4.4, 6.9)	6.5 (4.2, 6.9)	5.1 (3.2, 6.8)
<b>CCI (%)<sup>a</sup></b>							
0	65,412 (59.2)	58,292 (62.6)	44,208 (61.5)	30,400 (72.1)	22,522 (56.5)	7,535 (53.5)	229,280 (61.7)
1	24,529 (22.0)	18,251 (19.6)	14,520 (20.2)	6,072 (14.4)	9,288 (23.3)	3,451 (24.5)	75,807 (20.4)
≥2	20,552 (18.6)	16,575 (17.8)	13,155 (18.3)	5,692 (13.5)	8,052 (20.2)	3,098 (22.0)	66,517 (17.9)
<b>Hypertension status (%)<sup>a</sup></b>							
Incident cases	62,650 (56.7)	51,960 (55.8)	26,165 (36.4)	7,252 (17.2)	6,458 (16.2)	2,225 (15.8)	165,364 (44.5)
Prevalent cases	47,843 (43.3)	41,159 (44.2)	45,718 (63.6)	34,912 (82.8)	33,404 (83.8)	11,859 (84.2)	206,241 (55.5)
<b>Drug use status (%)<sup>a</sup></b>							
Incident users	72,925 (66.0)	60,993 (65.5)	29,184 (40.6)	9,698 (21.0)	9,328 (23.4)	3,606 (25.6)	193,978 (52.2)
Prevalent users	37,568 (34.0)	32,126 (34.5)	42,699 (59.4)	33,310 (79.0)	30,534 (76.6)	10,478 (74.4)	177,627 (47.8)

**(Note)** <sup>a</sup> p<0.001 from McNemar test; <sup>b</sup> p<0.001 Kruskal-Wallis test; **IQR**: interquartile range; **CCI**: Charlson comorbidity index; **ACEIs**: angiotensin converting enzyme inhibitors; **ARBs**: angiotensin receptor blockers; **CCBs**: calcium channel blockers; **BBs**: beta-blockers

**Table 2. Univariate analysis of the patient related factors with class and therapy PDC**

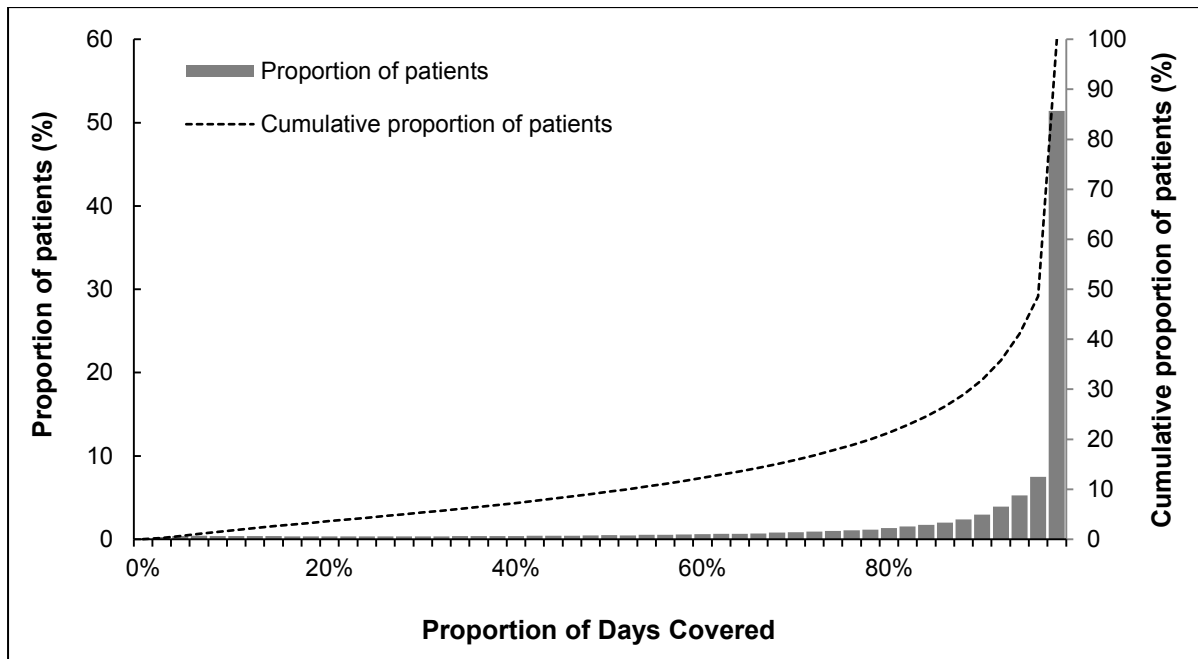
Covariates	Class PDC		Therapy PDC		Statistical test
	Median (IQR)	p-value	Median (IQR)	p-value	
<b>Index drug class</b>					
ACEIs	95.7 (51.3, 100)		98.3 (85.8, 100)		
CCBs	94.3 (50.4, 100)		98.6 (86.5, 100)		
Diuretics	90.6 (44.2, 100)	P=0.001	97.8 (85.9, 100)	P=0.0001	Kruskal-Wallis test
BBs	86.7 (24.3, 100)		98.5 (88.0, 100)		
ARBs	97.4 (74.2, 100)		98.3 (88.7, 100)		
“Others”	84.4 (26.7, 100)		98.5 (86.8, 100)		
<b>Gender</b>					
Male	94.2 (51.1, 100)	P<0.001	97.9 (85.7, 100)	P<0.001	Wilcoxon rank sum (Mann-Whitney) test
Female	93.6 (43.7, 100)		98.7 (87.2, 100)		
<b>Townsend deprivation score (quintile)</b>					
1 (Least deprived)	95.3 (48.3, 100)		98.8 (89.4, 100)		
2	94.6 (47.8, 100)		98.6 (88.2)		Kruskal-Wallis test
3	94.2 (47.7, 100)	P=0.001	98.3 (86.8, 100)	P=0.0001	
4	93.0 (46.9, 100)		97.9 (84.1, 100)		
5 (Most deprived)	89.6 (45.1, 100)		96.6 (78.1, 100)		
<b>Drug use status</b>					
Incident users	92.7 (37.9, 100)	P<0.001	98.2 (82.7, 100)	P<0.001	Wilcoxon rank sum (Mann-Whitney) test
Prevalent users	94.9 (57.0, 100)		98.4 (89.4, 100)		
<b>Hypertension status</b>					
Incident cases	93.3 (39.3, 100)	P<0.001	98.5 (84.0, 100)	P= 0.0079	Wilcoxon rank sum (Mann-Whitney) test
Prevalent cases	94.3 (53.1, 100)		98.2 (88.0, 100)		
<b>Switching index drug</b>					
No	NA	NA	98.7 (88.9, 100)	P<0.001	Wilcoxon rank sum (Mann-Whitney) test
Yes			96.7 (78.4, 100)		
<b>CCI</b>					
0	93.7 (45.5, 100)		98.1 (85.8, 100)		
1	94.3 (50.6, 100)	P=0.001	98.5 (87.1, 100)	P=0.0001	Kruskal-Wallis test
≥2	94.3 (49.1, 100)		98.9 (87.7, 100)		
<b>Age (years)</b>	0.08*	P<0.001	0.15*	P<0.001	Spearman's rank correlation test
<b>Follow up time (years)</b>	0.03*	P<0.001	0.02*	P<0.001	Spearman's rank correlation test

**(Note)** PDC: proportion days covered; IQR: interquartile range ; ACEIs: angiotensin converting enzyme inhibitors; ARBs: angiotensin receptor blockers; CCBs: calcium channel blockers; BBs: beta-blockers; CCI: Charlson comorbidity index; NA: not applicable

**Table 3. Results from the GLM regression of the patient related factors with class and therapy adherence**

Covariates	Class PDC		Therapy PDC	
	Coefficients (95%CI)	p-value	Coefficients (95%CI)	p-value
<b>Index drug class</b>				
Diuretics	1.0		1.0	
ACEIs	0.08 (0.074, 0.087)	<0.001	0.04 (0.035, 0.043)	<0.001
CCBs	0.052 (0.04, 0.06)	<0.001	0.02 (0.017, 0.025)	<0.001
BBs	-0.09 (-0.10, 0.-0.084)	<0.001	0.016 (0.011, 0.020)	<0.001
ARBs	0.13 (0.12, 0.14)	<0.001	0.03 (0.023, 0.032)	<0.001
“Others”	-0.11(-0.13, -0.09)	<0.001	-0.008 (-0.011, 0.0096)	0.869
<b>Gender</b>				
Male	1.0		1.0	
Female	-0.034 (-0.38, -0.029)	<0.001	0.004 (0.0012, 0.0060)	<0.001
<b>Townsend deprivation score (quintile)</b>				
1 (Least deprived)	1.0		1.0	
2	-0.003 (-0.009, 0.002)	0.294	-0.009 (-0.012, -0.01)	<0.001
3	-0.002 (-0.009, 0.004)	0.452	-0.013 (-0.02, 0.01)	<0.001
4	-0.008 (-0.012, -0.001)	0.022	-0.025 (-0.03, -0.02)	<0.001
5 (Most deprived)	-0.02 (-0.03, -0.01)	<0.001	-0.05 (-0.06, -0.04)	<0.001
<b>Drug use status</b>				
Incident users	1.0		1.0	
Prevalent users	0.13 (0.12, 0.14)	<0.001	0.06 (0.055, 0.065)	<0.001
<b>Hypertension status</b>				
Incident cases	1.0		1.0	
Prevalent cases	0.02 (0.008, 0.025)	0.02	0.03 (0.028, 0.04)	<0.001
<b>CCI</b>				
0	1.0		1.0	
1	0.0006 (-0.0049, 0.006)	0.837	0.03 (-0.00002, 0.0059)	0.052
≥2	-0.02 (-0.021, -0.0092)	<0.001	-0.046 (-0.078, -0.0020)	0.004
<b>Age (years)</b>	0.003 (0.0028, 0.0033)	<0.001	0.0032 (0.0032, 0.0033)	<0.001
<b>Follow up time (years)</b>	-0.014 ( -0.016, -0.013)	<0.001	0.007 (0.006, 0.0073)	<0.001
<b>Follow up time categories (years)</b>				
≤2	1.0		1.0	
>2-3	-0.017 (-0.022, -0.012)	<0.001	-0.08 (-0.14, -0.03)	0.001
>3-4	-0.024 (-0.030, -0.019)	<0.001	-0.04 (-0.09, 0.01)	0.148
>4-5	-0.035 (-0.040, -0.030)	<0.001	-0.02 (-0.07, 0.03)	0.756
>5	-0.035 (-0.039, -0.030)	<0.001	0.18 (0.21, 0.14)	<0.001
<b>Switching index drug</b>				
No	NA	NA	1.0	
Yes			-0.043 (-0.046, -0.040)	<0.001

**(Note)** PDC: proportion days covered; **ACEIs**: angiotensin converting enzyme inhibitors; **ARBs**: angiotensin receptor blockers; **CCBs**: calcium channel blockers; **BBs**: beta-blockers; **CCI**: Charlson comorbidity index; **NA**: not applicable

**Figures****Figure 1** Cumulative proportion of patients' adherence to any antihypertensive drug therapy

**Figure 2** Cumulative proportion of adherence of the episodes of the six antihypertensive drug classes

