




# Comparing the Sensitivities of Measures of Adherence to Antihypertensive Drugs Using Korean National Health Insurance Claims Data

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**Purpose:** Numerous studies have utilized various forms of adherence measures. However, methods for measuring adherence are inconsistent. Moreover, few studies are available that have compared sensitivities of the effects of several criteria on medication adherence. This study aims to compare measures of adherence using varied analytical decisions.

**Materials and Methods:** We included three measures for adherence: proportion of days covered with one or more medications ( $PDC_{with \geq 1}$ ), duration weighted mean PDC ( $PDC_{wm}$ ), and daily polypharmacy possession ratio (DPPR). We compared the sensitivities of the measures by changing parameters in the Korean nationwide claims database. First, we used  $PDC_{with \geq 1}$  as our base model. Then, we divided an adherence measure algorithm into three categories: (1) definition of data cleaning, (2) inclusion criteria and observation period, and (3) calculation methods of medication adherence. The categories included eight decision nodes that incorporated 25 alternative options. Finally, we assessed the medication adherence for the base-case with commonly used values and then varied to measure with each alternative option.

**Results:** The base-case included 14,288 beneficiaries with antihypertensives. Among eight decisions, both handling an end-date-of-study and overlaps had the strongest impacts on measuring  $PDC_{with \geq 1}$ ,  $PDC_{wm}$ , and DPPR, with small differences in sample size. Instead of the estimates of adherence from  $PDC_{wm}$ , those of  $PDC_{with \geq 1}$  and DPPR were similar. Furthermore, a tendency toward a higher medication adherence was observed with a smaller study population.

**Conclusion:** The decisions regarding identifying an end-date-of-study and overlaps showed meaningful impacts of all three measures including  $PDC_{with \geq 1}$ ,  $PDC_{wm}$ , and DPPR on measuring medication adherence.

**Keywords:** medication adherence, proportion of days covered, PDC, daily polypharmacy possession ratio, DPPR, sensitivity analysis, pharmacoepidemiology

## Introduction

Medication adherence affects clinical outcomes and healthcare costs for different individuals.<sup>1</sup> High levels of adherence to medications have been associated with decreased rates of morbidity and mortality compared to non-adherence.<sup>2</sup> Furthermore, improving medication adherence could reduce annual medical costs by between US\$5271 and US\$52,341, thereby relieving healthcare burden and saving on healthcare resources.<sup>3</sup>

Meanwhile, using accurate measures of medication adherence is important for both researchers and practitioners. This is because such accurate measures enable

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researchers to identify possible predictors related to the improvement and maximization of adherence. At the same time, these are important for practitioners in evaluating medication-taking, clinical outcomes, and cost-effectiveness, which can eventually resolve healthcare problems. Previous studies have demonstrated that the definition of medication adherence could influence the estimates of exposure risks, which could affect study outcomes.<sup>4–6</sup> Moreover, the quality of data on medication adherence can influence the interpretation of the results of epidemiologic studies and clinical trials.<sup>7,8</sup> Hence, it is critical to define medication adherence with accurate measures.

Although numerous studies have utilized adherence measures, inconsistent definitions, approaches, or methods for measuring adherence among these works remain.<sup>9–11</sup> Furthermore, the measured medication adherences depend on different modeling decisions.<sup>12–14</sup> A previous sensitivity analysis also revealed that the varying assumptions made during medication adherence estimation can impact the results of analyses.<sup>12</sup> Most researchers have emphasized the importance of standardization for measuring medication adherence.<sup>10,15</sup>

As part of understanding how to measure adherence related to chronic drug use, in the current study, patients with antihypertensive medications were examined for representative therapies with a conventional drug use of both monotherapy and combination therapy. Patients with hypertension commonly use the regimens to control their diseases with different mechanisms of actions for the drugs, with the expectation that the combination of these drugs can exert synergistic effects when used together.<sup>16</sup> However, little is known in terms of standardizing these approaches regarding measuring medication adherence. Related to this, a previous study emphasized the need to explore varied decisions on adherence with respect to combination therapy as well as monotherapy.<sup>13</sup>

The proportion of days covered (PDC) is one of the more commonly used metrics to estimate medication adherence during the implementation phase of the adherence.<sup>17</sup> According to the Pharmacy Quality Alliance, measures of PDC result in a more reliable and conservative value, particularly in cases of chronic drug therapies.<sup>18</sup> A recent systematic review of 147 studies (including observational studies, randomized controlled trials, and a validation study) has reported that PDC (17.7%) was the most frequently used concept of adherence to a combination therapy, although it did not have a uniformly defined or calculated standard.<sup>10</sup>

With the goal of analyzing the comparative impacts of varied options for medication adherence using claims data, PDC identified as considering one or more medications ( $PDC_{with \geq 1}$ ) was defined as the basic concept of PDC in the current study. Additionally, to investigate the robustness of our results, two more measures, namely, duration weighted mean PDC ( $PDC_{wm}$ ) and daily polypharmacy possession ratio (DPPR) were compared to  $PDC_{with \geq 1}$ .

## Materials and Methods

### Setting and Design

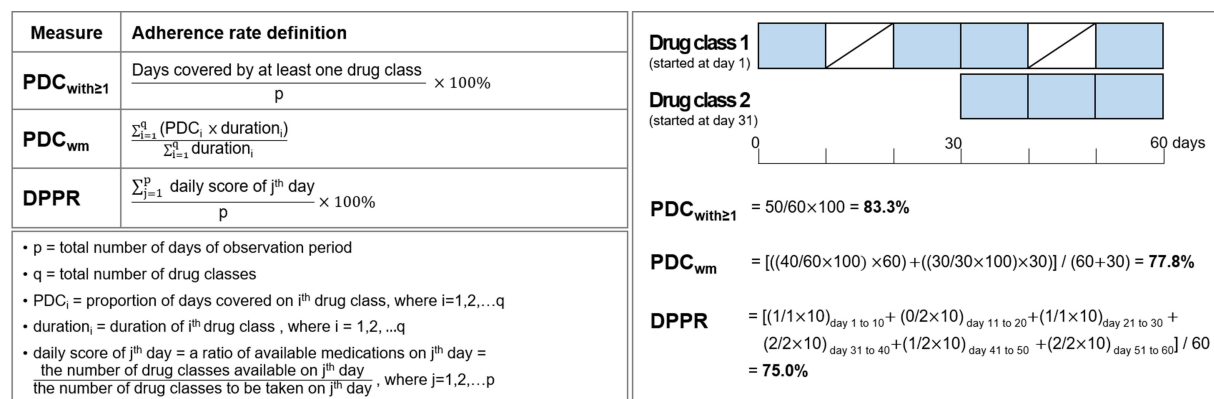
We used the population-based claim data between 2005 and 2015 from the Korean National Health Insurance Service–National Sample Cohort (NHIS-NSC) database (NHIS-2019-2-081).<sup>19</sup> First, we performed a sensitivity analysis using the NHIS-NSC data covering the following criteria: nationwide insured subjects aged over 20 years, had been diagnosed with hypertension, and had newly started antihypertensive medications, with at least one prescription in the index year of 2007. Hypertension diagnosis was identified with the International Statistical Classification of Diseases and Related Health Problems, Tenth Revision codes (I10–I15). The following five antihypertensive drug classes were included: angiotensin-converting enzyme inhibitors, angiotensin II receptor blockers, beta blockers, calcium channel blockers, and diuretics ([Supplemental Table 1](#)). The index date was identified as the first date of antihypertensive prescriptions in the year 2007.

To ensure privacy, the NHIS removed and encrypted the private and identifiable information of all subjects before providing the data. Thus, our study was exempted from acquiring patients' informed consent. Nevertheless, we still sought the approval of the Institutional Review Board of Ajou University (IRB No. 201905-HB-EX-001).

### Medication Adherence

#### Proportion of Days Covered with One or More Drugs ( $PDC_{with \geq 1}$ )

In this study,  $PDC_{with \geq 1}$  is defined as the ratio of the number of days a patient is covered by one or more medications to the total number of days within a certain period.  $PDC_{with \geq 1}$  was calculated as a ratio of two quantities, namely, the number of days covered by one or more prescriptions during the observation period, divided by the number of days from the first prescription to the end of the observation period ([Figure 1](#)).<sup>20,21</sup>



**Figure 1** Descriptive definitions of medication adherence measures including  $PDC_{with\geq 1}$ ,  $PDC_{wm}$ , and DPPR.

**Notes:** The right part describes how three measures are calculated and applied to derive medication adherences as hypothetical scenarios of claim data. Each box covers 10 days and a box with blue color means days' supply.

**Abbreviations:**  $PDC_{with\geq 1}$ , proportion of days covered with one or more medications;  $PDC_{wm}$ , duration weighted mean PDC; DPPR, daily polypharmacy possession ratio.

### Duration Weighted Mean PDC ( $PDC_{wm}$ )

$PDC_{wm}$  refers to the duration weighted mean of PDC, where some values contributed more than others, according to the duration of drug use.<sup>22</sup>  $PDC_{wm}$  was calculated as the sum of each value of multiplication of each drug's duration weight by its PDC, divided by the sum of the weights (Figure 1).

### Daily Polypharmacy Possession Ratio (DPPR)

DPPR is defined here as the proportion of all medications available for daily use. We calculated the DPPR by dividing the sum of each day's score by the number of days in the study period. Each day's score was estimated as a ratio of the number of drugs available for the day to the number of drugs to be taken for the day (Figure 1).<sup>11,23</sup>

Among various measures of medication adherence, as  $PDC_{wm}$  and DPPR, which were commonly used, novel and considered more probable and elaborate than  $PDC_{with\geq 1}$ , we compared both measures compared to  $PDC_{with\geq 1}$ .<sup>10,14,23,24</sup> We identified 80% as the threshold above which subjects were considered adherent to chronic medications.<sup>21,22,25</sup>

### Algorithm Definition

In order to compare various decisions related to antihypertensive medication adherence including both monotherapy and combination therapy using claims data, we divided the adherence-measuring algorithm into three categories: (1) "definition of data cleaning," (2) "inclusion criteria and observation period," and (3) "calculation methods of medication adherence." These three categories comprised eight decision nodes incorporating a total of 25 alternative options.

Each decision node included two or more options. Next, we defined a base-case with the most commonly used values, against which all other alternative options were compared. Then, medication adherence was measured by changing one alternative option in one decision node, keeping the base-case fixed at all but one option. Finally, to perform a sensitivity analysis that aimed to compare the various decisions, we identified which decision nodes and alternative options showed maximum variations in adherence measures, along with the sample size compared to a base-case.

### Category (1): "Definition of Data Cleaning"

The first category, "definition of data cleaning," aimed to handle the claims data to fit our study design. To calculate antihypertensive medication adherences, we analyzed differences of each alternative option selected in terms of irregular antihypertensive prescriptions and duplicate redundant claims. We considered in-hospital medication initiation, use of injectable agents, and short-term prescribing under seven days as the irregular antihypertensive medication uses. In addition, due to the nature of claims database, we considered duplicate claims as administrative error.<sup>12</sup>

### Category (2): "Inclusion Criteria and Observation Period"

This category included decisions pertaining to study designs of inclusion criteria including a minimum number of prescriptions, pre-index period, minimum cumulative number of days' supply and observation period.<sup>26</sup> In our study, a minimum of two prescriptions was identified as the base-case.<sup>27,28</sup> To align with the first prescribed episode, we defined a pre-index period or the prescription-

free period before the index date. A pre-index of 365 days, our base-case, is quite common in previous studies on chronic medications.<sup>29–31</sup> In addition, we defined the base-case for the minimum period of medication use, which was referred to how long medications were taken, as 90 days and set the observation period of 3 years in the current study.<sup>32</sup>

### Category (3): “Calculation Methods of Medication Adherence”

The last category sought to define medication adherence calculations by identifying the end-date-of-study and the overlaps. In case of handling an end-date-of-study, we compared two options: the prescription-based and the fixed period-based approach.<sup>4,22</sup> The prescription-based approach considers the end-date-of-study as the last prescription plus the supply for the inclusive days. Alternatively, the fixed period-based defines the end-date-of-study as the last day of the observation period. We also observed how different decisions made regarding overlaps of the same drug class can affect the measures used. Here, we defined the unconditional carryover of overlaps from a previous prescription to the following prescription as our base-case. Alternative approaches were considered carried over: (1) if the carryover of surplus days from the earlier prescription overlaps within a period of under 14 days, (2) if the carryover of those prescriptions was issued by the same institution, and (3) no carryover was granted. The underlying assumption for the option of the carryover was that the following prescription was taken earlier before the end of the first prescription.<sup>13,20</sup> We assumed the option of being left as it was meant being concurrently used at the same time. As combination therapies among the different drug classes are common, the carryover of overlaps was not granted for inter-classes.

### Statistical Analysis

Categorical variables were compared by using the chi-square test and were described as numbers and percentages. Continuous variables were compared by using the Kruskal–Wallis test. In spite of a large population, the variables were determined to be non-normally distributed. Thus, they were reported as medians with quantile 1–quantile 3 (Q1–Q3) and also as means with standardized deviations (SDs). The rate of adherent patients was calculated on the basis of the percentage of number of patients with over 80% adherence. When comparing three measures, percentage changes were calculated by dividing

differences in the estimates including medians and means or rates of adherent patients measured using  $PDC_{wm}$  and DPPR by those of  $PDC_{with\geq 1}$ , respectively. A p-value of  $<0.01$  was considered statistically significant. All data analyses were conducted using SAS Version 9.4 (SAS Institute, Cary, NC, USA).

## Results

### Baseline Characteristics

To compare sensitivities of the adherence using varied analytical decisions, we examined antihypertensive medication adherence including both monotherapy and combination therapy. The study population consisted of 117,771 adults who had a diagnosis of hypertension and had recently started on at least one prescription for any study antihypertensive medication in the index year of 2007 in Korea. From these, a sample of 14,288 patients with both monotherapy and combination therapy was finalized following the base-case criteria: conditions including each prescription over 7 days' supply, de-duplication of redundant prescriptions, patients with at least two prescriptions, a total of over 90 days' supply, a 365 days pre-index period, at least a 3-year observation period, identifying an end-date-of-study by a prescription-based way, and handling the carryover of overlaps from the earlier prescription for the same drug class unconditionally ([Supplemental Table 2](#)). In accordance with the base-case, the mean (SD) age of the subjects was 57.42 (12.86) years. There were 7440 (52.1%) male patients and 1219 (8.5%) disabled persons. The antihypertensive therapy at index date consisted of monotherapy (60.9%), dual therapy (29.9%), and more than triple therapy (9.3%). A total of 9857 (69.0%) patients had visited 1 or 2 medical care institutions during the observation period, whereas 4431 (31.0%) patients had visited 3 or more than institutions ([Table 1](#)). In the following sections, our results reveal the sensitivity of  $PDC_{with\geq 1}$  for each decision node, followed by a comparison of the results of  $PDC_{wm}$  and DPPR with those of  $PDC_{with\geq 1}$ , respectively. Additionally, we identified medication adherence while dividing monotherapy and combination therapy at the index date in [Supplemental Tables 3–5](#).

### Category (I): “Definition of Data Cleaning”

For the base-case of our sensitivity analysis, the sample had a mean (SD) and median (Q1–Q3)  $PDC_{with\geq 1}$  of

**Table I** Baseline Characteristics of Study Population in the Base-Case Setting

Characteristic		Total
		N(%)
Overall		14,288
Sex	Male	7440 (52.1)
	Female	6848 (47.9)
Age	Mean $\pm$ SD	57.42 $\pm$ 12.86
	20–39	1111 (7.8)
	40–49	3068 (21.5)
	50–59	3897 (27.3)
	60–69	3499 (24.5)
	$\geq 70$	2713 (19.0)
Disability	Yes	1219 (8.5)
Antihypertensive therapy	Mono therapy	8696 (60.9)
	Dual therapy	4267 (29.9)
	$\geq 3$ classes	1325 (9.3)
Type of health insurance	Regional	4672 (32.7)
	Employment-based	8527 (59.7)
	Medical aid	1089 (7.6)
Number of medical institution visited	1	5380 (37.7)
	2	4477 (31.3)
	$\geq 3$	4431 (31.0)
Insurance contributions	High	5504 (38.5)
	Moderate	4749 (33.2)
	Low	4035 (28.2)

76.70% (24.59) and 86.80% (63.05–96.15), respectively, with 8549 (59.8%) adherent patients. After comparing the elimination of prescriptions with irregular use of antihypertensive medications in the base-case (the short-term users were defined as subjects with under 7 days' supply), the other alternative options (including in-hospital medication initiation and in-hospital prescription or use of injectable agents) produced similar results (0.79% and 0.84%, respectively). Furthermore, the options of whether

duplicate claims were removed or retained showed the same results for  $PDC_{with \geq 1}$  (differences of mean and median were 0.33% and 0.57%, respectively), with a slight difference in sample size (12 individuals), as shown in Table 2.

## Category (2): “Inclusion Criteria and Observation Period”

In our study, three node-related inclusion criteria and one node-related observation period were identified sensitively to assess the medication adherence. As can be seen, the results of  $PDC_{with \geq 1}$  were similar when minimum numbers of prescriptions were 2 or 4, with only 0.19% of mean difference  $PDC_{with \geq 1}$  and with decreasing population size of 274 people. For the second decision node, compared to the 365-day pre-index period, the median  $PDC_{with \geq 1}$  decreased by 1.12% with the reduced 180-day pre-index period, and the median  $PDC_{with \geq 1}$  increased by 0.71% with the increased 730-day pre-index period. In addition, as the minimum cumulative number of days' supply increased from 30 days to 1 year, the study sample size decreased from 15,510 to 11,802 patients, and the percentages of adherent patients increased from 57.7% to 67.5% with the change of mean (SD)  $PDC_{with \geq 1}$  from 74.42% (27.11) to 83.12% (17.18). Moreover, we found that the mean (SD)  $PDC_{with \geq 1}$  decreased as the observation period became longer from the largest value of 84.04% (18.28) during the 1-year observation period to the smallest value of 73.43% (26.36) during the 8-year observation period. The differences between the largest and smallest values for the number of included patients and the rate of adherent patients were 2551 and 16.7%, respectively (Table 2 and Supplemental Table 6).

## Category (3): “Calculation Methods of Medication Adherence”

Table 2 presents the comparisons of  $PDC_{with \geq 1}$  concerned with an end-date-of study and overlaps for the same drug class. As can be seen, the mean  $PDC_{with \geq 1}$  decreased by 5.8% when we defined the end-date-of-study as a fixed period-based concept (70.94%) compared to the base-case of prescription-based concept (76.70%). The percentage of adherent patients was higher in the base-case than that in the alternative option (59.8% vs 53.2%), with a small difference in sample size (14,288 vs 13,950). Furthermore, the overlapped two prescriptions were observed to be quite common with the chronic use of



**Table 2** Diverse Impacts of Measures of Medication Adherence with Varied Options on PDC<sub>with≥1</sub>

Decision Node	N	PDC <sub>with≥1</sub>		Rate of Adherent Patients (≥ 80%)	
		Mean ± SD	Median (Q1–Q3)	N (%)	p-value
Category 1: Definition of data cleaning					
1. Elimination claims that are irregular for hypertension medication					
a. In-hospital medication initiation	14,311	75.91 ±24.85	85.96 (61.70–95.75)	8367 (58.5)	0.03
b. In-hospital medication initiation or use of injectable agents	14,306	75.92 ±24.84	85.98 (61.72–95.75)	8368 (58.5)	
c. Short-term prescribing under 7 days*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	
Difference (Max - Min)	23	0.79	0.84	182	
2. Elimination of redundant duplicate claims					
a. Deduplication*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	0.32
b. Leave as it was	14,276	77.03 ±24.65	87.37 (63.48–96.47)	8624 (60.4)	
Difference (Max - Min)	12	0.33	0.57	75	
Category 2: Inclusion criteria and observation period					
3. Inclusion criteria for a minimum number of prescriptions					
a. At least two prescriptions*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	0.84
b. At least four prescriptions	14,014	76.89 ±24.31	86.81 (63.39–96.12)	8402 (60.0)	
Difference (Max - Min)	274	0.19	0.01	147	
4. Inclusion criteria for pre-index period					
a. 180 days	16,397	75.82 ±24.79	85.68 (61.50–95.87)	9506 (58.0)	<0.0001
b. 365 days*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	
c. 730 days	12,656	77.26 ±24.46	87.51 (64.13–96.34)	7748 (61.2)	
Difference (Max - Min)	3741	1.44	1.83	1758	
5. Inclusion criteria for supply for the minimum cumulative number of days					
a. 30 days	15,510	74.42 ±27.11	85.74 (59.14–96.05)	8956 (57.7)	<0.0001
b. 90 days*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	
c. 180 days	13,194	79.19 ±21.73	87.93 (67.32–96.34)	8237 (62.4)	
d. 1 year	11,802	83.12 ±17.18	89.77 (73.60–96.72)	7964 (67.5)	
Difference (Max - Min)	3708	8.7	4.03	992	
6. Observation period					
a. 1 year	12,620	84.04 ±18.28	91.41 (76.61–97.83)	8998 (71.3)	<0.0001
b. 2 years	13,681	79.27 ±22.56	88.79 (67.77–96.69)	8723 (63.8)	
c. 3 years*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	
d. 5 years	14,888	74.11 ±26.17	84.56 (58.18–95.49)	8374 (56.2)	
e. 8 years	15,171	73.43 ±26.36	83.37 (57.92–95.06)	8277 (54.6)	

(Continued)

Table 2 (Continued).

Decision Node	N	PDC <sub>with≥1</sub>		Rate of Adherent Patients (≥ 80%)	
		Mean ± SD	Median (Q1–Q3)	N (%)	p-value
Difference (Max - Min)	2551	10.61	8.0	721	
Category 3: Calculation methods of medication adherence					
7. Handling end-date-of-study					
a. Prescription-based approach*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	<0.0001
b. Fixed period-based approach	13,950	70.94 ±28.82	82.76 (50.55–95.16)	7417 (53.2)	
Difference (Max - Min)	338	5.76	4.04	1132	
8. Handling overlaps					
a. No carryover was granted	14,222	74.47 ±23.65	84.08 (61.50–92.75)	8096 (56.9)	<0.0001
b. Carryover of overlaps unconditionally*	14,288	76.70 ±24.59	86.80 (63.05–96.15)	8549 (59.8)	
c. Carryover of overlaps if a period of overlaps was under 14 days	14,270	75.79 ±24.26	85.52 (62.45–94.83)	8375 (58.7)	
d. Carryover of overlaps if prescriptions were issued by the same institution	14,293	76.38 ±24.52	86.34 (62.74–95.74)	8492 (59.4)	
Difference (Max–Min)	71	2.23	2.72	453	

**Note:** \*The base-case of this study.

**Abbreviation:** PDC<sub>with≥1</sub>, proportion of days covered with one or more medications.

antihypertensive medications. In terms of handling the overlapped days of the same drugs, the three other approaches including the base-case, carryover when overlaps were under 14 days, or carryover when issued by the same institution had similar results, with higher than the option of left as it was. The highest median PDC<sub>with≥1</sub> was 86.80% in the base-case, and the lowest median PDC<sub>with≥1</sub> was 84.08% in left as it was, with similar number of included patients (14,288 vs 14,222). The results of three categories related to PDC<sub>with≥1</sub> are plotted in Figure 2.

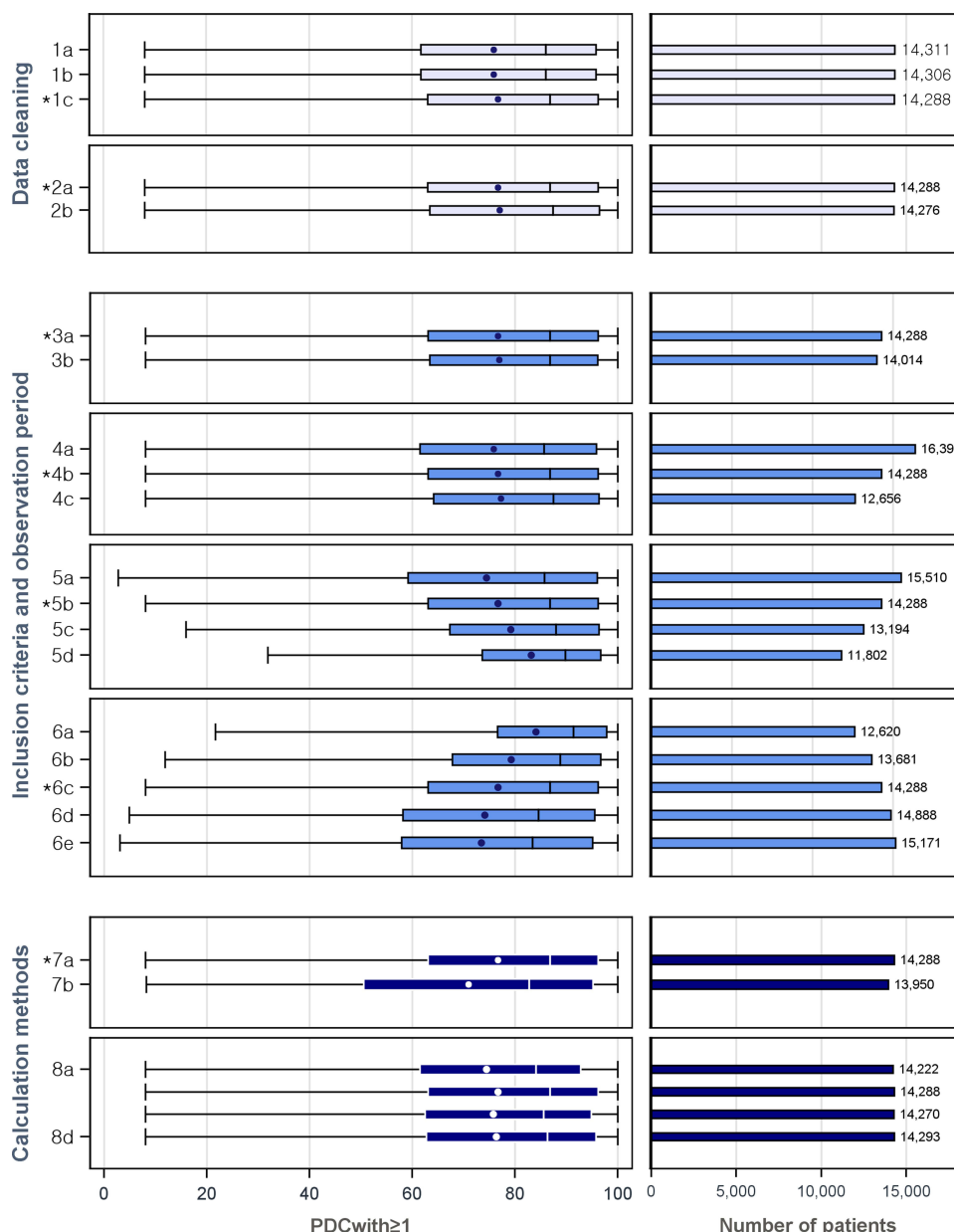
## Comparison with PDC<sub>w/m</sub> and DPPR to PDC<sub>with≥1</sub>

Thus far, we have described the relative effects on PDC<sub>with≥1</sub> of different options in each decision node. However, considering the nature of antihypertensive medications, such as combination therapy, more conservative measures, such as PDC<sub>w/m</sub> and DPPR, were comparatively evaluated. The corresponding results from these measures were compared with those of PDC<sub>with≥1</sub>. As revealed by the results, the overall trends of the abovementioned two measures were similar to those of PDC<sub>with≥1</sub> in every

single choice. At the base-case, we found that the estimates for DPPR were more similar to those for PDC<sub>with≥1</sub> and larger than the results for PDC<sub>w/m</sub>. The percentage changes in mean (SD), median (Q1–Q3), and the rate of adherent patients between PDC<sub>w/m</sub> and PDC<sub>with≥1</sub> and between DPPR and PDC<sub>with≥1</sub> were –2.76% vs –0.46%, –3.70% vs –0.47%, and –7.85% vs –0.69%, respectively (Table 3 and Supplemental Figure 1).

## Discussion

In this study, we examined the relative impacts of diverse methodological decisions on medication adherence using claims data, the results of which are relevant in studies that examine health or cost-effectiveness outcomes. We compared the influences of diverse options for PDC<sub>with≥1</sub>, PDC<sub>w/m</sub>, and DPPR by measuring medication adherence while considering both monotherapy and combination therapy. Although our findings are small absolute differences with statistical significances, we could provide evidence regarding which decision is proper depending on research purposes and clinical situations. To the best of our knowledge, this is the first sensitivity analysis with



**Figure 2** Diverse impacts of measures of medication adherence and rates of adherent patients with varied options on PDCwith≥1.

**Notes:** Among the eight decision nodes measured, the two choices that had the most impact were handling the end-date-of-study and overlaps in the third category, with few differences in the sample size. Decision node 1. Elimination claims that are irregular (1a: In-hospital medication initiation, 1b: In-hospital medication initiation or use of injectable agents, 1c: Short-term prescribing under 7 days). Decision node 2. Elimination of duplicate claims (2a: Deduplication, 2b: Leave as it was). Decision node 3. Minimum number of prescriptions (3a: At least two prescriptions, 3b: At least four prescriptions). Decision node 4 Pre-index period (4a: 180 days, 4b: 365 days, 4c: 730 days). Decision node 5. Minimum cumulative number of days' supply (5a: 30 days, 5b: 90 days, 5c: 180 days, 5d: 1 year). Decision node 6. Observation period (6a: 1 year, 6b: 2 years, 6c: 3 years, 6d: 5 years, 6e: 8 years). Decision node 7. Handling end-date-of-study (7a: Prescription-based, 7b: Fixed period-based). Decision node 8. Handling overlaps (8a: No carryover was granted, 8b: Carryover of overlaps unconditionally, 8c: Carryover of overlaps if a period of overlaps was under 14 days, 8d: Carryover of overlaps if prescriptions were issued by the same institution). \*The base-case of this study.

**Abbreviation:** PDCwith≥1, proportion of days covered with one or more medications.

different options, with the aim of comparing three adherence measures of antihypertensive medications in Korea.

We calculated medication adherence in the base-case with commonly used values and then varied each alternative option individually to compare their impacts on the adherence. While less variation was observed when

comparing most of the alternatives to the base-case, certain decision nodes seemed to have had meaningful impacts on measures of adherence. Among the eight decision nodes measured, the two choices that had the most impact were handling the end-date-of-study and overlaps in the third category. These two nodes showed influential



**Table 3** Comparing the Varied Impacts of PDC<sub>w/m</sub> and DPPR Measures Compared to PDC<sub>with≥I</sub>

Decision Node	Case	PDCwith≥I Median (Q1–Q3)	PDCwm		DPPR	
			Median (Q1–Q3)	%Change from PDCwith≥I	Median (Q1–Q3)	%Change from PDCwith≥I
Base-case setting		86.80 (63.05–96.15)	83.59 (59.89–94.92)	–3.70	86.39 (62.58–95.84)	–0.47
Category 1: Definition of data cleaning						
1. Elimination claims that are irregular for hypertension medication						
Smallest	a. In-hospital medication initiation	85.96 (61.70–95.75)	82.80 (58.66–94.58)	–3.68	85.66 (61.35–95.49)	–0.35
Largest	c. Short-term prescribing under 7 days	86.80 (63.05–96.15)	83.59 (59.89–94.92)	–3.70	86.39 (62.58–95.84)	–0.47
2. Elimination of redundant duplicate claims						
Smallest	a. Deduplication	86.80 (63.05–96.15)	83.59 (59.89–94.92)	–3.70	86.39 (62.58–95.84)	–0.47
Largest	b. Leave as it was	87.37 (63.48–96.47)	84.03 (60.29–95.17)	–3.82	86.98 (62.99–96.13)	–0.45
Category 2: Inclusion criteria and observation period						
3. Inclusion criteria for a minimum number of prescriptions						
Smallest	a. At least two prescriptions	86.80 (63.05–96.15)	83.59 (59.89–94.92)	–3.70	86.39 (62.58–95.84)	–0.47
Largest	b. At least four prescriptions	86.81 (63.39–96.12)	83.59 (60.23–94.85)	–3.71	86.40 (62.92–95.79)	–0.47
4. Inclusion criteria for pre-index period						
Smallest	a. 180 days	85.68 (61.50–95.87)	82.27 (58.10–94.49)	–3.98	85.29 (61.01–95.52)	–0.46
Largest	c. 730 days	87.51 (64.13–96.34)	84.38 (61.09–95.13)	–3.58	87.15 (63.79–96.00)	–0.41
5. Inclusion criteria for supply for the minimum cumulative number of days						
Smallest	a. 30 days	85.74 (59.14–96.05)	82.36 (55.92–94.77)	–3.94	85.33 (58.64–95.71)	–0.48
Largest	d. 1 year	89.77 (73.60–96.72)	86.73 (69.46–95.71)	–3.39	89.48 (73.17–96.44)	–0.32
6. Observation period						
Smallest	e. 8 years	83.37 (57.92–95.06)	76.22 (52.19–91.86)	–8.58	82.97 (57.59–94.75)	–0.48
Largest	a. 1 year	91.41 (76.61–97.83)	90.12 (74.28–97.07)	–1.41	90.93 (76.08–97.33)	–0.53
Category 3: Calculation methods of medication adherence						
7. Handling end-date-of-study						
Smallest	b. Fixed period-based approach	82.76 (50.55–95.16)	79.10 (48.08–93.80)	–4.42	82.44 (50.09–94.89)	–0.39
Largest	a. Prescription-based approach	86.80 (63.05–96.15)	83.59 (59.89–94.92)	–3.70	86.39 (62.58–95.84)	–0.47

(Continued)

Table 3 (Continued).

Decision Node	Case	PDCwith≥I Median (Q1–Q3)	PDCwm		DPPR	
			Median (Q1–Q3)	%Change from PDCwith≥I	Median (Q1–Q3)	%Change from PDCwith≥I
8. Handling overlaps						
Smallest	a. No carryover was granted	84.08 (61.50–92.75)	81.06 (58.07–91.76)	–3.59	83.73 (61.00–92.56)	–0.42
Largest	b. Carryover of overlaps unconditionally	86.80 (63.05–96.15)	83.59 (59.89–94.92)	–3.70	86.39 (62.58–95.84)	–0.47

**Abbreviations:** PDC<sub>with≥1</sub>, proportion of days covered with one or more medications; PDC<sub>w</sub>, duration weighted mean PDC; DPPR, daily polypharmacy possession ratio.

impacts on adherence measures, with few differences in the sample size. For our results, diverse options for handling end-date-of-study and overlaps varied in terms of sample size only: 338 (2.4%) and 71 (0.5%), respectively. The small difference between the numbers of study samples could mean that each population of each option is similar, resulting in little selection bias and revealing the impacts of decisions. Furthermore, the impacts of these decisions on medication adherence might be related to methodological approaches, rather than data sources or study purposes, thus ruling out the possibility of bias related to the study cohort. On the basis of our findings, we demonstrated that two nodes must be taken into account when performing analyses, interpreting the research outcomes, and comparing them to others. Consistent with our findings, the results of recent sensitivity analyses highlighted the importance of using an analytical approach in examining the end-date-of-study and overlaps.<sup>12,13</sup> Previous studies also supported our findings, which indicated that PDC<sub>with≥1</sub> from the fixed period-based approach was lower than that of the prescription-based approach.<sup>22,24</sup> As for the carryover, the values of PDC<sub>with≥1</sub> between the condition with under 14 days and the same institution were similar. This can be inferred to mean that most prescriptions regarding overlaps were issued by the same institution.

Moreover, a tendency toward a higher medication adherence was observed with a smaller study population. Previous studies supported our finding that the longer observation period with the larger study population led to the smaller PDC<sub>with≥1</sub> observed.<sup>9,33</sup> We assumed that more requirements for inclusion would be relevant to a smaller sample size remaining a sample

with a higher adherence. Moreover, selection bias from different samples may affect to broaden diversity of medication adherence estimates, thus indicating the importance of controlling the bias.<sup>34</sup> These findings are in line with those reported by a previous study, which examined the impacts of the medication possession ratio for a single drug to modeling decisions in large claims databases.<sup>13</sup>

One outstanding finding of our study is that the three measures of PDC<sub>with≥1</sub>, PDC<sub>w</sub>, and DPPR had similar results that handling the end-date-of-study and overlaps in the third category were impactful, thus ensuring the robustness of this study. Especially, the estimates for PDC<sub>with≥1</sub> and DPPR had similar results. This could mean that, among the three, the conceptually simplest approach (PDC<sub>with≥1</sub>) was almost the same as the most theoretically precise and complicated approach (DPPR) in measuring antihypertensive medication adherence while tracking combination therapy. Meanwhile, PDC<sub>w</sub> was adjusted by the average duration; thus, it seemed to be lower than those two measures. Choosing a simple and accurate method can provide efficient processes and results for both researchers and clinicians, considering both the factors of reliability and applicability. In this study, we verified that applying PDC<sub>with≥1</sub>, instead of DPPR, is sufficient when measuring antihypertensive medication adherence. The findings of this study may, therefore, provide better insights in identifying an easier approach for measuring adherence.

However, a careful interpretation of our findings is needed due to some limitations. First, we did not examine all the impacts of different decisions simultaneously, but only identified those of each decision given that there were 5760 cases. Thus, our findings might be conservative, and

further research is warranted. Second, we did not examine medication adherence upon the antihypertensive therapy or dosage change, as treatment regimens of antihypertensive medications tend to change frequently.<sup>16</sup> Moreover, a previous study showed that different methods of assuming antihypertensive medications' prescribed daily dose had significant differences on calculating PDC in patients with chronic heart failure using claims data.<sup>35</sup> Further research is warranted to validate these factors. Third, each antihypertensive drug class has a specific indication and regimen.<sup>36</sup> However, as we mainly focused on the methodological impacts of each decision on adherence, we did not perform analyses according to drug classes. Previous studies have shown associations between antihypertensive medication adherence and drug classes, with lower adherence to diuretics and highest adherence to angiotensin II receptor blockers.<sup>37,38</sup> Finally, our results regarding antihypertensive prescriptions based on claims data, representing proxy measures of adherence, might be altered according to different types of drugs, patterns of use, or data sources. According to these factors, as sensitivities of measures of medication adherence could be different,<sup>12</sup> additional research is warranted to validate the outcomes on each occasion.

Despite these limitations, our study reflects the real-world setting based on a Korean representative claims database, the NHIS-NSC. Although a large number of medication adherence measurement methods exist, no standard has yet to be identified.<sup>10</sup> Imprecise measures of medication adherence can induce misinterpreted efficacy or safety of therapy, thus leading to possibly costly and hazardous problems. To help resolve these problems, the present study examined various approaches for measuring adherence and provided evidence on which factor was the most impactful. Our results could add more insights into the measures of medication adherence, reflecting patterns of antihypertensive medication use in Korea.

## Conclusion

We identified the comparative impacts of different options for medication adherence measures including  $PDC_{with \geq 1}$ ,  $PDC_{wm}$ , and DPPR using a Korean nationwide claims database with antihypertensive prescriptions. The decisions regarding identifying an end-date-of-study and overlaps showed meaningful impacts of all three measures on measuring medication adherence. Our findings suggest that researchers focus on these decisions when deciding on a proper approach for their research. Further sensitivity

studies are needed to examine the validity and reliability of measuring adherence for antihypertensive drugs.

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## Disclosure

The authors report no conflicts of interest in this work.

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