

1 **REDUCING ROAD CONGESTION THROUGH INCENTIVES: A CASE STUDY**

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1 ABSTRACT

2 Traffic congestion is a burden on modern society: hundreds of billions of dollars are wasted each
3 year due to extra fuel consumed, wasted time of commuters, traffic accidents, etc. Congestion
4 occurs when the demand for transport capacity exceeds supply either in a sustained manner each
5 day or just during peak hours. In the former case it is necessary to add transport capacity,
6 whereas adding capacity to combat peak-hour congestion can be very expensive and create a
7 vicious cycle which attracts “latent demand” to fill in the added capacity. Peak-hour demand
8 must therefore be actively managed. Currently, several cities and agencies employ congestion
9 charging which, while effective, often lacks popular support. In this paper we describe an
10 approach that uses incentives to nudge commuter behavior towards congestion-reducing times
11 and modes of travel. Specifically, we describe an incentive program, CAPRI (Congestion And
12 Parking Relief Incentives), which aims to reduce peak-hour traffic in Stanford University.
13 CAPRI rewards commuters who drive during off-peak hours and those who walk or bike to
14 work. Commuter behavior is monitored using RFID sensors for automobiles and a smartphone
15 app—*My Beats*—for walkers and bikers. A commuter who signs up for CAPRI earns points for the
16 “good trips” she makes, and these points can be redeemed for rewards (both monetary and
17 in-kind) in a fun, online game. CAPRI also nudges commuters through features like personalized
18 offers, social influence and status. The paper describes the program, highlighting the effect of
19 each of these features in nudging commuter behavior.

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24 *Keywords:* Traffic Congestion, Travel Demand Management, Incentives, Personalization, Social
25 Influence, Behavior Shift

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1 INTRODUCTION

2 Traffic congestion is a serious problem occurring every day in cities around the world.
3 According to a recent study (1), congestion-related costs surpassed \$121 billion in the US in
4 2011, and this figure is projected to jump to \$199 billion by 2020. The consequences of traffic
5 congestion include increased air pollution, extra fuel consumption, increase in traffic accidents,
6 and the diversion of commuters' time from more desirable pursuits.

7 Given the severity of the problem, cities and transport agencies are employing various
8 methods of addressing congestion. The methods can be classified into two broad categories:
9 increasing transport capacity and curbing the demand. It is necessary to increase transport
10 capacity when the demand exceeds the supply in a sustained fashion and this is expensive; for
11 example, Illinois has a plan underway (2) to widen tollways in the Chicago region at a projected
12 cost of \$12 billion. Adding capacity to address just peak-hour congestion can not only be
13 prohibitively expensive, it can create a vicious cycle of attracting additional peak traffic—the
14 so-called “latent or induced demand”—and drive the transport system back to a state of
15 congestion (3). Hence, peak demand must be actively managed and reduced. Several demand
16 management schemes have been studied and deployed; notably, (a) congestion
17 charging—levying a charge on peak-hour commuters, (b) road rationing—prohibiting road use
18 for vehicles on some days depending on license plate numbers, (c) carpool lanes—restricting
19 highway lanes for single occupancy vehicles, (d) ridesharing and vanpooling—encouraging
20 commuters to share vehicles, and (e) travel mode-shifting programs—encouraging commuters to
21 use public transit or bikes. The effectiveness of these measures ultimately rests on the
22 commuters' willingness to support and adopt them.

23 This paper studies the use of incentives to increase the willingness of commuters at
24 Stanford University for adopting certain congestion-reducing behaviors. Specifically, we aim to
25 reward commuters who travel during off-peak hours (in the shoulder hours adjacent to the
26 morning and evening peak hours) and those who walk or bike to work. Thus, in terms of
27 reducing peak traffic, our approach can be viewed as the complement of congestion charging
28 schemes. Vickrey (4) was an early and remarkably prescient proponent of congestion charging.
29 He pointed out that road congestion is an example of the “tragedy of the commons (5)” effect: a
30 free public good (namely, the road) will tend to suffer from overuse. Therefore, he argued,
31 congestion charges need to be levied to operate road networks at loads well below the point at
32 which they are congested. Several cities (6,7,8) have implemented “congestion charging”
33 mechanisms; essentially, drivers pay a fee for entering *congestion zones* or using pre-designated
34 roadways during peak hours. However, congestion charging schemes can suffer from a lack of
35 popular support and thereby fail to be implemented in the first place (see (9), for example).

36 By being a “carrot” rather than a “stick”, incentive approaches are viewed favorably by
37 commuters and, therefore, enjoy certain advantages over congestion charging schemes.
38 Participation in incentive programs is voluntary and on an opt-in basis. Therefore, incentive
39 programs are incrementally deployable, i.e. it is not necessary to include all commuters at the
40 outset. On the other hand, for reasons of fairness, congestion charging schemes must include all
41 the concerned commuters. Incentive programs invite commuters to reveal their good commuting
42 behavior so they may claim rewards rather than go to great lengths to hide their bad behavior so
43 as to avoid charges and penalties. This can considerably reduce the burden of enforcement.

44 The major questions for a commuter incentive program are: (i) is it effective in inducing
45 the right behavior, and (ii) if so, at what monetary cost? An experiment conducted at a
46 reasonable scale is necessary to answer these questions. This paper describes such an

1 experiment, called CAPRI (Congestion And Parking Relief Incentives), which we have designed
2 and implemented at Stanford University (10). The aims of CAPRI are to nudge Stanford
3 University drivers away from peak-time travel, to park at underutilized parking lots and to adopt
4 alternate modes such as biking and walking. The nudges are administered through a
5 combination of monetary rewards, social influence and personalized recommendations.

6 CAPRI was launched in April 2012 and invited the participation of about 10,200
7 permit-holding car commuters at Stanford; of these, 3,082 registered for the program. An
8 additional 975 registered for the walking and biking incentives using a smartphone app. We
9 describe the main elements of the CAPRI program and the technological platform, which is a
10 combination of RFID and smartphone-based sensing of commuter movements, and a
11 cloud-based incentive or “nudge” engine. CAPRI builds on the INSTANT (11), Steptacular (12)
12 and Insinc (13) programs in terms of the behavioral interventions and technological elements.
13 The rest of the paper analyzes the commuter behavior shift in detail. We summarize the main
14 findings as follows: compared to the general Stanford population, CAPRI participants are 21.2%
15 less likely to commute during the morning peak hours of 8–9AM, and 13.1% less likely to
16 commute during the afternoon peak hour of 5–6PM. CAPRI also successfully converted tens of
17 commuters from driving to walking or biking to the campus. Through a regression analysis, we
18 show that incentives have played an important role in shifting the commuting behavior of CAPRI
19 participants over time.
20

21 2. LITERATURE REVIEW

22 *Incentive programs for reducing congestion.* Spitsmijden (Dutch for “peak avoidance”) is a
23 series of commuter incentive program conducted in the Netherlands to reward drivers for driving
24 during off-peak hours. It was shown that positive incentives were able to reduce the amount of
25 peak traffic by around 60% (14,15,16) in the test population of 340 participants. The rewards in
26 the ten-week program were high: from 3 to 7 euros per avoided peak-hour trip. Sensors were
27 deployed on the road system and in the vehicles of the participants for behavior detection. An
28 experiment in Osaka, Japan gave drivers a one-month free bus ticket. It was observed that drivers
29 who were given free bus tickets took buses more frequently than before, but the effect did not
30 persist after the trial (17). A similar study in Germany investigated the effect of offering pre-paid
31 bus tickets to college students (18). In Melbourne, Australia, an incentive program offered free
32 rail fares to rail travelers who completed their trips before 7AM in order to alleviate the
33 overcrowding issue during morning peak hours (19). A similar program is currently underway in
34 Singapore (20).

35 A recent mobile traffic app, Smartrek, was designed to predict the length of commute
36 time with different departure times, and give varying levels of reward credits to users choosing
37 different departure times (21). The prediction of future traffic conditions and suggestion of routes
38 used both historical and real-time traffic data. The program ran for 10 weeks and gathered 308
39 completed observations. It was found that users were willing to change departure times and
40 routes for as many as 35% of their trips.
41

42 *Incentive programs for influencing driver behavior.* There exists research that focuses on using
43 insurance-based incentives to influence the driving behavior of participants. Usage-based
44 insurance (UBI), for example, uses devices plugged into a car’s diagnostic port to record and
45 analyze information such as distance driven, the time at which the trip was undertaken, and
46 unsafe driving behaviors (e.g., hard braking). These data are used to improve the risk profile of

1 the insured and offer potential discounts in insurance premiums. One experiment showed that
2 drivers who were offered mileage-based insurance premium on average drove 5%, or 560 miles,
3 less per year (22).

4 Since the advent of smartphones, there are an increasing number of applications of
5 smartphone apps to monitor commutes and encourage environmentally friendly traffic modes.
6 UBIGreen is a smartphone app that senses users' transportation behaviors and gives graphical
7 feedback (23).

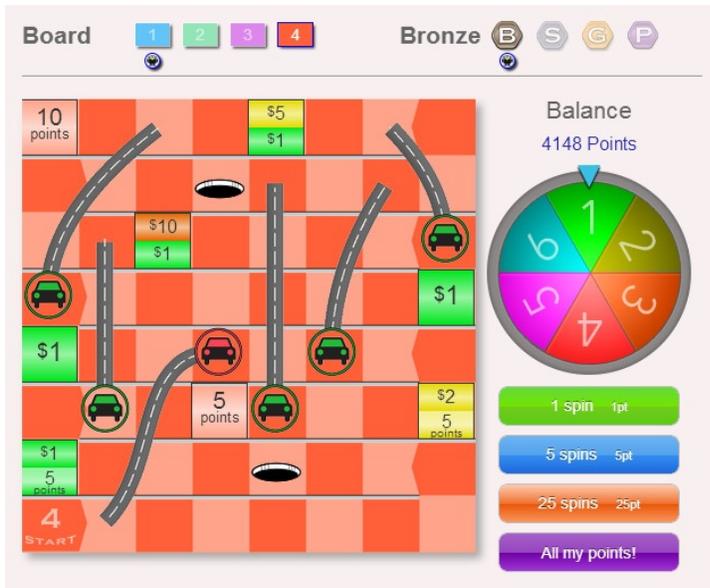
8
9 As compared to these programs, the scale of the CAPRI program is higher in terms of the
10 number of participants (excluding the Melbourne and Singapore public transit early travel
11 incentive programs), the number of trips observed, and the duration of the project. Secondly,
12 and in common with some of the above programs, CAPRI has made extensive use of technology
13 for sensing (a custom-designed RFID system, and mobile app) and for interaction with the
14 participants (an "always on" online portal). Finally, CAPRI gives monetary rewards through a
15 lottery-like random reward mechanism rather than deterministically. This has allowed CAPRI to
16 use a budget of just \$1/participant/week (or 10 cents per morning or evening trip) for nudging.
17

18 **3. DESIGN AND IMPLEMENTATION OF CAPRI**

19 **Background and motivation**

20 Stanford University is one of the largest employers in the San Francisco Bay Area (24), and its
21 growing population has led to an increase in traffic volume. This has caused congestion on the
22 roads around the campus and strained the parking resources on campus. Stanford University
23 signed a General Use Permit (GUP) with the County of Santa Clara in 2000 (25). One of the
24 provisions in GUP states that Stanford is required to manage its transportation impact under a
25 "no net new commute trips" standard: the amount of traffic during peak hours must not increase
26 by more than 1% over 3,319 vehicles in the morning peak hour, and 1% over 3,446 vehicles in
27 the afternoon peak hour (these numbers were measured in 2000). Failure to adhere to this
28 standard has severe consequences for Stanford, including fines and the refusal of building
29 permits for new construction projects.

30 Existing efforts to mitigate the traffic in Stanford include free campus shuttles from and
31 to nearby transit stations, cash incentives (\$300/person/year) to off-campus staff and students
32 who do not buy parking permits on campus, subsidized public transit passes, and discounts on
33 parking permits for carpooling commuters. These measures have been effective in reducing the
34 total number of commuters who drive alone. It does not directly address peak-hour commuters
35 whose number has been increasing over time. The CAPRI program directly addresses peak-hour
36 commuters and aims to reduce their number.
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1

2 **FIGURE 1 Chutes-and-ladders game for redeeming cash reward in CAPRI.**

3

4 **Design**

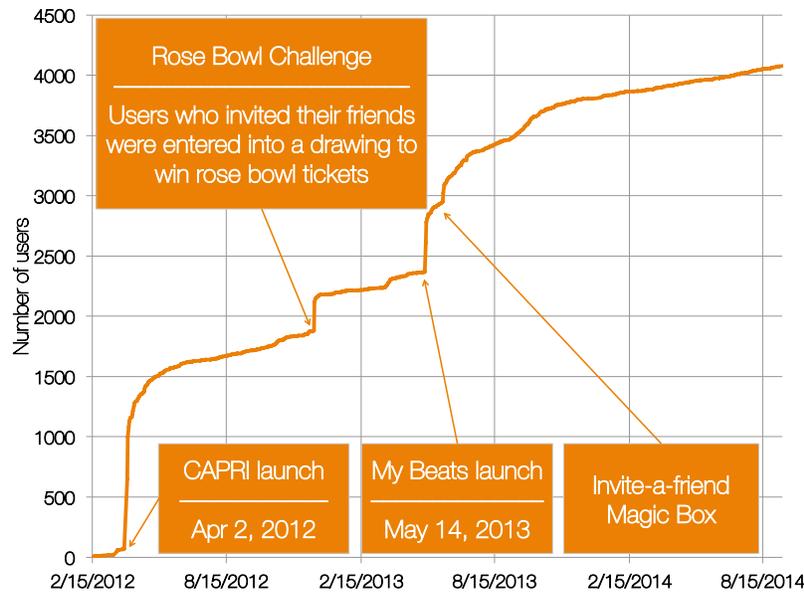
5 Stanford University parking permit holders who parked inside the “congestion cordon” were
 6 invited to participate in the program on Apr 2nd, 2012. Those who enrolled were given passive
 7 RFID tags to place on their windshield. The tags were sensed at ten main entrances/exits of the
 8 Stanford campus. The sensors detected participants’ vehicles for 3 hours in the morning
 9 (7–10AM) and in the afternoon (4–7PM) on each weekday. The peak-hours were defined to be
 10 8–9AM and 5–6PM. For each automobile detected by the sensors during the off-peak shoulder
 11 hours (7–8AM and 9–10AM, and 4–5PM and 6–7PM), the participant was awarded 10 points.
 12 Additionally, CAPRI assigned each participant a *boost day*, a day on which their off-peak trip
 13 earned them 30 points. From May 14th, 2013, CAPRI incentivized walkers and bikers by
 14 awarding them between 10 and 25 points depending on the length of the commute. Walking and
 15 biking activity was monitored using a smart-phone app, called *My Beats*. The CAPRI program
 16 concluded on Sept 30th, 2014.

17

18 Participants can redeem their points for cash in one of two ways: (i) deterministically, by
 19 trading 100 points for \$1 (or a full week’s worth of off-peak trip points, ignoring boost points),
 20 and (ii) randomly, by playing a chutes-and-ladders game using their points on the CAPRI
 21 website (FIGURE 1). Playing the game is tantamount to a “self-administered lottery”. The
 22 game gave cash rewards ranging from \$1–\$50. 87.3% of the participants used the random
 23 rewards option. Since participants were allowed to change the manner of redeeming rewards
 24 (deterministic or random), it is interesting to note that 13.2% of the participants chose to switch
 25 from the deterministic option to the random option at some point during the program. The
 26 rewards were paid out as paycheck supplements or through bank deposits.

27

28 CAPRI used a number of other interventions to increase the popularity (and hence the
 29 enrollment through friend recommendations), engagement and behavior shift among the
 participants. We review the important ones below.



1
2 **FIGURE 2 Number of CAPRI registrants over time**

3
4 *Status system.* Participants were placed in one of four status levels: Bronze, Silver, Gold,
5 and Platinum. Participants started at the Bronze level. Depending on the number of off-peak
6 shoulder hour trips they made on a weekly basis, they could upgrade their status from their
7 current level to the next. At the Silver, Gold and Platinum levels, failure to make the number of
8 off-peak shoulder hour trips required for that status level results in a degrading of the status by
9 one level.

10 Status is only worth something if it affords a privilege; CAPRI used status as follows.
11 Participants with higher status had higher odds of winning rewards in the game. Further,
12 higher-valued rewards were only available at the higher status levels. For example, the \$50
13 reward was only available at the Platinum level. These two features—higher status means higher
14 odds of winning and higher-valued rewards—have been also been used in the INSTANT (11),
15 Steptacular (12) and Insinc (13) projects.

16
17 *Friends.* Participants were allowed to invite their friends (such as those who were eligible
18 to participate in CAPRI) to join the program as well as to connect with their friends on the
19 CAPRI portal. A participant could see their friends' recent updates; e.g., a status upgrade, a cash
20 award won, etc. This feature provided a basis for social influence to spread.

21
22 *"Magic Box".* From the data it gathered, CAPRI learned a participant's preference for
23 commuting off-peak (e.g., only twice a week, only on Monday and Tuesday mornings and
24 Wednesday evenings). It also learned a participant's propensity for shifting to off-peak based on
25 their response to past offers. Based on these learnings, CAPRI offered personalized incentives
26 through a tab in a commuter's portal called *"Magic Box"*. Each week some of the participants
27 received such offers. Magic Box offers were previously used in Insinc (13).

28
29 *"Trendjacking".* Stanford is involved in numerous high profile sporting events like the
30 Rose Bowl, the "Big Game" (the annual football match between the Stanford Cardinals and the

1 Berkeley Golden Bears), etc. CAPRI offered tickets to some of these events and used them to
2 incentivize behavior shift or increase enrollment.

3
4 *The My Beats smartphone app and parking incentives.* CAPRI detected walking and
5 biking activity using the My Beats app (for both Android and iOS). The app was also developed
6 so as to detect where a person parks so that we could reward those parked in designated and
7 underutilized parking lots. However, due to poor cellular signals at some of the parking
8 structures, My Beats could not accurately and reliably detect parking behaviors. Therefore, we
9 could not implement the parking incentives portion of CAPRI.

10
11

12 **4. ANALYSIS OF CAPRI**

13 The CAPRI program started on April 2, 2012 and ended on September 30, 2014. Over the two
14 and a half years, 4,057 Stanford affiliates completed the registration process; this includes 3,082
15 car commuters and 975 biking/walking commuters. These car commuters cover about 30.0% of
16 the 10,290 car commuters in Stanford who were ever eligible to participate during this period. A
17 graph of the number of registrations over time is shown in FIGURE 2. The figure indicates that
18 there were significant increases in the number of registrations during promotional programs like
19 the Rose Bowl Challenge and from personalized offers like the invite-a-friend magic box. These
20 users conducted a total of 734,562 RFID scans and 98,751 My Beats trips. CAPRI has given out
21 a total of \$211,989 in incentives.

22

23 **Demographics**

24 A closer look at the demographics of CAPRI participants reveals that 63% are female; 72% are
25 staff members, 14% are students and the rest are faculty; and 68% are younger than 45 years of
26 age. On an average day, 795 unique participants make 1,196 automobile trips. Note that a
27 unique participant can contribute both a morning and an evening trip. Moreover, the scanners do
28 not operate other than during the 6 hours specified; thus, a participant's morning trip may not be
29 detected even though their evening trip is detected. The My Beats app records 158 unique
30 participants making a total of 225 walking or biking trips on an average weekday. 1,095
31 participants have at least one CAPRI friend and 46 users have six or more friends. The top
32 friend-maker in CAPRI has 31 friends.

33

34 **Automobile Commutes**

35 In this section we investigate the behavior of automobile commuters in CAPRI. Stanford
36 University employs an independent consultant company to count the number of vehicles entering
37 and exiting campus each day for six weeks in the spring and two weeks in the fall each year.
38 This generates the "cordon count data" and gives an aggregate profile of all the traffic entering
39 and leaving the campus during the morning and evening peak and shoulder hours.

40 FIGURE 3 compares the profile of the overall Stanford traffic monitored in 2013 and
41 traffic due to CAPRI users since the start of program. For the three-hour monitoring window of
42 CAPRI in the morning and afternoon, i.e. 7–10AM and 4–7PM, we average over days and
43 normalize the traffic time density so that the area under each curve in FIGURE 3 is exactly one.
44 From the aggregate profile collected by the university, we find that during an average day a total

1 of 11,620 inbound trips and 13,315 outbound trips occur in the corresponding monitoring
2 windows (this includes pass-through traffic, visitors, construction vehicles, etc.). CAPRI
3 commuters make 619 inbound trips and 577 outbound trips per day during the same hours. The
4 percentage of the number of commutes before, during and after the peak hour is also shown in
5 the figure. The fraction of trips in the gray-shaded area, indicating the peak hour, is the
6 *peak-hour trip ratio*.

7 From FIGURE 3, we see that:

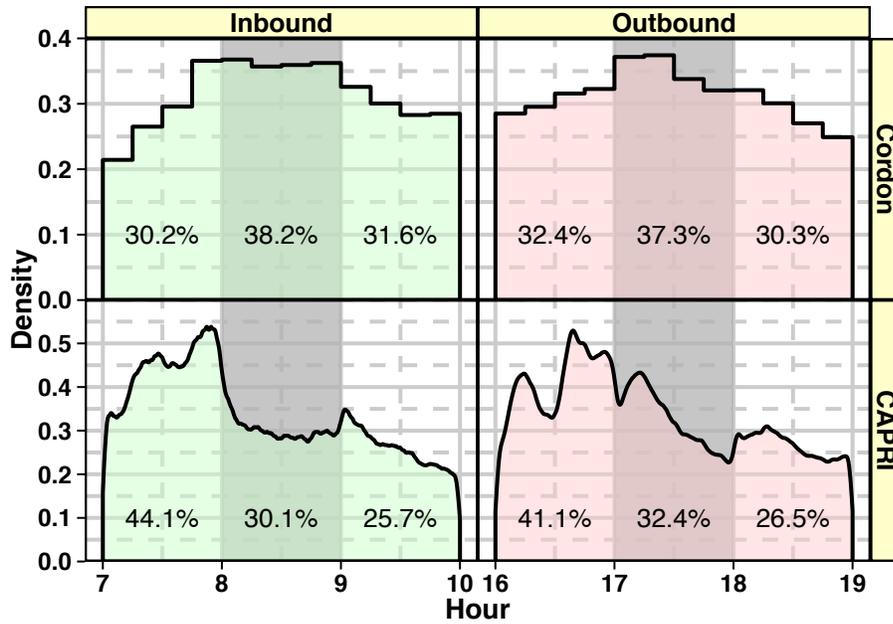
- 8 • **The overall Stanford traffic density is heaviest during the peak-hours.** The morning
9 peak-hour trip ratio is 38.2% and the evening peak-hour trip ratio is 37.3%.
- 10 • **CAPRI users avoid peak hours.** For CAPRI participants, the peak-hour trip ratio is only
11 30.1% in the morning and 32.4% in the evening, which is a 21.2% and a 13.1% reduction
12 from the Stanford-wide traffic.
- 13 • **CAPRI users respond to incentives.** The commute density for CAPRI participants
14 peaks adjacent to (but just outside) the peak hours. Furthermore, CAPRI users prefer
15 commuting during the hour before the peak hour as compared to the hour after the peak
16 hour.

17 As only aggregate statistics about Stanford traffic are available from the cordon count, we
18 cannot unambiguously determine whether there is a significant shift in the commute hours of
19 CAPRI users after they joined program. It is possible that CAPRI users may have been
20 commuting during off-peak hours even before the program. To support the hypothesis of a
21 behavioral shift under incentives, we conducted a survey in May 2013, in which users were
22 asked if they had shifted their commute time after joining CAPRI. In the next section, we
23 compare the performance between self-declared shifters and non-shifters and gain some insight
24 into their behavioral change.

26 *Shift statistics segmented by user-declared behavior change*

27 In May 2013, a survey of CAPRI participants was conducted. The survey aimed to obtain a
28 better understanding of CAPRI participants' behavioral change, especially given the lack of
29 commute data for these users before they joined CAPRI. The responses from the survey are used
30 to determine if there has been a shift in the participant's commute times since joining CAPRI,
31 the amount of shifting, the reason for shifting, etc. In total, exactly 1,000 CAPRI users
32 responded, among which 602 users claimed that they had shifted commute time by some amount
33 since joining CAPRI, while 398 users declared that they did not shift. Based on this information,
34 we split the users into two groups: self-declared shifters and self-declared non-shifters.

35 As shown in FIGURE 4, self-declared shifters did indeed shift: the peaks of the time
36 density of these commuters occur at 7:58 and 9:02 for morning inbound commutes, and at 16:58
37 and 18:01 for afternoon outbound commutes. These are times just adjacent to, but outside, the
38 peak hours. The peaks of the commute time density for self-declared non-shifters occur well
39 before peak hours: around 7:30 for morning inbound commutes, and 16:15 and 16:40 for
40 afternoon outbound commutes, which are *drastically* different from the self-declared shifters.
41 From these observations, we infer:

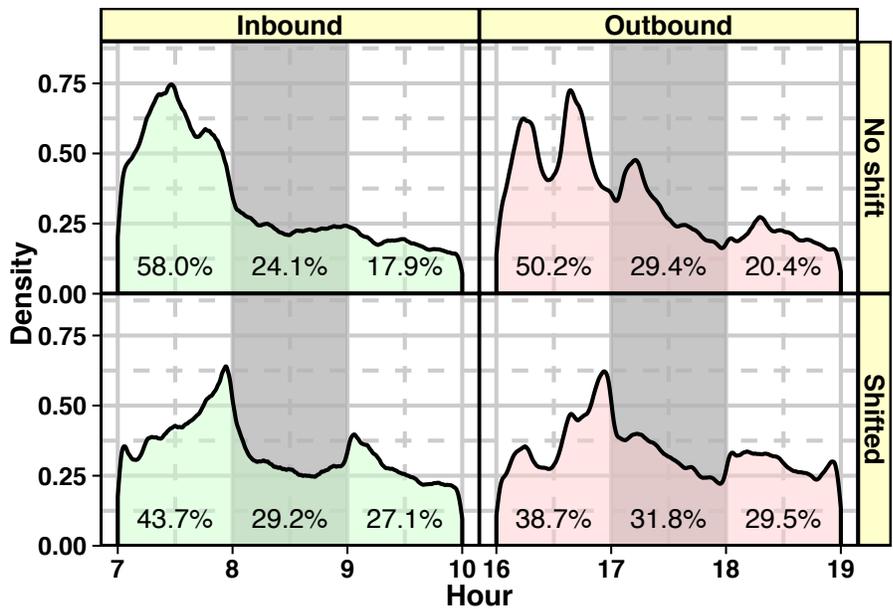


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2 **FIGURE 3** Stanford commute time density (cordon counts) vs. CAPRI participants'
 3 commute time density on an average day. The time window is 7–10AM and 4–7PM. Areas
 4 shaded gray indicate the designated peak hours: 8–9AM and 5–6PM. Each density is
 5 normalized: the area under each curve is 1.

6

7



8

9 **FIGURE 4** Commute time density for self-declared shifters and non-shifters in CAPRI
 10 survey. Areas shaded gray indicate the designated peak hours: 8–9AM and 5–6PM.

- 1 • The self-declared shifters shifted the *minimum amount necessary* to receive rewards
- 2 from CAPRI; these shifts can be reasonably attributed to CAPRI.
- 3 • The self-declared non-shifters were likely already commuting *far away from peak*
- 4 *hours*. Although CAPRI did not cause this commute behavior, it provides incentives
- 5 for these users to *maintain* their behavior.
- 6

7 *Linear regression on features affecting automobile commute time*

8 To further examine the behavioral shift of CAPRI users, we conduct a linear regression analysis
 9 to quantify the effects of different factors that affect the commute times of CAPRI participants.
 10 The commute time of each participant on every morning/afternoon is regressed on:

- 11 • Whether the user has won any rewards from CAPRI in the past week;
- 12 • Whether the user has any friends winning rewards from CAPRI in the past week;
- 13 • Whether the day of the commute is a boost day for this user;
- 14 • The day of the week.

15 These features capture the impact from incentives, friends' winnings and daily schedules.
 16 As the peak hours in CAPRI are in the middle of the 3-hour windows (7AM–10AM and 4PM–
 17 7PM), commuters may shift their commute times *earlier* or *later*. To consider shifts in different
 18 directions, we categorize users into two groups: those whose median commute time is before the
 19 center of the peak hour (early users) and those whose median commute time is after the center of
 20 the peak hour (late users). Thus, for morning commutes, early users consist of those whose
 21 median commute time is prior to 8:30, while late users consist of those whose median commute
 22 time is after 8:30; for afternoon commutes, early users are those whose median commute time is
 23 prior to 17:30, and late users are those whose median commute time is after 17:30.

24 TABLE 1 shows the linear regression coefficients for different features affecting CAPRI
 25 participants' commute time. As shown in the table, for early users, winning rewards in the past
 26 week can have an immediate impact on their commute time: they shift 4 to 5 minutes earlier in
 27 both morning and afternoon commutes (hence away from peak hours). The winning of a
 28 participant's friend also has an impact, albeit a smaller one: early commuters who have friends
 29 winning rewards in the past week travel around 1.5 minutes earlier. Early commuters also
 30 advance their commutes by an additional minute on their boost days to ensure receiving bonus
 31 award points.

32 For late users, those who won rewards in the past week shift about 3 minutes later in
 33 morning and afternoon commutes (hence away from peak hours). Rewards won by friends and
 34 boost days do not significantly influence the afternoon commutes of late users.

35 On Fridays, both early and late users tend to leave earlier. In particular, late users leave
 36 10.8 minutes earlier on Fridays.

37 In summary, the rewards in CAPRI have a direct effect on participants' commute time:
 38 they shift their commute time away from peak hours when receiving rewards in the recent past.
 39 The influence of friends and the boost day is also observed.

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 41
 42

1 **TABLE 1 Linear Regression Coefficients for Different Features on CAPRI Participants'**
 2 **Commute Time. Statistically Significant Coefficients Are Shown (5% Significance Level).**

Morning Commutes	Early Users		Late Users	
	Time	p-value	Time	p-value
Mean commute time	7:50	$< 2 \times 10^{-16}$	9:01	$< 2 \times 10^{-16}$
The user won rewards in the past week	-5.3min	$< 2 \times 10^{-16}$	+3.2min	$< 2 \times 10^{-16}$
The user has friends who won rewards in the past week	-1.5min	3.51×10^{-10}	-	-
It is the user's boost day	-1.0min	3.41×10^{-7}	+1.6min	1.48×10^{-9}
It is Friday	+0.9min	7.02×10^{-4}	-2.3min	2.05×10^{-11}
Afternoon Commutes	Early Users		Late Users	
	Time	p-value	Time	p-value
Mean commute time	16:58	$< 2 \times 10^{-16}$	17:57	$< 2 \times 10^{-16}$
The user won rewards in the past week	-4.0min	$< 2 \times 10^{-16}$	+2.8min	1.18×10^{-13}
The user has friends who won rewards in the past week	-1.7min	5.00×10^{-9}	-	-
It is the user's boost day	-1.0min	4.10×10^{-5}	-	-
It is Friday	-3.0min	$< 2 \times 10^{-16}$	-10.8min	$< 2 \times 10^{-16}$

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5 **My Beats App Commutes**

6 In My Beats, variable numbers of points are given for walking and biking commutes based on
 7 distance. Most of these trips are between 10 and 30 minutes in duration.

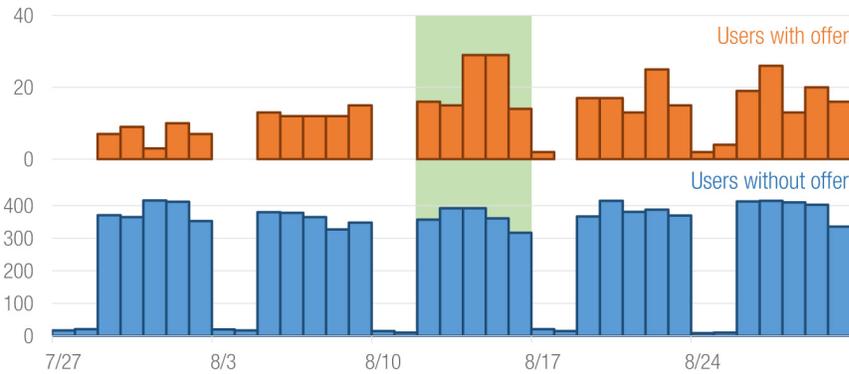
8 We observe that there are some My Beats app users who also have automobile commutes
 9 logged in CAPRI. Indeed, some users almost completely converted to walking and biking
 10 commutes after the launch of My Beats. To find these users, we define a *convert* as a participant
 11 who had at least 10 automobile commutes logged before the My Beats launch, has since logged
 12 more than 10 My Beats trips, and has more My Beats trips than automobile commutes since the
 13 launch. In total, there are 28 such converts. For example, there is a participant who made 192
 14 automobile commutes before the launch but made *exclusively* bike commutes after the launch of
 15 the app. This shows that CAPRI is effective in persuading some participants to shift modes from
 16 driving to walking and biking.

17

18 **“Magic Box” and Trendjacking**

19 In total, CAPRI deployed more than 10 rounds of magic boxes offers in the fall of 2013. These
 20 included instant status upgrades, bonus points for friend invites, bonus points for eligible trips,
 21 and points for completing a survey on commute behavior.

22 Magic boxes are effective in shaping participants' commute behavior. For instance, in
 23 August 2013, CAPRI issued a one-week magic box that gives bonus points for eligible My Beats
 24 trips. 202 users accepted the offer. FIGURE 5 shows the number of My Beats trips made by
 25 CAPRI participants. As shown, those who accepted the offer increased the number of My Beats
 26 trips they made by 61% compared to the week before; whereas, other participants had an
 27 increase of only 1% in trip count over the previous week. Additionally, a memory-effect can be
 28 observed in those who accepted the offer: the two weeks following the end of the magic box saw
 29 increases of 36% and 47% in the number of My Beats trips over the week prior to the magic box.



1

2 **FIGURE 5** The number of My Beats trips made by CAPRI participants, grouped by
 3 whether the user received and accepted the magic box offer of bonus points for eligible
 4 trips. Shaded week indicates the bonus period.

5

6 During the 2013 Rose Bowl Challenge (an annual collegiate football championship for
 7 which Stanford qualified), CAPRI offered five pairs of tickets to some lucky participants. In
 8 order to qualify for the draw, a participant had to invite their friends to join CAPRI and each
 9 accepted invitation earned the participant one entry into the raffle. Over the 2 weeks period of
 10 the challenge, over 500 friend invitation emails were sent and 287 invited users successfully
 11 completed registration, resulting in an increase of over 13% in CAPRI's total user population.
 12

12

13 5. SUMMARY

14 This paper analyzed CAPRI, an incentive program that aimed to shift drivers out of the peak
 15 hours and towards alternate modes such as walking and biking. We conducted an analysis of
 16 CAPRI based on 734,562 automobile trips and 98,751 walking and biking trips collected from
 17 4,057 CAPRI participants during two and a half years. We found that the rewards in CAPRI
 18 were effective in incentivizing users to drive during off-peak hours, and to walk and bike. In
 19 particular, rewards received in the recent past shifted users' commute times away from the peak
 20 hours, and there were some users who converted from driving to walking or biking to work.
 21 Furthermore, personalized offers—the “magic box”—was useful in shaping users' commute
 22 behavior and increasing their participation.

23 As mentioned in the paper, the My Beats app could not detect parking behaviors
 24 accurately and reliably due to poor cellular coverage at some of the designated parking lots. In
 25 future, we intend to study parking incentives.
 26

26

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