

# Nonurgent Patients' Preferences for Emergency Department Versus General Practitioner and Effects of Incentives: A Discrete Choice Experiment

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## Abstract

**Objective.** This study investigates potential of a new financial incentive policy, the GP-referral discount scheme introduced in Singapore, in reducing nonurgent emergency department (ED) visits, and compares it with alternative interventions. **Methods.** A discrete choice experiment (DCE) was designed to elicit patients' preferences for ED and general practitioner (GP) under hypothetical nonurgent medical conditions. Through latent class multinomial logistic regression, choice models were estimated to quantify how patients' choices are influenced by GP-referral discount, other ED/GP attributes (waiting time, test facilities, and payment), patient demographics, and their perception of severity. The choice models were used to predict uptake of the GP-referral discount scheme and other countermeasures suggested by these models. **Results.** Survey responses from 849 respondents recruited from a public hospital in Singapore were included in the study. The choice model identified two prominent classes of patients, one of which was highly sensitive to GP-referral discount and the other to test-facility-availability. Patients' perceptions of severity ("critical" v. "not critical" enough to go to ED directly) were highly significant in influencing preference heterogeneity. Predictive analysis based on the choice model showed that GP-referral discount is more effective when patients visit ED expecting "shorter" waits, as opposed to test-facility provision at GPs and perception-correction measures that showed stronger effects under "longer" expected waits. **Conclusions.** The new GP-referral financial incentive introduced in Singapore can be effective in reducing nonurgent ED visits, if it reasonably covers the (extra) cost of visiting a GP. It may serve as a complement to test-facility provision at GPs or perception-correction measures, as the financial incentive and the latter two measures appear to influence distinct classes (discount-sensitive and facility-sensitive) of patients.

## Keywords

discrete choice experiment, emergency department crowding, GP-referral discount, nonurgent emergency department visits, patient perception, patient choice model

Date received: April 16, 2021; accepted: May 14, 2021

## Introduction

Nonurgent patients' visits to emergency departments (EDs) has been a global problem for many years. In several countries these patients constitute a large proportion of ED demand; for example, 9% to 60% in the United States and Canada, 19.6% to 40.9% in Europe, and 30%

to 57% in Singapore and Hong Kong.<sup>1–3</sup> The nonurgent use of ED causes crowding and longer waiting times at

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EDs, unnecessary testing, excessive health care spending, and weaker patient-primary care provider relationships.<sup>4</sup> Despite such severity, nonurgent ED visits largely remain an unresolved problem.<sup>1,5</sup>

A big challenge in addressing nonurgent ED visits is that hospitals cannot bar patients from directly accessing ED, and that nonurgent patients often “prefer” ED to general practitioners (GPs) due to factors such as wrong perception of severity, belief that ED provides better care, or lack of incentives to visit GPs even when they are accessible.<sup>2</sup> In Singapore, where this study was conducted, there are about 1700 private GP clinics that provide within walking distance primary care service in most neighborhoods, and accept walk-in patients without requiring advance appointment. Yet, nonurgent ED visits remain an issue.<sup>3</sup> Therefore, in general, interventions/incentives must be designed to encourage nonurgent patients to choose primary care.

In the past, some countries have tried interventions such as higher ED co-payment, tele-advice services, education on proper ED usage, strengthening primary care linkage, and improving access to GP care. Among studies that examined tele-advice services in the United Kingdom and the United States, most found no significant reduction in ED visits,<sup>6</sup> but explanations are lacking whether this is due to limitations of tele-advice services or other potential factors. Public education campaigns in the past showed mixed results in the United States.<sup>7</sup> In Singapore, such campaigns were initially effective, but a few months after each campaign, ED visits reverted to precampaign levels.<sup>8</sup> Some recent studies in the United Kingdom and Australia have shown positive impact of extended GP hours and in-hospital GP walk-in centers, but in the past, opening of such centers in Singapore showed no effect.<sup>9</sup> As discussed in reviews,<sup>7,10</sup> studies on interventions have often failed to reach a consensus

about their potential/effectiveness due to lack of understanding of the interplay of various factors.

Compared to other interventions, ED co-payment/fee increase has shown more consistent positive results.<sup>7,10</sup> It has been observed though that such disincentives need to be significant to have an impact. For example, ED co-payments for Medicaid beneficiaries in the United States showed reduction in ED visits in Oregon (with \$50 co-payment per ED visit), but no influence on ED visits in other states (with \$2–\$8 ED co-payment).<sup>10</sup> In Singapore, ED fee increases between 1987 and 1997 showed reduction in ED visits only after ED fees significantly exceeded the charges levied by primary care services.<sup>8</sup> Though high financial penalties seem to be effective, there have been concerns on imposing high ED co-payments/fees, as penalties for ED use could delay the needed care, especially for low-income populations.<sup>10</sup>

Recently, a new financial incentive policy—the *GP-referral discount scheme*—was introduced as a pilot by a public hospital in Singapore (here onward referred to as the Public Hospital) so as to encourage nonurgent patients to first visit GPs instead of directly visiting the ED. Under this scheme, patients who are referred to the Public Hospital ED by participating GPs receive a discount on their ED fee, whereas those who directly go to the ED receive no discount.<sup>3</sup> (In 2017, ED fee at the Public Hospital was flat S\$120, and the pilot scheme offered S\$50 discount for GP-referred patients.) As opposed to increase in ED co-payment/fee (which undesirably imposes penalty for ED use even on urgent patients), the above incentive has the advantage that it does not impose penalty for ED use. Such an incentive also has the potential to be adopted by health care systems where ED fee increase is not even feasible (e.g., universal publicly funded systems\*). As it is the first-of-its-kind incentive scheme, it is of interest to explore its influence; which is one of the main motivations of our study.

In this study, we evaluate the potential of the above GP-referral discount scheme in reducing nonurgent ED visits, and also compare it with other possible interventions in Singapore context, for example, test-facility provision at GPs and publishing of ED waiting times. To understand the interplay of various factors and appropriately evaluate the interventions, we follow a discrete choice experiment (DCE)<sup>11</sup> approach that enables us to quantify patients’ sensitivity to different factors (including GP-referral discount, other service attributes of ED/GP such as waiting time, test-facilities and payment, as

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Engineering Systems and Design, Singapore University of Technology and Design, Singapore (YS, SS); Health Services and Systems Research Program, Duke-NUS Medical School, Singapore (SO); Health Services Research Department (WLC, HCO) and Accident & Emergency Department (LT), Changi General Hospital, Singapore. The author(s) declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article. The author(s) disclosed receipt of the following financial support for the research, authorship, and/or publication of this article: Financial support for this study was provided by grants from SUTD-MIT International Design Center (IDC) (Grant Award No. IDG21500103) and Ministry of Education, Singapore (MOE AcRF Tier-2 Grant No. MOE2019-T2-2-168). Any findings, conclusions, or recommendations expressed in this article are those of the authors and do not necessarily reflect the views of the IDC or the Ministry of Education, Singapore. The funding agreement ensured the authors’ independence in designing the study, interpreting the data, writing, and publishing the report.

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\*Such systems could possibly implement a referral-based incentive by compensating the patients referred from GP to ED for the extra costs (due to travel, wait, etc.) incurred in visiting a GP.

well as patients' personal characteristics) in choosing ED versus GP. By virtue of DCE, we are able to investigate effectiveness of GP-referral discounts other than the specific discount offered in the pilot, which provides useful insights for further implementation/adoption of the scheme. Furthermore, we are able to assess and compare it with other possible interventions before their implementation. Through our DCE design, we also quantify effects of presenting nonurgent *medical conditions* and patients' *perceptions* (self-perceived urgency of the medical condition) on their choices. Though some previous studies<sup>12-14</sup> have observed importance of perceptions in patients' choices between ED and GP, we are the first to quantify its influence and thereby assess the benefits of perception-improvement measures (e.g., tele-advice services).

We are among the first to use a DCE approach to quantify nonurgent patients' preferences, and thereby assess interventions for nonurgent ED visit reduction. Previous studies on nonurgent patients' preferences have mostly pursued qualitative investigation,<sup>15</sup> which have identified their primary driving factors as cost, perceived severity, expecting to require tests, greater trust on ED services, and lack of access to timely care outside ED. A few studies have explored quantitative choice models using "revealed preference" data of patients' visits to EDs/GPs.<sup>16-18</sup> However, they have estimated only the influence of patients' sociodemographic characteristics and cost of care. Recently, DCE with mixed logit regression has been used<sup>12,19,20</sup> to quantify influence of ED/GP service attributes on patients' preferences for ED/GP, and to compare effects of different medical scenarios. Some common findings from these studies are that patients prefer lower costs, shorter waiting times, and quality care.

Our study differs from the above-mentioned DCE studies in multiple aspects. First, we investigate influence of the "GP-referral discount" along with other factors. Second, we study patients' preferences under four different nonurgent scenarios (previous studies considered only one or two nonurgent scenarios) that capture the most common health conditions for which nonurgent patients visit EDs in Singapore. A comprehensive analysis of these helps not only study heterogeneity due to medical scenarios but also to better understand the potential of GP-referral discount scheme and other investigated measures in reducing nonurgent ED visits. Third, we also quantify influence of patients' perceptions (self-perceived urgency for the presented scenario) on their choices, and thereby also investigate implications of perception improvement, which has not been done before. Thus, our study and experiment design differ significantly from previous DCE studies, and aim to provide a

good understanding of patients' tradeoffs that can provide useful guidelines for designing effective interventions/reforms.

## Methods

### *Study Setting*

This study was conducted in Singapore in collaboration with a public hospital, Changi General Hospital (CGH). The study was ethically reviewed and approved by SingHealth Centralized Institutional Review Board (Reference Nos. 2016/2006, 2016/2619, 2016/2620, 2017/2927) and Singapore University of Technology and Design Institutional Review Board (Reference No. 15-088).

In Singapore, patients can seek primary care treatment at private GP clinics or subsidized polyclinics, the former of which meet about 80% of the total primary care demand.<sup>21</sup> Emergency care is provided by EDs at both private and public acute hospitals. EDs, polyclinics, as well as GPs accept "walk-in" patients. Compared to an ED, a GP has limited test facilities and fewer senior doctors, but has lower wait (compared to nonurgent patients' wait at EDs) and lower consultation costs in general. For this study, we focused on patients' choices between ED and GP, in particular for "nonurgent" health conditions that constitute nonurgent ED demand.

### *Study Design*

A DCE is a commonly used technique in health economics (and many other fields) to elicit patient preferences for health care products and services.<sup>11</sup> A DCE presents respondents with a sequence of *choice tasks*, where they are asked to make a choice between *alternatives* (e.g., health care providers) that are described by a set of *attributes* (e.g., waiting time, payment, etc.). These attributes can be designed to capture various characteristics of health care services and policies of interest. By varying *attribute-levels* among the choice tasks, respondents' tradeoffs for attributes can be observed, which enables estimation of choice models using statistical techniques.<sup>22</sup>

For our study, DCE was the most suitable approach for the following reasons. First, since GPs in Singapore have private management and do not record patients' visit statistics, we did not have patients' data from the GPs to be able to analyze effectiveness of the new GP-referral discount scheme. Second, we were also interested in examining influence of discount amounts other than the specific amount offered in the pilot, which was not possible with pilot data. Third, our factors of interest also included patients' perceptions of medical conditions

**Table 1** List of Attributes and Levels Used for Discrete Choice Experiment Design

Attributes	ED Levels	GP Levels
Waiting time to see a doctor	30 minutes 1 hour 2 hours 4 hours	15 minutes 30 minutes 1 hour 2 hours
Test facilities (blood test, X-ray, ECG, etc.)	All available	None available All available
Out-of-pocket payment	S\$50 S\$100 S\$200	S\$50 S\$100 S\$200
GP-referral discount (when GP refers to ED)	No discount	No discount S\$50 discount on ED fee Entire GP fee waived

ECG, electrocardiogram; ED, emergency department; GP, general practitioner.

and ED/GP service attributes. For these a revealed preference study is not possible in general because ED/GP visit data lacks information on patients' perceptions, and though real data of service attributes (waiting time, payment, etc.) can be observed for the medical provider a patient visits, corresponding data for the other provider does not exist. DCE thus provided a feasible approach for our objectives, whereby we could elicit patients' perceptions and could explicitly specify ED/GP attributes one of which also captured the GP-referral discount.

Unlike previous DCE studies on preferences between ED and GP, which used unlabeled\* DCE designs, we used a *labeled* DCE<sup>23,24</sup> where choice alternatives were labeled (Option-ED, Option-GP) and were constructed from ED/GP-specific attribute-levels. Labeled design enabled us to capture influence of the GP-referral discount scheme (as the discount applies only if patients choose the GP option), and estimate patients' inherent preference for ED/GP. Furthermore, it offered the advantage that we could specify ED/GP-specific waiting times and test facilities.

### Study Protocol

*Attribute and Level Identification for DCE.* To identify relevant attributes for DCE design, we first conducted focus group discussions (FGDs) with 20 patients recruited from CGH, who had visited in the past 3 months 1) an ED, without any referral (8 patients); 2) an ED after referral from GP (7 patients); or 3) a GP and

were not referred to ED (5 patients). In the appendix, we provide participants' demographics (Table 6) and details of FGD execution and analyses. Based on FGD results and our follow-up discussions with ED physicians and administrators, we included in our DCE the ED/GP *attributes* shown in Table 1. Since we were particularly interested in investigating influence of the aforementioned GP-referral discount scheme, we also included "GP-referral discount" as an attribute in the DCE.

In our labeled DCE, each choice task asked respondents to choose between an ED and a GP alternative comprising of the four attributes shown in Table 1. For each attribute, the ED/GP-specific *levels* (also shown in Table 1) were chosen based on Singapore health care setting, after discussions with ED physicians and administrators. Details of these levels' selection is provided in the appendix. For the attribute "GP-referral discount," the GP-level "No discount" represented the status quo, "S\$50 discount on ED fee" represented the pilot scheme offer, and "Entire GP fee waived" represented a possible alternative for future. If in a choice task, GP's out-of-pocket payment was S\$50 (respectively, S\$100, S\$200) and GP-referral discount was "Entire GP fee waived," then it implied that the patient would get back her S\$50 (respectively, S\$100, S\$200) GP payment if the GP referred her to ED.

*DCE Design.* Having specified Table 1, we adopted a blocked fractional factorial design<sup>25</sup> (using Lighthouse Studio, version 9.5.3, Sawtooth Software, Inc.) to generate 40 choice tasks. Each task contained one ED level and one GP level from Table 1 corresponding to each attribute. Figure 1 shows an example of one such *random* choice task. We randomly divided the above tasks into five *blocks* with eight tasks each. The conventional

\*In the unlabeled DCEs, respondents' choice tasks consisted of generic health care provider alternatives (Option A v. Option B), both of which were constructed from the same set of attribute-levels.

If you have to choose between your preferred ED and preferred GP under the following conditions, which one would you choose?						
	Random choice task		Holdout choice task 1		Holdout choice task 2	
Medical provider	ED	GP	ED	GP	ED	GP
Waiting time to see a doctor	30 minutes	2 hours	2 hours	30 minutes	1 hour	15 minutes
Test facilities (blood test, X-ray, ECG, etc.)	All Available	None Available	All Available	None Available	All Available	All Available
Out-of-pocket payment	S\$200	S\$200	S\$100	S\$50	S\$200	S\$100
GP-referral discount (when GP refers to ED)	None	Entire GP fee waived	None	S\$50 discount on ED fee	None	Entire GP fee waived
Choice	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>

**Figure 1** Example of a random choice task and two holdout tasks.

approach to investigating rationality of DCE subjects is to include non-satiation tests where one alternative is made worse than the other in all attributes.<sup>26</sup> Since our DCE used a labeled design, such validity tests were not suitable because subjects might make inferences based on the identity tag (ED v. GP) on alternatives.<sup>27</sup> Therefore, for validity test in our labeled DCE, we constructed two *holdout* choice tasks,<sup>28</sup> as shown in Figure 1. The first of these represented the status quo, whereas in the second all ED attributes were dominated by those of GP. We added these two holdout tasks to all five choice task blocks. Thus, in each block, there were 10 choice tasks with eight being different across the blocks (*random tasks*) and two being common (*holdout tasks*). As we will explain later, we estimated patient preference models based on their responses to random choice tasks, and used their responses to holdout tasks to validate the estimated preference models.

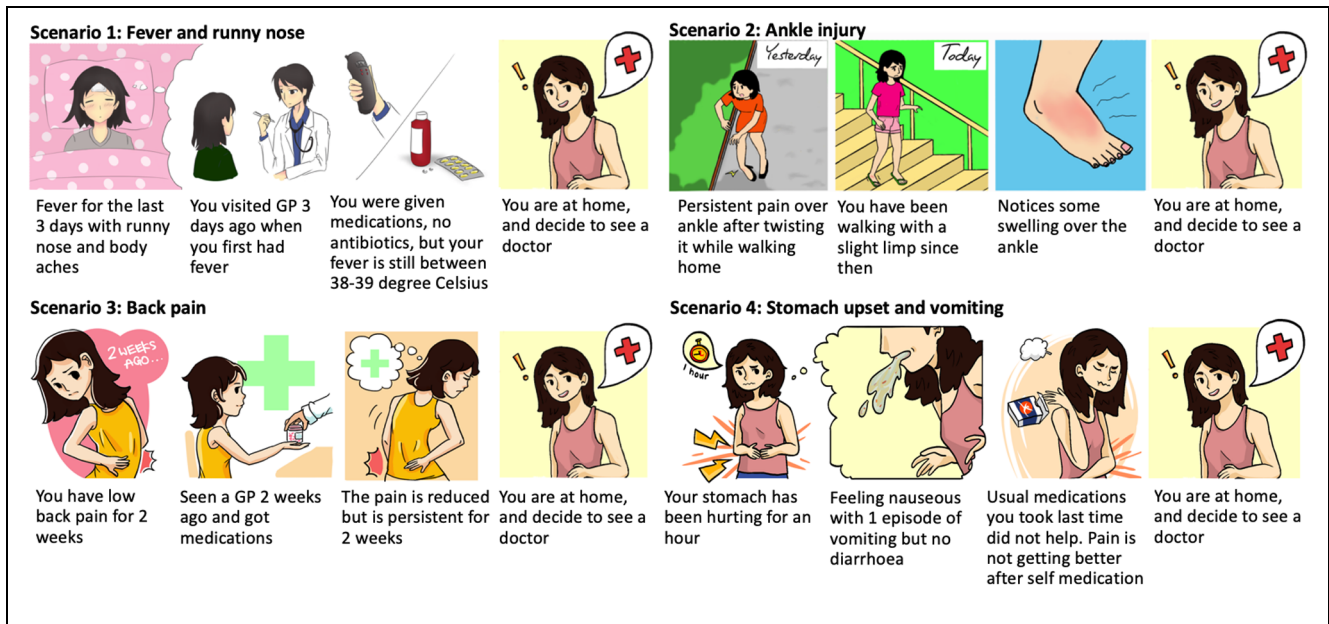
**Survey Design.** The survey was designed to have two parts. The first part contained questions about patients' demographics (age, gender, race, employment status, income range, and travel times to patients' preferred ED and GP). The second part contained the DCE. At the beginning of the DCE, respondents were presented with one of the four *hypothetical medical scenarios* depicted in Figure 2. These scenarios represent the most common cases of nonurgent patient visits to ED, and were developed through discussions with ED physicians of CGH. The survey asked the respondents about their *perception* of the presented scenario (*Do you think this condition is*

*critical enough to go to ED directly*), and their belief about GP referral under that scenario (*Do you think your GP will (a) provide adequate medical treatment for this scenario; or (b) will refer you to a specialist; or (c) will refer you to ED for urgent treatment?*). Respondents were then presented with one of the five choice task blocks described under "DCE design." In each task, they were asked to choose between their preferred ED and GP under the presented conditions, imagining that they faced the presented medical scenario.

To help the respondents properly interpret all attributes before starting with the choice tasks, the DCE also contained several warm-up questions, which were presented before the hypothetical medical scenario and the choice task block. In addition, at the beginning of the choice task block, an example choice task was presented to clearly explain how to interpret a choice task.

### Study Population

The survey was conducted in CGH from December 2017 to February 2018. Respondents were recruited from the ED, specialist outpatient clinics, discharge lounge, and public areas in CGH such as food court and cafes. Because of the reasons described later in "Limitations" section, we used convenience sampling for recruitment. The eligibility criteria for the respondents were 1) Singapore citizen or Permanent Resident, 2) English speaking, and 3) 21 years old or above. Trained research assistants (RAs) screened the eligibility of participants at the beginning of the survey and obtained



**Figure 2** Hypothetical nonurgent medical scenarios.

eligible participants' written consent. The survey was conducted using electronic tablets, and RAs accompanied each respondent throughout the survey to assist in case of any queries.

According to Orme's rule of thumb formula,<sup>29</sup> the minimum sample size required for our survey design is  $157$  respondents ( $500 \times \text{maximum number of levels for any attribute (5 levels corresponding to "waiting time")} \div \text{number of random choice tasks in each survey (8)} \div \text{number of alternatives (2)}$ ), i.e.  $(500 \times 5 \div 8) \div 2$ ). In the survey, a total of 849 respondents were recruited, among which 204 were given scenario 1 (fever and runny nose), 216 were given scenario 2 (ankle injury), 218 were given scenario 3 (back pain), and 211 were given scenario 4 (stomach upset and vomiting). Among these respondents, 7% were recruited from specialist outpatient clinics, 8% from ED, 27% from public areas in the hospital and food court, and 58% from the waiting area at the entrance of inpatients' discharge lounge.

### Analysis

We first estimated patient choice models using latent class multinomial logit regression.<sup>30</sup> Multinomial logit is widely used for analyzing DCE data.<sup>31</sup> With latent class it incorporates preference heterogeneity by identifying (based on choice data) groups of respondents sharing

similar preferences.<sup>23</sup> Thus, it essentially generates different choice models for different groups.

To generate the latent class model we used data from only random choice tasks from the survey. We used effects coding<sup>32</sup> to construct attribute-levels such that in the generated model, the coefficients of all levels within an attribute sum up to zero. For the attribute-level "Entire GP fee waived," the amount of discount depended on the value of GP's out-of-pocket payment (explained under "Study Protocol"). Therefore for our analysis we divided "Entire GP fee waived" into three sublevels: "Entire GP fee waived when GP payment is S\$50," "Entire GP fee waived when GP payment is S\$100," and "Entire GP fee waived when GP payment is S\$200." We conducted all statistical analyses using Sawtooth Software (Latent Class, version 9.5.3, Sawtooth Software, Inc.).

To investigate whether specific respondent characteristics made them more likely to belong to the different classes identified by the latent class model, we performed logistic regression (using R, version 3.4.1), which mapped respondents' class memberships generated by the latent class model to their presented medical scenario, their perception and belief about GP referral for presented scenario, and their demographics collected in the survey. In the appendix we provide mathematical description of our latent class and logistic class membership models.

**Table 2** Attribute-Levels for Status Quo Tasks

	Task 1		Task 2		Task 3		Task 4	
	Short ED Wait				Long ED Wait			
	Day Hour		Night Hour		Day Hour		Night Hour	
Medical provider	ED	GP	ED	GP	ED	GP	ED	GP
Waiting time to see a doctor	30 minutes	30 minutes	30 minutes	15 minutes	4 hours	30 minutes	4 hours	15 minutes
Test facilities	All	None	All	None	All	None	All	None
Out-of-pocket payment	\$\$100	\$\$50	\$\$100	\$\$100	\$\$100	\$\$50	\$\$100	\$\$100
GP-referral discount	No discount	No discount	No discount	No discount	No discount	No discount	No discount	No discount

Next, we tested the validity of the choice model and class membership model generated above. First, we predicted class membership of each respondent using logistic regression, and using the corresponding class' choice model we predicted each respondent's choice under the two holdout tasks. For each task we then computed the proportion of respondents predicted to choose ED. Because holdout tasks were completed by all survey respondents, we also had the actual proportion of respondents from the survey who selected ED under each holdout task. We compared these actual proportions with predicted proportions above, and showed that they are similar, suggesting that patients' preference are reasonably captured by our models.

Last, we predicted uptake of the GP-referral discount scheme and other countermeasures suggested by the choice models. For this we considered four "status quo" tasks from Table 2 that capture differences in GP fees and waiting times between day and night hours, and account for possibilities that patients' decisions to visit ED/GP may depend on "short" or "long" expected ED wait (attribute-level-selection for Table 2 is explained in the appendix). For each status quo task, we predicted the proportion of survey respondents who would choose ED (following similar prediction method as for holdout tasks). We then predicted this proportion under different countermeasures, each of which changed one specific attribute-level in the status quo tasks (e.g., "GP-referral discount" for GP alternative changed to "Entire GP fee waived"). With the above predicted proportions  $P_{ED}^s$  (under status quo) and  $P_{ED}^c$  (under countermeasure), we computed "percentage reduction in nonurgent ED visits" due to the countermeasure as

$$\frac{P_{ED}^s - P_{ED}^c}{P_{ED}^s} \times 100.$$

To predict such reduction for a countermeasure that corrects patients' misperceptions of severity, we generated a "perception-corrected" sample from the survey sample by changing all respondents' "perception" to "not critical" (cf. Table 3). For this sample, we computed the proportion choosing ED under the status quo tasks ( $P_{ED}^{pc}$ ), and compared it against that of survey sample ( $P_{ED}^s$ ) to obtain the corresponding reduction in nonurgent ED visits as

$$\frac{P_{ED}^s - P_{ED}^{pc}}{P_{ED}^s} \times 100.$$

## Results

### *Sociodemographic Profiles and Perception*

Table 3 describes socioeconomic characteristics of the survey respondents, and their perceptions under the presented scenarios. In summary, the average age of respondents was 35.9 years with 46.4% being males and 66.5% Chinese. As we explain later under "Limitations," the survey sample reasonably represents the nonurgent patient population of interest that considers choice between ED and GPs. The survey response showed that 96% respondents knew that ED has priorities in serving patients; and 97% respondents knew that they would have to wait longer at the ED if they were nonurgent.

An interesting observation from Table 3 is respondents' perception of severity for the four scenarios. As per ED physicians, all four scenarios described conditions that could be managed adequately by GPs and would not require urgent treatment at the ED. However, patients' perceptions varied. For example, the percentage of respondents who perceived their presented scenario as "critical enough to go to ED directly" was 34.3% among

**Table 3** Characteristics of Survey Respondents

Patient Characteristics	Full Sample ( <i>n</i> = 849)	Scenario 1 ( <i>n</i> = 204)	Scenario 2 ( <i>n</i> = 216)	Scenario 3 ( <i>n</i> = 218)	Scenario 4 ( <i>n</i> = 211)	Facility-sens' class ( <i>n</i> = 305)	Discount-sens' class ( <i>n</i> = 544)	Class Difference and 95% CI
Age in years, mean (SD)	35.9 (13.3)	38.2 (14.3)	36.6 (13.9)	34.5 (12.4)	34.5 (12.5)	37.6 (14.4)	35.0 (12.6)	2.6 (0.7, 4.5)
Gender, <i>n</i> (%)								
Men	394 (46.4)	90 (44.1)	116 (53.7)	102 (46.8)	86 (40.8)	140 (45.9)	254 (46.7)	-0.8 (-7.8, 6.2)
Race, <i>n</i> (%)								
Chinese	565 (66.5)	131 (64.2)	143 (66.2)	137 (62.8)	154 (73.0)	189 (62.0)	376 (69.1)	-7.1 (-13.8, -0.4)
Malay	202 (23.8)	52 (25.5)	53 (24.5)	60 (27.5)	37 (17.5)	85 (27.9)	117 (21.5)	6.4 (0.3, 12.5)
Indian	60 (7.1)	17 (8.3)	15 (6.9)	14 (6.4)	14 (6.6)	24 (7.9)	36 (6.6)	1.3 (-2.4, 5.0)
Other	22 (2.6)	4 (2.0)	5 (2.3)	7 (3.2)	6 (2.8)	7 (2.3)	15 (2.8)	-0.5 (-2.7, 1.7)
Yearly household income, <i>n</i> (%)								
Less than S\$25,000	205 (24.1)	52 (25.5)	55 (25.5)	50 (22.9)	48 (22.7)	88 (28.9)	117 (21.5)	7.4 (1.3, 13.5)
S\$25,000 to S\$49,999	246 (29.0)	45 (22.1)	66 (30.6)	71 (32.6)	64 (30.3)	76 (24.9)	170 (31.3)	-6.6 (-12.8, -0.4)
S\$50,000 to S\$99,999	241 (28.4)	67 (32.8)	66 (30.6)	54 (24.8)	54 (25.6)	78 (25.6)	163 (30.0)	-4.4 (-10.6, 1.8)
S\$100,000 and above	157 (18.5)	40 (19.6)	29 (13.4)	43 (19.7)	45 (21.3)	63 (20.7)	94 (17.3)	3.4 (-2.1, 8.9)
Employment status, <i>n</i> (%)								
Full-time	577 (68.0)	141 (69.1)	145 (67.1)	148 (67.9)	143 (67.8)	195 (63.9)	382 (70.2)	-6.3 (-12.9, 0.3)
Part-time	102 (12.0)	28 (13.7)	29 (13.4)	24 (11.0)	21 (10.0)	45 (14.8)	57 (10.5)	4.3 (-0.4, 9.0)
Other: retired, homemaker, etc.	170 (20.0)	35 (17.2)	42 (19.4)	46 (21.1)	47 (22.3)	65 (21.3)	105 (19.3)	2.0 (-3.7, 7.7)
Perception, <sup>a</sup> <i>n</i> (%)								
Critical	153 (18.0)	70 (34.3)	23 (10.6)	30 (13.8)	30 (14.2)	76 (24.9)	77 (14.2)	10.7 (5.0, 16.4)
Not critical	696 (82.0)	134 (65.7)	193 (89.4)	188 (86.2)	181 (85.8)	229 (75.1)	467 (85.8)	-10.7 (-16.4, -5.0)
Belief about GP referral, <sup>b</sup> <i>n</i> (%)								
No	733 (86.3)	144 (70.6)	198 (91.7)	204 (93.6)	187 (88.6)	253 (83.0)	480 (88.2)	-5.2 (-10.2, 0.2)
Yes	116 (13.7)	60 (29.4)	18 (8.3)	14 (6.4)	24 (11.4)	52 (17.0)	64 (11.8)	5.2 (-0.2, 10.2)
Travel time to ED – Travel time to GP in minutes, mean (SD)	8.5 (10.9)	9.94 (11.3)	8.5 (12.1)	7.9 (10.6)	7.9 (9.2)	7.5 (10.5)	9.1 (11.0)	-1.6 (-3.1, -0.1)

CI, confidence interval; ED, emergency department; GP, general practitioner.

<sup>a</sup>Perception: For the perception question, “Do you think this condition is critical enough to go to ED directly?” the response *Yes* was coded as “critical”; and *No* was coded as “not critical.”

<sup>b</sup>Belief about GP referral: For the question “Do you think your GP will (a) provide adequate medical treatment for this scenario; or (b) will refer you to a specialist; or (c) will refer you to ED for urgent treatment?” responses (a) and (b) were coded as “No” and response (c) was coded as “Yes.”

those who got the fever-related scenario, but only 10.6% among those who got the ankle injury scenario.

### Choice Models and Class Membership

Our latent class model identified two classes of respondents based on their choices and provided distinct choice models for each class. The model validity test described under “Analysis” generated the following choice percentages, which suggests that patients’ preferences are reasonably captured by these models. The actual proportions of respondents from the survey who chose ED under holdout task-1 and holdout task-2 were 16% and 2.8%, respectively. By predicting respondents’ choices based on the generated models, we obtained the corresponding proportions as 21.3% and 5.5%, respectively.

Table 4 presents parameter estimates for the two-class model. The two classes clearly differ in their sensitivities to

availability of test-facilities and GP-referral discount; hence, we will refer to them as facility-sensitive and discount-sensitive class, respectively. These classes contain 38.5% and 61.5% respondents, respectively. The model indicates that discount-sensitive class has an inherent preference for GP, as their attribute-level coefficient for GP (0.32) is higher than that for ED (-0.32).

To compare the impacts of different ED/GP attributes on the two classes, we plot in Figure 3 the attribute-level coefficients of each class against attribute-levels. For a meaningful comparison of the two classes, we generated these plots using normalized coefficients corresponding to Table 4 (obtained by dividing the coefficients in each class by the absolute value of the out-of-pocket payment coefficient of that class).

Figure 3A and B show combined attribute-level coefficient of GP-referral discount and out-of-pocket payment for the two classes. To see how it is computed, consider



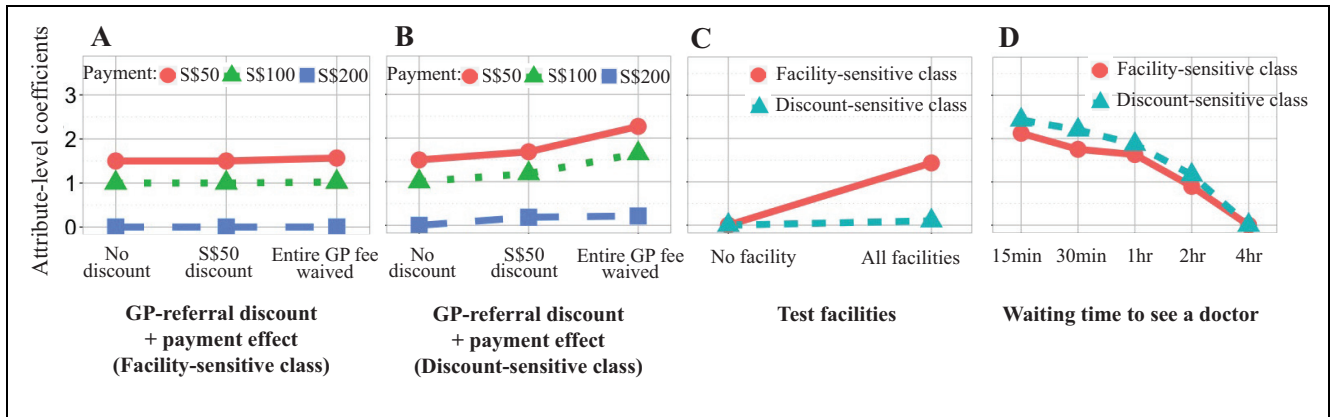
**Table 4** Results From the Discrete Choice Experiment: Latent Class Model

	Facility-Sensitive Class (38.5%)			Discount-Sensitive Class (61.5%)		
	Coefficient (95% CI)	SE	P Value	Coefficient (95% CI)	SE	P Value
<b>Attribute-level coefficients</b>						
ED	0.02 (-0.09 to 0.13)	0.06	0.722	-0.32* (-0.43 to -0.22)	0.05	<0.001
GP	-0.02 (-0.13 to 0.09)	0.06	0.722	0.32* (0.22 to 0.43)	0.05	<0.001
Out-of-pocket payment						
Payment	-1.21* (-1.33 to -1.09)	0.06	<0.001	-1.67* (-1.81 to 1.53)	0.07	<0.001
GP-referral discount						
No discount	-0.02 (-0.20 to 0.16)	0.09	0.819	-0.60* (-0.83 to -0.38)	0.11	<0.001
S\$50 discount on ED fee	-0.02 (-0.21 to 0.16)	0.09	0.824	-0.29 <sup>†</sup> (-0.53 to -0.06)	0.12	0.033
Entire for S\$50 <sup>a</sup>	0.06 (-0.25 to 0.36)	0.16	0.710	0.67 <sup>‡</sup> (-0.04 to 1.39)	0.37	0.092
Entire for S\$100 <sup>a</sup>	0.00 (-0.24 to 0.25)	0.13	0.971	0.47* (-0.12 to 0.82)	0.18	0.022
Entire for S\$200 <sup>a</sup>	-0.02 (-0.33 to 0.29)	0.16	0.897	-0.25 (-0.54 to 0.05)	0.15	0.130
Test facility						
No facility	-0.87* (-0.98 to -0.76)	0.06	<0.001	-0.08 <sup>‡</sup> (-0.18 to 0.01)	0.05	0.104
All facilities	0.87* (0.76 to 0.98)	0.06	<0.001	0.08 <sup>‡</sup> (-0.01 to 0.18)	0.05	0.104
Patient waiting time						
15 minutes	1.11* (0.88 to 1.34)	0.12	<0.001	1.50* (1.28 to 1.73)	0.11	<0.001
30 minutes	0.55* (0.41 to 0.69)	0.07	<0.001	1.12* (0.97 to 1.26)	0.07	<0.001
1 hour	0.40* (0.28 to 0.53)	0.06	<0.001	0.56* (0.44 to 0.68)	0.06	<0.001
2 hours	-0.49* (-0.61 to -0.36)	0.06	<0.001	-0.62* (-0.75 to -0.49)	0.07	<0.001
4 hours	-1.58* (-1.81 to -1.34)	0.12	<0.001	-2.56* (-2.82 to -2.29)	0.14	<0.001

CI, confidence interval; ED, emergency department; GP, general practitioner.

<sup>a</sup>“Entire for S\$50,” “Entire for S\$100,” and “Entire for S\$200” represent “Entire GP fee waived” with GP’s out-of-pocket payment being S\$50, S\$100, and S\$200, respectively, in the choice task.

\*,<sup>†</sup>, and <sup>‡</sup> indicate 1%, 5%, and 10% levels of significance respectively.



**Figure 3** Normalized attribute-level coefficients versus attribute-levels.

this example. For facility-sensitive class, the combined coefficient of “Entire GP fee waived” discount and S\$100 out-of-pocket payment (for GP) is calculated as: *Normalized “Entire for S\$100” coefficient for facility-sensitive class + Coded value for S\$100 payment × (-1)*. Thus, in each Figure 3A and B, the three points corresponding to “no discount” represent combined coefficients of “no

discount” and S\$50/S\$100/S\$200 out-of-pocket payment (for either ED or GP). They imply that higher the ED/GP payment, lower is the utility for ED/GP. The red (solid), green (dotted), and blue (dashed) plots further show how the utility for GP changes with different levels of GP-referral discount, when out-of-pocket payment for GP is S\$50, S\$100, and S\$200, respectively. The figures

**Table 5** Results of Logistic Class Membership Analysis<sup>a</sup>.

	Coefficient (95% CI)	SE	P Value
Facility-sensitive class			
Constant	-0.15 (-0.52 to 0.22)	0.19	0.422
Age <sup>1</sup> : >40 years	0.48 <sup>†</sup> (-0.19 to 0.78)	0.15	0.001
Employment <sup>2</sup> : PartTime	0.41 <sup>‡</sup> (-0.03 to 0.84)	0.22	0.068
Employment <sup>2</sup> : EmployOther	0.15 (-0.21 to 0.52)	0.18	0.402
Travel time to ED – Travel time to GP	-0.01 <sup>‡</sup> (-0.03 to 0.00)	0.01	0.067
Perception <sup>3</sup> : NotCritical	-0.71* (-1.07 to -0.35)	0.18	<0.001
Log-likelihood	-538.1		

CI, confidence interval; ED, emergency department; GP, general practitioner.

<sup>a</sup>Level of reference: 1, Age: 21–40; 2, Employment: full-time; 3, Perception: critical.

\*, †, and ‡ indicate 1%, 5%, and 10% levels of significance, respectively.

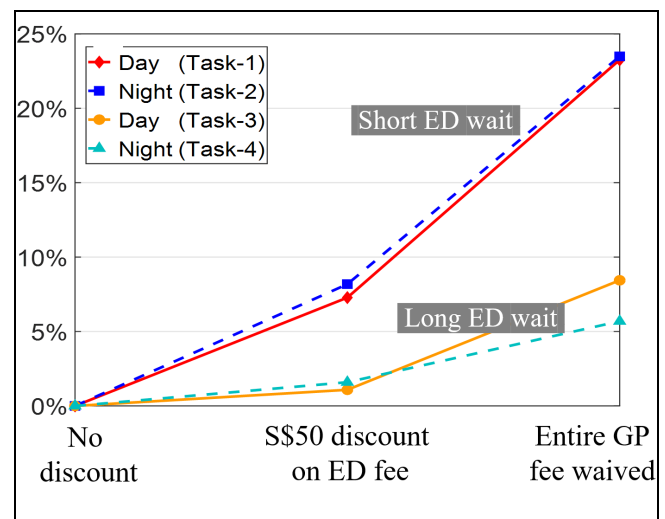
imply that GP-referral discount has very mild influence on facility-sensitive class, but has significant impact on discount-sensitive class, especially when GP payment is S\$50 or S\$100.

Figure 3C shows test-facility-level coefficients for the two classes. It implies that with test-facility provision, utility for GP significantly increases for the facility-sensitive class, but remains almost unchanged for discount-sensitive class.

Figure 3D shows that effects of waiting time to see a doctor (at either ED or GP) are similar on both classes—both classes prefer shorter waiting times, with their attribute-level coefficients decreasing almost linearly in waiting time.

The logistic class membership model in Table 5 shows that among all patient characteristics collected in the survey, only age, employment status, difference in travel times to ED and GP, and patients' perception of severity significantly influence their class membership. It shows that patients younger than 40 years or with longer travel time to ED are more likely to belong to the discount-sensitive class. This implies, based on the choice model from Table 4, that such patients have higher inherent preference for GP and are more sensitive to GP-referral discount. Patients older than 40 years or with part-time employment exhibit higher likelihood of belonging to the facility-sensitive class. This implies their higher sensitivity to test facilities than GP-referral discount, which may be because older patients may care more about health diagnosis than cost. Also, since part-time workers have more time, they may be willing to wait longer at the ED, which has all test-facilities.

An important observation from Table 5 is that patients' perceptions of their severity significantly influence their class membership and thus heterogeneity of preferences. Patients who perceived their presented scenario as “critical

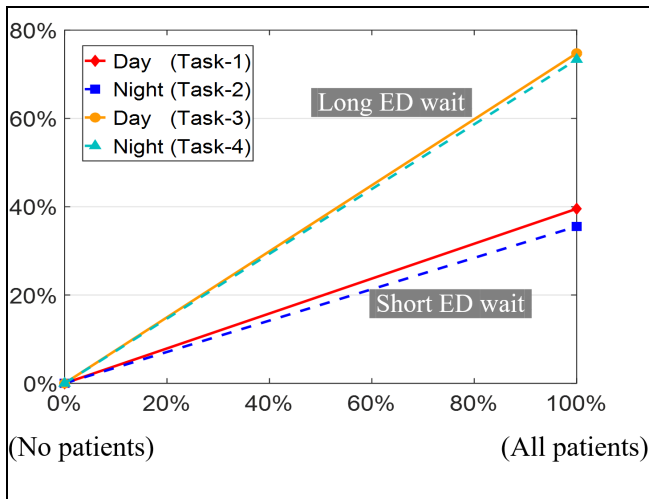


**Figure 4** Percent reduction in nonurgent ED visits versus “GP-referral discount.”

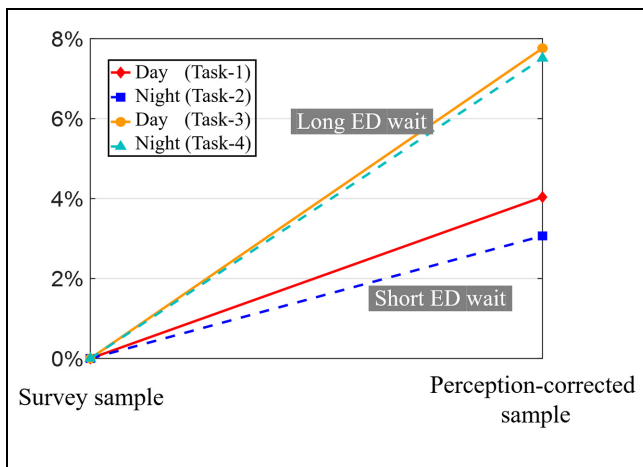
enough to go to ED directly” exhibit much higher likelihood of belonging to the facility-sensitive class, and hence of preferring ED. Another important observation from this model is that the “Scenario” variable is not significant, indicating that the effects of the different nonurgent scenarios on patients' preference are sufficiently captured by patients' perceptions (“critical” v. “not critical”) of those scenarios.

### Predicted Uptake of Countermeasures

Figures 4 to 7 show percentage reduction in nonurgent ED visits due to the different countermeasures, as per our predictive analysis described in “Analysis” section. In all figures, 0% reduction corresponds to the status quo attribute-level (e.g., “no discount” in Figure 4), and other



**Figure 5** Percent reduction in nonurgent ED visits versus % of patients whose GPs have “all test facilities.”



**Figure 6** Percent reduction in nonurgent ED visits due to perception correction.

attribute-levels on  $x$ -axis represent countermeasures. A common observation for all countermeasures is that their uptake is similar for day versus night hours if, under both situations, patients visit ED expecting the same wait. However, if patients visit ED expecting “short” versus “long” waits, then uptakes are quite different.

Figure 4 shows that the effect of GP-referral discount is stronger when patients visit ED expecting “short” wait, with 23.5% reduction in nonurgent ED visits in the best case. Figures 5 and 6 show that test-facility provision at GPs and correcting patients’ misperceptions of severity have stronger effect when patients visit ED expecting

“long” waits, with best case reductions of 74.8% and 7.8%, respectively. Finally, Figure 7 shows that publishing ED wait information could reduce up to 76.2% nonurgent ED visits if actual waits are much higher than current beliefs (Figure 7A); however, it could also drastically increase nonurgent ED visits if actual waits happen to be lower than common beliefs (Figure 7B).

## Discussion

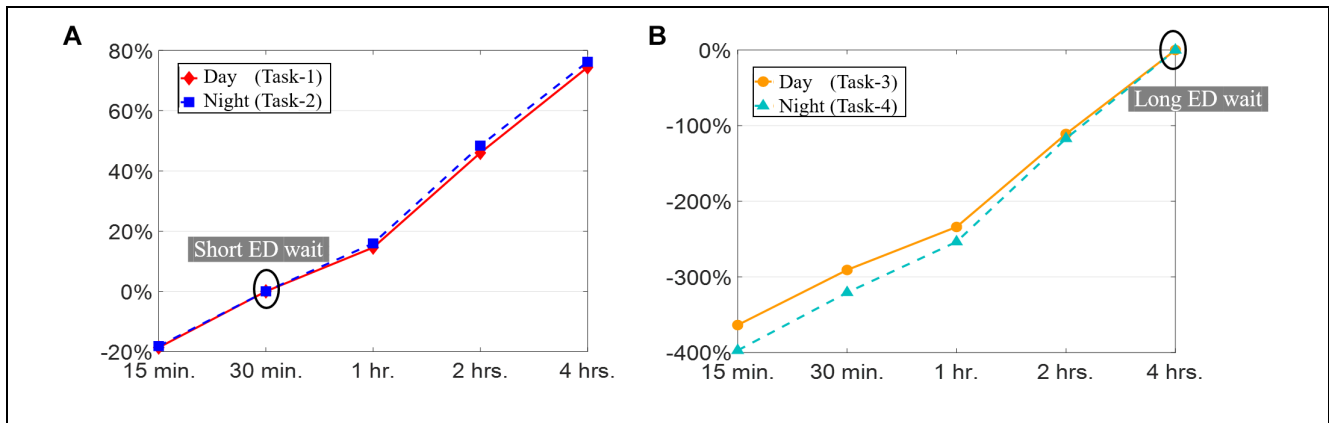
We now discuss policy implications of our results. The latent class multinomial logit model presented in the previous section identified two patient classes. Among these, for the discount-sensitive class, introduction of “GP-referral discount” significantly increases preference for GPs. In particular, providing entire GP fee waiver upon referral to ED appears to be quite effective with standard day/night GP charges in Singapore (represented by S\$50/S\$100 in the survey). Though the discount shows very mild influence on facility-sensitive class, the large size of discount-sensitive class suggests that hospitals can benefit overall from such incentive schemes for diverting nonurgent patients from ED to GPs.

Our patient choice model also suggests other potential countermeasures that may be useful in reducing nonurgent ED visits, which we discuss below.

The model shows that patients are heterogeneous in their preferences for test-facility-availability, with facility-sensitive class being highly sensitive to it and discount-sensitive class not. Given significant size of the former class, increasing test facilities at GPs can also help in reducing nonurgent ED visits. This observation is similar to those of concurrent DCE studies<sup>19,20</sup> conducted in China, which also found that improvements to medical equipment/skill at primary care level could be effective in directing patients to primary care.

Previous DCE studies<sup>12,19,20</sup> that explored patients’ preferences between ED and GP have identified waiting time as an important attribute. Our model corroborates this as it shows that both patient classes are highly time-sensitive and have clear preferences for shorter waiting time. Currently, hospitals in Singapore as well as in many other countries do not publish ED wait information online. Since patients are highly time-sensitive, information made available about long ED waits should discourage nonurgent patients from visiting ED. According to the follow-up questions to our DCE, 73% respondents would like to have a mobile application for ED estimated waiting times.

The logistic class membership model shows that patients who perceived their presented nonurgent scenarios as “critical enough to go to ED directly” have significantly



**Figure 7** Percent reduction in nonurgent ED visits versus patients' belief on "ED wait" (based on published ED waits): Circled point represents assumed patients' belief under status quo.

higher preference for ED. Furthermore, patients' "perceptions" sufficiently capture preference heterogeneity due to the presented nonurgent scenarios. These results suggest that measures to correct patients' misperception of their ailments may also be useful. In our survey, 77% respondents said that they would like to have a call facility or application which can help them assess their symptoms prior to arrival at ED.

The predictive analysis further shows that uptake of GP-referral discount would be higher when patients visit ED expecting "short" wait, whereas uptakes of test-facility provision at GPs and perception-correction would be higher when patients visit ED expecting "long" waits, because patients' choice of ED despite expecting long wait implies their strong preference for test facilities or misperception of severity. Publishing of ED wait information, on the other hand, can have good uptake if actual ED waits are much higher than patient' beliefs, but it could also drastically increase nonurgent ED visits if actual waits happen to be lower than common beliefs. Therefore, before considering this measure, it is important to understand how patients' beliefs compare to actual ED waits.

Overall, the predictive analysis shows that compared to other countermeasures, test-facility provision at GPs has much higher uptake, whereas perception-correction has much smaller uptake as it only influences preferences of those with "critical" perception, which is a relatively small proportion of patients. Interestingly, for GP-referral discount the analysis shows that, for the same GP fee of S\$50, describing the discount as "entire GP fee waived" as opposed to "S\$50 discount on ED fee" is much more effective, which indicates that describing the

incentive as "entire GP fee waiver" (possibly with minimum discount of S\$50) may be a more effective strategy for further implementation of the scheme.

### Limitations

There are some potential limitations of our study that are worth mentioning. The first limitation comes from the use of DCE which uses hypothetical scenarios. It could be argued that respondents may behave differently under real versus hypothetical situations.<sup>33,34</sup> Nevertheless, to minimize such differences, we designed the attributes and levels in our survey as close as possible to those in a real-world situation. Also, though we emphasized to make choices based on the hypothetical situation, it is possible that some respondents made choices based on their past medical experience. To avoid misinterpretation of attributes, we constructed several warm-up questions to help the respondents properly understand all attributes before starting with choice tasks.

The other limitation relates to sampling of participants as we do not know the profile of real patient population that considers choice between ED and GPs, in particular for the "nonurgent" conditions of interest (those constituting nonurgent ED demand). Although nonurgent patient statistics were available for CGH-ED, it was not available for GPs as GPs in Singapore do not record such statistics. Because of this nontriviality we chose to use convenience sampling to recruit participants. However, after building the choice model, we predicted the distribution of patients from the survey sample that would choose ED under status quo ED/GP attribute-levels (from Table 2). We found that this is comparable

to the real profile distribution of nonurgent patients at CGH-ED, which implies that the survey sample is a reasonable representation of the patient population of interest.

## Conclusions

This study showed that the new GP-referral financial incentive introduced in Singapore can be effective in reducing nonurgent ED visits, if it reasonably covers the (extra) cost of visiting a GP. The ease of implementation of such financial incentive makes it a more viable solution compared to other infrastructure-heavy solutions such as test-facility-provision at GPs or perception-correction measures. In Singapore context, however, our results implied that increase in GP test facilities can have much bigger overall impact; hence, it may be worth considering for a long-term solution. In fact, the financial incentive may serve as a complement to the other two measures, as the former and the latter appear to influence distinct classes (discount-sensitive and facility-sensitive) of patients.

Results of this study should be of interest to health care professionals in other places too, because GP-referral financial incentive can be easily adapted to other systems. Compared to financial disincentives such as ED co-payment/fee increase that have been tried in some countries in the past, this novel incentive has the advantage that it does not impose penalty for ED use. Hence, it can provide a better/feasible alternative where ED fee increase may not be preferable/feasible.

## Appendix

### Focus Group Discussions (FGDs)

We conducted five FGD sessions with different subgroups of recruited participants. The sessions were conducted by an experienced moderator using a topic guide, which consisted of a list of open-ended questions pertaining patients' health seeking behavior. Each session lasted between 1 and 1.5 hours, and data saturation was obtained after five sessions. All discussions were audio recorded and transcribed verbatim, and then were analyzed using inductive thematic analysis approach.<sup>35</sup>

Based on the themes emerged from FGDs, we identified the salient factors that affect patients' choices between ED and GP, and grouped them in two categories: 1) *Service attributes* of ED and GP, which include patient's waiting time to receive treatment, availability of test facilities (blood test, X-ray, ECG, etc.), and cost of care (in particular the out-of-pocket payment after

deductions of any insurance or subsidies); and 2) *Patient characteristics*, which include their perception of severity based on observed symptoms, their belief about doctors' experience, and their travel time to ED/GP. The above-mentioned factors are consistent with findings of previous qualitative studies in other countries.<sup>15</sup> The "service attribute" factors identified above were included as ED/GP *attributes* in our DCE design, as shown in Table 1. The "patient-characteristic-related" factors were included in other part of the survey, as explained under the section "Survey Design."

### Attribute-Level Selection for DCE

For our DCE, the ED/GP-specific attribute-levels shown in Table 1 were chosen as follows.

- *Patients' waiting time to see the doctor:* We chose its levels to include the range of nonurgent patients' waiting times at EDs and GPs in Singapore. Across different EDs, the daily\* median waiting time (to see the doctor) of the lower-urgency P3 patients ranges from 47 to 68 minutes, and their 95th percentile waiting time ranges from 97 to 221 minutes.<sup>36</sup>
- *Test facilities:* In Singapore, EDs have laboratory and radiology facilities with a short turnaround time for the results. Most solo GP practices do not have these ancillary facilities on site. Therefore, to simplify our DCE design, we set the "test-facilities" level for ED as "All available," and varied its levels for GP between "None available" and "All available."
- *Out-of-pocket payment:* Hospitals in Singapore charge a flat ED fee (differing a little between hospitals), which covers consultation, medication, and all tests done at the ED. In 2017, ED fee at public hospitals was around S\$100. On the other hand, GPs' consultation fees in Singapore range from less than S\$50 to more than S\$120 depending on the time of the day (night charges being much higher than day charges),<sup>37</sup> and additional fees apply for medication and any tests (such as X-ray or blood tests) that may be required. To cover the above range, we chose three levels—S\$50, S\$100, and S\$200—of "out-of-pocket payment" for GP. Despite the relatively uniform ED fee, we also chose (same) three levels of out-of-pocket payment for ED to be able to investigate sensitivity of patients' preference with respect to (w.r.t.) out-of-pocket payment, and to be able to adapt to any future changes in ED fee.

\*Averaged between January 2013 and June 2016.

- *GP-referral discount*: Under the GP-referral discount scheme piloted in Singapore, patients who directly go to the Public Hospital ED pay the regular ED fee and are not eligible for the “GP-referral discount.” Therefore, in our DCE, we kept the “GP-referral discount” level for ED alternative (i.e., for visiting ED directly) as “No discount.” The above pilot scheme provides “GP-referral discount” to patients who visit a participating GP and are referred by the GP to the Public Hospital ED. If those patients are not referred to the ED, they do not receive any discount. Though the pilot scheme offers S\$50 discount on ED fee,<sup>3</sup> some patients suggested during our FGDs that they would prefer a scheme that waives the entire GP fee (if GP refers the patient to ED) as GP fees can be high, especially during night hours. Hence, to capture this suggestion as well as the pilot scheme offer and the status quo, we chose three levels of “GP-referral discount” for the GP alternative—“entire GP fee waived,” “S\$50 discount on ED fee,” and “no discount.” In our DCE, if a choice task showed GP’s out-of-pocket payment as S\$50 (respectively, S\$100, S\$200) and GP-referral discount as “entire GP fee waived,” it implied that the patient would get back her S\$50 (respectively, S\$100, S\$200) GP payment if she chose to visit GP and the GP referred her to ED.

### *Latent Class Multinomial Logit Choice Model and Logistic Class Membership Model*

Here we provide mathematical description and interpretation of the latent class (LC) multinomial logit (MNL) choice model and logistic class membership model in our context.

For an LC-MNL choice model with  $C$  classes and two choice alternatives  $j, k \in \{ED, GP\}$ , the probability that individual  $i$  chooses alternative  $j$  can be expressed as

$$P_{ij} = \sum_{c=1}^C Pr(i \text{ belongs to class } c) \times Pr(\text{choice} = j | \text{class} = c) = \sum_{c=1}^C \pi_{ic} \frac{e^{V_j^c}}{e^{V_j^c} + e^{V_k^c}}, \quad (1)$$

where  $\pi_{ic}$  denotes the probability that individual  $i$  belongs to class  $c$ . In our context, for each class  $c$  and  $j \in \{ED, GP\}$ ,  $V_j^c$  in (1) can be written as

**Table 6** Characteristics of Focus Group Discussion (FGD) Participants

Characteristics	Patients ( $n = 20$ )
Age, $n$ (%)	
21–30	7 (35)
31–40	2 (10)
41–50	4 (20)
51–60	1 (5)
≥61	6 (30)
Gender, $n$ (%)	
Male	13 (65)
Female	7 (35)
Race, $n$ (%)	
Chinese	11 (55)
Malay	7 (35)
Indian	1 (5)
Others (Eurasian)	1 (5)
Education, $n$ (%)	
Primary	0 (0)
Secondary	6 (30)
A-level/polytechnic/diploma/ITE	9 (45)
University and above	5 (25)
Current occupation, $n$ (%)	
Professional, executive and managerial	3 (15)
Production, technical and mechanical	2 (10)
Clerical and administrative	1 (5)
Housewife	1 (5)
Others: Retirees, NS men, etc.	13 (65)
Monthly personal income, $n$ (%)	
< S\$1,000	3 (15)
S\$1,000 to S\$2,999	5 (25)
S\$3,000 to S\$3,999	1 (5)
S\$4,000 to S\$4,999	0 (0)
≥ S\$5,000	1 (5)
Decline to answer	4 (20)
NA	6 (30)
Dwelling type, $n$ (%)	
HDB (1–2 rooms)	0 (0)
HDB (3 rooms)	2 (10)
HDB (4 rooms)	6 (30)
HDB (5 rooms or more)	3 (15)
Condominium/private flat	1 (5)
Others: e.g., Landed	3 (15)
Decline to answer	5 (25)

$$V_j^c = \beta_j^c + \beta_{\text{Out-of-pocket payment}_j}^c \times \text{Out-of-pocket payment}_j + \beta_{\text{GP-referral discount}_j}^c + \beta_{\text{Test Facility}_j}^c + \beta_{\text{Waiting time to see doctor}_j}^c, \quad (2)$$

where  $\beta_{(\cdot)_j}^c$  are *attribute-level coefficients* for class  $c$  corresponding to the attribute-levels of alternative  $j$ . In

LC-MNL model estimation, the *attribute-level coefficients*  $\beta$  and *class membership probability*  $\pi_{ic}$  of each individual are estimated simultaneously to maximize the log-likelihood. This estimation is done by using only respondents' *choice data* (responses to choice tasks) without incorporating their individual characteristics. For our model, estimated coefficients  $\beta$  are presented in Table 4.

The logistic class membership model uses individual characteristics of respondents to further describe which type of respondents are more likely to belong to the different classes generated by the LC-MNL model. In our context, with dependent variable being the class membership (facility-sensitive class v. discount-sensitive class) generated by the LC-MNL model, this logistic regression can be represented as follows. Let  $\pi_{if}$  denote the probability that individual  $i$  belongs to the facility-sensitive class, then

$$\log\left(\frac{\pi_{if}}{1-\pi_{if}}\right) = \alpha_0 + \alpha_{Demographic_i} + \alpha_{Scenario_i} + \alpha_{Perception_i} + \alpha_{Belief-about-referral-from-GP-to-ED_i} \quad (3)$$

In (3),  $\alpha_0$  is the constant term,  $Demographic_i$  represents respondent  $i$ 's demographic profile (implying one coefficient for each demographic characteristic—age, gender, race, household income, employment status, and difference between respondent's travel times to her preferred ED and preferred GP),  $Scenario_i$  represents the hypothetical medical scenario presented to respondent  $i$ ,  $Perception_i$  represents respondent  $i$ 's perceived severity of the presented medical scenario, and  $Belief-about-referral-from-GP-to-ED_i$  represents respondent  $i$ 's belief about being referred to ED by the GP under the presented scenario. For our model, estimated coefficients  $\alpha$  are presented in Table 5.

### Attribute-Level Selection for Status Quo Tasks for Uptake Analysis

We explain here how we constructed the four status quo tasks in Table 2. In Singapore, GP fees and waiting times usually differ between day and night. Therefore, we constructed different status quo tasks to represent day and night situations. Based on standard GP fees in Singapore,<sup>37</sup> we selected the “out-of-pocket payment” for GP to be S\$50 and S\$100 for the day and night tasks, respectively. For “waiting time to see a doctor” at the GP, we selected the levels 30 minutes and 15 minutes for day and night tasks, respectively, as GPs have lighter load during night hours. Since waiting times at ED are more

unpredictable and patients may make their choices based on their own beliefs about ED waiting times (as they are not published online), we considered two extreme possibilities—that patients may make their decision based on “short ED wait” experience, or “long ED wait” experience for nonurgent visits. To approximate these two situations, we selected “waiting time to see a doctor” at the ED as 30 minutes and 4 hours, respectively.<sup>36,\*</sup> Considering the two extreme possibilities helps in analyzing sensitivity of our uptake analysis. Thus we constructed four status quo tasks in total to represent “day” versus “night” hours and “short” versus “long” expected ED waits, as shown in Table 2. According to a prior survey conducted at the Public Hospital in 2016, most of the patients were not aware of the GP-referral discount pilot scheme. Therefore, to simulate patients' decision making, we set the GP-referral discount (for GP alternative) as “no discount” for all four status quo tasks. Finally, since most solo GP practices do not have test-facilities on site, we set GP's test-facility as “None available” for all four status quo tasks.

### Acknowledgments

We would like to thank the administration of Changi General Hospital, Singapore, for their support in the recruitment of study participants.

We would also like to thank Dr. Adam Oei, Mr. Tony Lim, Ms. Suxin Liaw, and Mr. Andrew Tan for survey administration, and Professors Yow Wei Quin, Bikramjit Das, and Costas Courcoubetis from Singapore University of Technology and Design for their valuable feedback during study design.


### Authors' Note

- *Institutions where work was done:* The work was done at the Engineering Systems and Design Pillar, Singapore University of Technology and Design, Singapore, and Health Services Research Department, Changi General Hospital, Singapore.
- *Presentations:* The work presented in this manuscript has been orally presented at the following conferences. These conferences do not have any published proceedings.
- POMS-HK International Conference, Hong Kong, 2020 (POMS-HK Chapter; January 4–5, 2020; HKU Business School).

\*Across different EDs in Singapore, the daily (averaged between January 2013 and June 2016) median waiting time (to see the doctor) of the lower-urgency P3 patients ranges from 47 to 68 minutes, and their 95th percentile waiting time ranges from 97 to 221 minutes.<sup>36</sup>

- INFORMS Healthcare Conference 2019 (INFORMS; July 27–29, 2019; MIT, Cambridge, MA).
- INFORMS MSOM Conference 2019 (INFORMS; July 1–2, 2019; National University of Singapore, Singapore).

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### Supplemental Material

Supplementary material for this article is available on the *Medical Decision Making Policy & Practice* website at <https://journals.sagepub.com/home/mpp>.

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