An Integrated System Model for Evaluating the Impact of the Dynamic ICC Toll Policy on the Regional Network Mobility



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# AN INTEGRATED SYSTEM MODEL FOR EVALUATING THE IMPACT OF THE DYNAMIC ICC TOLL POLICY ON THE REGIONAL NETWORK MOBILITY

FINAL REPORT

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By

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#### **1. INTRODUCTION**

Road pricing has been advocated as an efficient travel demand management to alleviate congestion since the seminal work by Pigou (1920) and Knight (1924) (see Lindsey, 2006, for recent reviews). More specifically, dynamic toll pricing has received greater interest among policy makers and public agencies due to its potential for lowering energy costs for society. Some analytical studies (e.g., Arnott et al., 1990) have found that dynamic toll pricing generally yield greater efficiency gains than static toll pricing because the former reduce queueing delays by altering travelers' departure times as well as routes.

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This study proposes the model that enables practitioners to integrate user response to dynamic toll pricing. The analysis accounts for cost and time savings perceived by regional drivers and the users' response to time-varying road charges. More specifically, the study captures difference in behavioral characteristics of the willingness to pay (WTP) for toll across users socioeconomics and trip related characteristics such as time of day, and trip purpose. The analysis is expected to be useful for transportation agency responsible for the ICC operations.

### 2. PREVIOUS WORK ON VALUE OF TIME AND WILLINGNESS TO PAY

High occupancy toll (HOT) lanes are a special case of road pricing in that they refer to high occupancy vehicle (HOV) facilities that are open to lower-occupancy vehicles upon payment of a fee or toll (Goodman, Jurasin, Larwin, Orski, Turnbull, & Cunard, 1998). They can be characterized as a commodity that offers faster, safer and more reliable travel time for the drivers that are willing to pay to access it.

Currently, HOT lanes exist in several U.S. cities, and more are under construction or planned, such as Atlanta (GA), Austin (TX), Oakland (CA), and Washington (DC). As of 2010, existing HOT lanes include the following (Los Angeles County Metropolitan Transportation Authority, 2010):

- I-95 Express Toll Lanes in Miami, Florida (since 2008)
- SR-167 HOT Lanes Pilot in Seattle, Washington (since 2008)
- I-25 Express Lanes in Denver, Colorado (since 2006)
- I-15 Express Lanes Pilot in Salt Lake City, Utah (since 2006)
- I-394 in Minneapolis, Minnesota (since 2005, employs dynamic tolling)
- I-15 Express Lanes in San Diego, California (since 1998, employs dynamic tolling)
- HOT Lanes on the I-10 Katy Freeway and US-290 Northwest Freeway, Houston, Texas (since 1998)
- SR-91 Express Lanes in Orange County, California (since 1995)

The previous list supports to the notion that congestion pricing, and HOT lanes, is becoming a more attractive alternative for agencies to deal with: 1) high congestion levels and 2) underused HOV lanes. Nevertheless, the key question always is how much drivers are willing to pay to save time by avoiding congested roadways.

Lam and Small (2001) measured value of time (VOT) and value of reliability (VOR) using data on actual travel behavior in a real pricing context. In this study, they collected data from the toll highway State Route 91 (SR-91) drivers. SR-91 users face a choice between two parallel routes, one free but congested and the other with time-varying tolls. At the time of this study, high occupancy vehicles (3 plus occupants) paid half of the toll. The authors created several models with different variables' interaction and weight. Their proposed model yielded a VOT of \$22.87 per hour, whereas one of the other reliable models estimated VOT as \$16.37 per hour. Both VOT are estimates for congested travel.

Brownstone, et al. (2003) estimated willingness to pay to reduce travel time on a congested road. The approach uses RP data from the I-15 congestion pricing project in San Diego, collected over a two month period, in October and November 1998. They found that the average willingness to pay to reduce travel time of morning commuters is roughly \$30 per hour. In addition, they explain that I-15 drivers mostly preoccupy about unexpected delays in their morning commute. As a result, they are willing to pay a higher than usual toll to avoid unusual congestion. Furthermore, the authors highlight the fact that the toll lanes along the I-15 are separated from the rest of the freeway, which may improve driving conditions and safety. As a result, this may affect the willingness to pay estimates by biasing the estimates upward.

Later on, Brownstone and Small (2005) compared their previous results. Their results from the SR91 and I-15 corridors based on RP lane-choice data almost always obtain median VOT of \$20 to \$40 per hour. They found that the VOT estimated using stated preference data (SP) are about half of the ones estimated using revealed preference (RP) data, SR-91 and I-15 respectively. They explain this by stating that: 1) people display time inconsistency in their actual behavior, but not in their hypothetical behavior (i.e. drivers make the higher-cost choice more often in real life than on hypothetical surveys); and 2) the difference is caused by systematic misperception of travel times (i.e., RP respondents overestimate their time saving). The authors then implemented sample weights in order to match the income and commute distance distribution of both facilities. This approach yielded similar VOT around \$20 per hour for I-15 and SR-91.

More recently, Finkleman, Casello and Fu (2011) conducted a willingness to pay (WTP) study in the Greater Toronto Area, Ontario, Canada. Given that the study was performed in Canada, all monetary value related to it are in Canadian dollar. The study estimates WTP based on common trip and travelers' characteristics (i.e. travel time, income, trip distance, etc.) and introduces "trip urgency" as a factor. For this, they collected SP data of 255 Toronto residents. Their results indicate that under a trip condition of high urgency and high congestion level on parallel lanes, travelers are willing to pay \$4.12 for 24 minute trip on the HOT lane, which equates to \$10.3 per hour. On the other hand, for the least urgent trips, their WTP decreases to \$0.58 for a 13 minute trip, which equates \$2.7 per hour. As expected, they found that trip urgency plays a significant role on travelers' decision to pay for admission to the HOT lanes.

Literature on value of time for dynamic toll systems is scarce given the lack of facilities to perform this type of studies. In summary, the existing literature universally concludes that traveler's value of time and willingness to pay varies by the trip and driver characteristics. In addition, value of time is sensitive to geographic (i.e. regional) effects, as evidenced by the different values obtained in each study.

#### **3. SURVEY DESIGN**

The stated preference survey was conducted in the Maryland Capital Beltway (I-495) to capture response of the potential regional driver on the ICC in the presence of dynamic toll pricing. The questionnaire was designed as a web-based survey where the survey recruitment was conducted by flyer distributions at several exit locations of I-495. The sample population consisted of car drivers traveling on I-495 during weekday extended peak periods (8:00 AM - 11:00 AM and 3:30 PM. - 6:00 PM) on March 21-25, and May 23-27, 2011. A total of 200 respondents from a sample of 4,000 who received the flyers responded to the questionnaire which results in the overall response rate of 5%. Within the 200 responded surveys, 173 of the respondents completed the survey, which results in the effective sample size of 173 observations for the model estimation.

The questionnaire consists of two parts: revealed preference (RP) and stated preference (SP) questions. The description for each part of the survey is described as follows:

#### 3.1 Revealed Preference (RP) Questionnaire

The revealed preference (RP) questionnaire consists of two sections: respondents' socioeconomics and recent trip information.

### **3.1.1 Socioeconomic Information**

The purpose of this section is to investigate socioeconomic data of the potential HOT lane users in I-495. The respondent is asked to describe his/her socioeconomic information via the following constructs:

- Gender
- Age
- Household income range
- Education
- Occupation
- Number of worker per household
- Number of vehicle in the household
- Vehicle type most used by the respondent
- Number of years the vehicle owned
- ZIP code of work place

# **3.1.2 Recent Trip Information**

The recent trip information gathers data about the respondent's most recent trip on I-495. The purpose of this section is to use respondent's experienced trip condition as the pivot point when designing the stated preference (SP) question. This ensures that the stated scenario in the SP part is realistic for each respondent. The respondent is asked to describe his/her most recent trip information on I-495 via the following constructs:

- Mode choice
- Number of passenger
- Trip purpose
- Departure time (DT)
- Arrival time (AT)
- Preferred departure time (PDT)
- Preferred arrival time (PAT)
- Total travel time in minutes (TT)
- Total trip distance in miles (D)

- Fuel cost (FC)
- Parking cost (\$)
- Toll cost (\$)
- Entry and exit ramp locations
- Shortest (TT min) and longest (TT max) travel time experienced on the whole trip in minutes
- Shortest (ST) and longest (LT) travel time experienced on the beltway in minutes
- Number of departure time alternatives respondents have considered
- Corresponded departure and arrival time for the alternative departure time
- Work starting/ending time, work schedule flexibility (whether can start work 30 minutes later)

# 3.2 Stated Preference (SP) Questionnaire

The stated preference (SP) portion of the survey aims at investigating traveler lane choice in response to time-of-day traffic condition and dynamic toll pricing scheme. It presents respondents with 7 scenarios of stated experiments on the lane choice alternatives. Given that a prevalent form of congestion pricing in the US is high-occupancy/toll (HOT) lanes, the dynamic toll pricing is introduced through the HOT lane alternative.

The game consists of three alternatives and four variables. These variables have the maximum of five levels of variation per alternative. Three alternatives presented to respondents are: (1) Solo driver on normal lane, (2) High Occupancy Toll (HOT) lane, and (3) High Occupancy Vehicle (HOV) lane. The variables included in the stated preference are: (1) Departure time, (2) Travel time range, (3) Fuel cost, and (4) Toll. These variables are designed to account for traffic conditions by time of day taking into account observed respondents' departure time where the peak period is defined as 8:00 AM. to10:00 AM., and 3:00 PM. to 7:00 PM. (Crunkleton, 2008). The description of the variables used in the survey is as follows:

1) Departure time: Departure time is pivoted from respondent's reported departure time in the RP part.

- 2) Total travel time range: This variable is designed to account for both time-of-day conditions based on the respondent's reported departure time and travel condition on toll lane. It is aimed at capturing travel time uncertainty.
- Fuel cost: The fuel cost is designed to reflect higher expenses in the peak period and on the normal lane. The fuel cost is pivoted from the reported fuel cost in the RP part.
- 4) Toll cost: The toll cost is designed as a mileage based using the Intercounty Connector toll rates (MDTA, 2010). The toll rate for the HOT lane accounts varies depending on whether the respondents' reported departure time is in the peak or non-peak period.

The survey is designed with orthogonal design approach where numerical evaluations in a wide range of parameters values was undertaken to guarantee sufficient efficiency of the design. The pilot study, in combination with expert judgments, was also used to arrive at the final levels of attribute in the SP experiment. The questionnaire design of the stated preference survey is shown in Table 1.

| Variable          | Normal Lane             | HOT Lane                | HOV Lane (passengers >1) |
|-------------------|-------------------------|-------------------------|--------------------------|
| Departure time    | DT-40min                | DT-40min                | DT-40min                 |
|                   | DT-20min                | DT-20min                | DT-20min                 |
|                   | DT                      | DT                      | DT                       |
|                   | DT+20min                | DT+20min                | DT+20min                 |
|                   | DT+40min                | DT+40min                | DT+40min                 |
| Total travel time | If DT in peak hour      | If DT in peak hour      | If DT in peak hour       |
| range (minute)    | TTmin+20 to TTmin+30;   | TTmin+10 to TTmin+20;   | TTmin+10 to TTmin+20;    |
|                   | TTmin+20 to TTmin+40;   | TTmin+10 to TTmin+25;   | TTmin+10 to TTmin+25;    |
|                   | TTmin+20 to TTmax;      | TTmin+10 to TTmin+30;   | TTmin+10 to TTmin+30;    |
|                   | TTmin+20 to TTmin+45;   | TTmin+10 to TTmin+25;   | TTmin+10 to TTmin+25;    |
|                   | TTmin+20 to TTmin+35;   | TTmin+10 to TTmin+20;   | TTmin+10 to TTmin+20;    |
|                   | If DT not in peak hour, | If DT not in peak hour, | If DT not in peak hour,  |
|                   | TTmin+10 to TTmin+20;   | TTmin+5 to TTmin+10;    | TTmin+5 to TTmin+10;     |
|                   | TTmin+10 to TTmin+30;   | TTmin+5 to TTmin+15;    | TTmin+5 to TTmin+15;     |
|                   | TTmin+10 to TTmax-10;   | TTmin+5 to TTmin+20;    | TTmin+5 to TTmin+20;     |
|                   | TTmin+10 to TTmax-20;   | TTmin+5 to TTmin+15;    | TTmin+5 to TTmin+15;     |
|                   | TTmin+10 to TTmax-30;   | TTmin+5 to TTmin+10;    | TTmin+5 to TTmin+10;     |
| Fuel cost (\$)    | If DT in peak hour      | If DT in peak hour      | If DT in peak hour       |
|                   | FC(1+10%)               | FC                      | FC                       |
|                   | FC(1+20%)               | FC(1+10%)               | FC(1+10%)                |
|                   | FC(1+30%)               | FC(1+20%)               | FC(1+20%)                |
|                   | If DT not in peak hour, | If DT not in peak hour, | If DT not in peak hour,  |
|                   | FC(1+10%)               | FC                      | FC                       |
|                   | FC(1+15%)               | FC(1+15%)               | FC(1+15%)                |
|                   | FC(1+20%)               | FC(1+20%)               | FC(1+20%)                |
| Toll cost (\$)    | 0                       | If DT in peak hour      | 0                        |
|                   |                         | \$0.3/mile*D            |                          |
|                   |                         | \$0.35/mile*D           |                          |
|                   |                         | \$0.4/mile*D            |                          |
|                   |                         | \$0.45/mile*D           |                          |
|                   |                         | \$0.5/mile*D            |                          |
|                   |                         | If DT not in peak hour, |                          |
|                   |                         | \$0.1/mile*D            |                          |
|                   |                         | \$0.15/mile*D           |                          |
|                   |                         | \$0.2/mile*D            |                          |
|                   |                         | \$0.25/mile*D           |                          |
|                   |                         | \$0.3/mile*D            |                          |

Table 1 Stated Preference Survey Design

Figure 1 shows the interface of the stated preference survey on the website.

| following travel option   | is are available for your trip alo  | ng the Capitol Beltway.                 |                          |
|---|---|---|--------------------------|
| trip is from Exit 36 to   | Exit 25.  | - 10 - 11 - 11 - 11 - 11 - 11 - 11 - 11 |                          |
|   | Normal Travel Lane  | SOV Lane (No<br>Passengers)             | HOV Lane<br>(Passengers) |
| Departure Time  | 8:40 AM   | 7:40 AM                                 | 7:40 AM                  |
| Travel Time   | 45 - 75 mins  | 30 - 40 mins                            | 30 - 40 mins             |
| Fuel Cost   | \$3.90  | \$3.30                                  | \$3.30                   |
| Toll Cost   | \$0.00  | \$3.84                                  | \$0.00                   |
| th travel option would<br>I Will Use the Normal<br>I Will Use the SOV La<br>I Will Use the HOV La<br>I Will Not Use the Bel | you prefer for your trip?<br>Travel Lanes.<br>ne (Single-Occupant Vehicle)<br>ne (High-Occupancy Vehicle)<br>tway (I will use an alternate ro | oute)                                   |                          |

Figure 1 Stated Preference Questionnaire

# 4. DESCRIPTIVE STATISTICS

A sample size of 173 completed surveys was collected. The respondent's characteristics are divided into two groups, socioeconomics and trip characteristics. The distribution of the sample is presented next.

# **4.1 Socioeconomics Results**

Socioeconomic data was collected for the respondent's gender, age, education, and occupation; and the household's income, number of worker in the household, number of vehicle in the household, and others. The summary of socioeconomic statistics is shown in Table 2.

Gender Statistics show that 51% of respondents are male.

<u>Age</u> Respondents' ages are distributed with an average age of 43 and median age of 45. The youngest respondent is 19 and the oldest is 82.

*Education* Approximately 49% of the respondents are at a graduate or professional level. In addition, 34% of the respondents have a bachelor's degree and 8% have some college education. The remaining respondents are distributed almost uniformly across the other levels.

<u>Occupation</u> Approximately 46% of the respondents work for a private company, whereas 30% work for the government. Only 1.2% of the respondents are unemployed.

*Income* Approximately 32% of households have incomes above \$150,000. Approximately 24% of households have incomes between \$100,000 and \$149,999. Similarly, 24% of households have income between \$50,000 and \$99,999. Around 10% of households have incomes less than \$50,000 with the remaining households (9%) refusing to answer the question.

<u>Number of workers in the household</u> Only 2.3% of the sample have no worker in the household. Nearly 27% have 1 worker and 54% of households have 2 workers. Almost 8% of households have 3 workers, whereas 3.5 have 4 or more workers.

<u>Number of vehicle per household</u> Nearly 2% of households have no vehicles. On the other hand, 27% reported having 1 vehicle and 54% had 2 vehicles. 7.5% of households have 3 vehicles and 3.5% have 4 or more vehicles with the remaining households (6.4%) refusing to answer the question.

| Table 2 | Socioeconomic | Statistics |
|---------|---------------|------------|
|---------|---------------|------------|

| Category              | Case                            | Respondents | Percentage* |
|-----------------------|---------------------------------|-------------|-------------|
| Gender                | Male                            | 77          | 45.00%      |
|                       | Female                          | 89          | 51.00%      |
| Age                   | 18-25                           | 13          | 7.51%       |
|                       | 26-35                           | 40          | 23.12%      |
|                       | 36-45                           | 33          | 19.08%      |
|                       | 46-60                           | 63          | 36.42%      |
|                       | Greater than 60                 | 15          | 8.67%       |
| Education             | Associate Degree                | 3           | 1.70%       |
|                       | Bachelor Degree                 | 59          | 34.10%      |
|                       | Graduate or Professional Degree | 84          | 48.60%      |
|                       | High School Graduate            | 2           | 1.20%       |
|                       | Less than High School           | 2           | 1.20%       |
|                       | Some College                    | 14          | 8.10%       |
| Occupation            | Faculty or School Staff         | 5           | 2.90%       |
|                       | Retired                         | 2           | 1.20%       |
|                       | Self Employed                   | 10          | 5.80%       |
|                       | Student                         | 8           | 4.60%       |
|                       | Work for Private Company        | 80          | 46.20%      |
|                       | Work for the Government         | 52          | 30.10%      |
|                       | Unemployed                      | 2           | 1.20%       |
|                       | Other                           | 6           | 3.50%       |
| Household Income      | Less than \$50,000              | 18          | 10.40%      |
|                       | \$50,000-\$99,999               | 41          | 23.70%      |
|                       | \$100,000-\$149,999             | 42          | 24.28%      |
|                       | Greater than \$150,000          | 56          | 32.37%      |
| Workers per household | 0                               | 4           | 2.31%       |
|                       | 1                               | 46          | 26.59%      |
|                       | 2                               | 93          | 53.76%      |
|                       | 3                               | 13          | 7.51%       |
|                       | 4                               | 2           | 1.16%       |
|                       | Greater than 4                  | 4           | 2.31%       |
| Vehicle per household | 0                               | 4           | 2.31%       |
|                       | 1                               | 46          | 26.59%      |
|                       | 2                               | 93          | 53.76%      |
|                       | 3                               | 13          | 7.51%       |
|                       | 4                               | 2           | 1.16%       |
|                       | Greater than 4                  | 4           | 2.31%       |

\* Travelers who skipped the answer are excluded from the statistics.

#### **4.2 Trip Characteristics Results**

Respondents were also asked about their trips, mainly focusing on the mode of transportation, number of passengers, different times and costs associated with the trip, distance traveled, and others. The summary of trip characteristic statistics is shown in Table 3.

*Carpool.* Approximately 17% of the respondents are carpool travelers.

Shared Expenses. Of the carpool travelers, 80% of them do not share fuel and parking expenses.

<u>Number of Passengers.</u> Of the carpool travelers, 60% and 23% of them have 1 passenger and 2 passengers respectively. The remaining have 3 or more passengers.

<u>Travel Time.</u> The average travel time of the respondents is 30 minutes, without considering the trips that took more than 2 hours. Around 29% of the respondents spent 15 minutes or less in their trip, whereas 41% spent between 16 and 30 minutes. Almost 15% spent between 31 and 45 minutes and nearly 7% took between 46 and 60 minutes. The remaining spent more than 60 minute to complete their trip.

<u>Departure Time</u>. The majority of the trips departed within peak hours (33% in AM peak and 32% in PM peak).

*Distance Traveled.* The average trip length is 23.4 miles. 41% of the respondents traveled between 20 and 40 miles, followed by 10 to 20 miles with 32%. Around 15% traveled less than 10 miles with the remaining traveling 40 miles or more.

*Fuel Cost.* On average, each respondent spent \$6 on fuel. Most of the respondent (60%) spent between \$2.5 and \$10, whereas nearly 24% spent between 0 and \$2.5. The remaining 15.6% is somewhat equivalently distributed among the other cost groups.

<u>Parking Cost.</u> The average parking cost is \$0.96, without taking into consideration two respondents that combined paid \$200 for parking. 85% of the respondents did not pay for parking, while 11% paid less than \$10.

*Toll Cost.* The average toll cost is \$0.24. 94% of the respondents did not pay for toll, while 3% paid less than \$5. 1.7% paid between \$5 and \$10, with the remaining paying more than \$10 in toll.

<u>Shortest Trip Time.</u> The majority of the respondents (56%) performed their shortest trip between 16 and 30 minutes. On the other hand, 14% took 15 minutes or less, while 21% took between 31 and 45 minutes. Only 6.4 % took between 46 and 60 minutes and the remaining needed more than 60 minutes to complete their trip.

*Longest Trip Time.* The average longest trip time is 93.5 minutes. Only 3% of respondents performed their longest trip in 15 minutes or less, while 4.6% needed between 16 and 30 minutes. The remaining respondents are somewhat equivalently distributed among the other time ranges.

<u>Shortest Travel Time on Beltway.</u> On average, respondents performed their shortest trip on the Beltway in 15 minutes. Nearly 46% of respondents traveled for 10 minutes or less, whereas approximately 39% needed between 11 and 20 minutes to complete their trip. 11% took between 21 and 30 minutes, and the remaining 4% needed more than 30 minutes.

<u>Longest Travel Time on Beltway.</u> On average, respondents performed their longest trip on the Beltway in 62 minutes. Nearly 9% of respondents traveled for 15 minutes or less, whereas approximately 7% needed more than 120 minutes to complete their trip. The remaining respondents are somewhat equivalently distributed among the other time ranges.

| Category                          | Case            | Respondents | Percentage* |
|-----------------------------------|-----------------|-------------|-------------|
| Carpool                           | Carpool         | 30          | 17.30%      |
|                                   | Non-Carpool     | 143         | 82.70%      |
| Travel Time (min)                 | 0-15            | 50          | 28.9%       |
|                                   | 16-30           | 71          | 41.0%       |
|                                   | 31-45           | 25          | 14.5%       |
|                                   | 46-60           | 11          | 6.4%        |
|                                   | Greater than 60 | 16          | 9.2%        |
| Departure Time                    | Before 6 AM     | 4           | 2.4%        |
|                                   | 6AM-8AM         | 30          | 17.8%       |
|                                   | 8AM-10AM        | 55          | 32.5%       |
|                                   | 10AM-12PM       | 9           | 5.3%        |
|                                   | 12PM-2PM        | 8           | 4.7%        |
|                                   | 2PM-4PM         | 10          | 5.9%        |
|                                   | 4PM-7PM         | 54          | 32.0%       |
|                                   | After 7PM       | 3           | 1.8%        |
| Distance Traveled (mile)          | 0-5             | 5           | 2.9%        |
|                                   | 5-10            | 20          | 11.6%       |
|                                   | 10-20           | 56          | 32.4%       |
|                                   | 20-40           | 71          | 41.0%       |
|                                   | 40-60           | 7           | 4.0%        |
|                                   | Greater than 60 | 13          | 7.5%        |
| Fuel Cost                         | 0-2.5           | 41          | 23.70%      |
|                                   | 2.5-5           | 54          | 31.20%      |
|                                   | 5-10            | 51          | 29.50%      |
|                                   | 10-20           | 12          | 6.90%       |
|                                   | 20-40           | 6           | 3.50%       |
|                                   | Greater than 40 | 9           | 5.20%       |
| Parking Cost                      | 0               | 147         | 85.00%      |
|                                   | 0-5             | 10          | 5.80%       |
|                                   | 5-10            | 9           | 5.20%       |
|                                   | 10-20           | 5           | 2.90%       |
|                                   | Greater than 20 | 2           | 1.20%       |
| Toll Cost                         | 0               | 162         | 93.60%      |
|                                   | 0-5             | 5           | 2.90%       |
|                                   | 5-10            | 3           | 1.70%       |
|                                   | Greater than 10 | 3           | 1.70%       |
| Minimum Freeway Travel Time (min) | 0-10            | 79          | 45.70%      |
|                                   | 11-20           | 68          | 39.30%      |
|                                   | 21-30           | 19          | 11.00%      |

Table 3 Trip Characteristics Statistics

|                                   | 31-45            | 3  | 1.70%  |
|-----------------------------------|------------------|----|--------|
|                                   | Greater than 45  | 4  | 2.30%  |
| Maximum Freeway Travel Time (min) | 0-15             | 15 | 8.70%  |
|                                   | 16-30            | 33 | 19.10% |
|                                   | 31-45            | 40 | 23.10% |
|                                   | 46-60            | 27 | 15.60% |
|                                   | 61-90            | 23 | 13.30% |
|                                   | 91-120           | 23 | 13.30% |
|                                   | Greater than 120 | 12 | 6.90%  |

\* Travelers who skipped the answer are excluded from the statistics.

# **5. MODELS AND RESULT**

In this section, the information on traveler's socioeconomic and trip characteristics presented in section 4 are used for the model estimation, the variable definitions are as follows:

- The variable Time is the traveler's travel time in minutes.
- Trip time is divided into two groups:
  - PeakHr: dummy variable that takes the value of 1 if the trip takes place between peak hours (8-10 AM or 3-7 PM) and 0 otherwise.
  - OffPeakHr: dummy variable that takes the value of 1 if the trips takes place during off-peak hours and 0 otherwise.
- The variable Tc refers to the toll cost in US dollars.
- Carpool is a dummy variable that takes a value of 1 if the person carpools and 0 otherwise.
- Distance refers to the distance traveled in miles.
- Trip purpose in divided into four dummy variables which take the value of 1 if trip purpose is the same as the variable and 0 otherwise:
  - Hbw\_sch: work or school related trip.
  - Hbsoc: social trip.
  - Hbshop: shopping trip.
  - Hbo: other trips.
- Traveler's age is divided into five dummy variables which take the value of 1 if age falls between the specified range and 0 otherwise:
  - o AgeLess25
  - o Age25to35
  - o Age36to45
  - o Age46to60
  - o AgeMore60
- The education level is captured by 4 dummy variables which take the value of 1 if the education level is the same as the variable and 0 otherwise:
  - o Somecollege

- Ass\_degr: represents associate degree.
- Bach\_degr: represents bachelor degree.
- Gradprof\_degr: represents a graduate or professional degree.
- Income refers to the household income level. These variables are represented as categorical values which range from 0 to 3 depending on whether the income is less than \$50k, \$50k-\$100k, \$100k-\$150k, and more than \$150k, respectively.
- Cars\_per\_hh represents the number of vehicles in the household.
- Traveler's occupation is divided into seven dummy variables which take the value of 1 if the occupation is the same as the variable and 0 otherwise:
  - Occ\_other: refers to unrevealed or miscellaneous occupation.
  - Occ\_facsch: refers to school faculty.
  - Occ\_priv: refers to working for a private company.
  - Occ\_self: refers to self employed.
  - Occ\_ret: refers to retired.
  - Occ\_stud: refers to student.
  - Occ\_unemp: refers to unemployed.
  - Occ\_gov: refers to working for the government.

The following significance levels are used in this study:

- \* Significant at the 10% level;
- \*\* Significant at the 5% level;
- \*\*\* Significant at the 1% level.

### 5.1 Model 1: Generic Willingness to Pay (WTP)

To estimate the willingness to pay (WTP) toll to reduce travel time, a model is developed focusing on time and toll cost, see Table 4. The resulting willingness to pay is 12.93 \$/hr, with a 95% confidence interval of [11.63 \$/hr, 16.53 \$/hr]. Notice that the variables time and toll cost are significant at the 1% level and the signs are in line with intuition.

| Scenario        | Coef.       | P> z      | ]          |        |            |        |
|-----------------|-------------|-----------|------------|--------|------------|--------|
| Time            | -0.0402***  | 0.0000    |            |        |            |        |
| Tc_             | -0.1866***  | 0.0000    |            |        |            |        |
| Chaina          | Normal Lane |           | НОТ        |        | HOV        |        |
| Choice          | Coef.       | P> z      | Coef.      | P >  z | Coef.      | P >  z |
| _cons           | Base Al     | ternative | -0.8328*** | 0.000  | -2.2371*** | 0.000  |
| Log Likelihood: | -746.8564   | # obs:    | 2682       |        |            |        |
| Wald Chi2(42):  | 52.56       | # cases:  | 894        |        |            |        |
| Prob > Chi2:    | 0.0000      | # Alt:    | 3          |        |            |        |

Table 4 Model 1 Result

### 5.2 Model 2: Willingness to Pay (WTP) by Income Level.

A second model is estimated to calculate the willingness to pay by income level, see Table 5. One would expect that the WTP monotonically increases by level of income. Nevertheless, the results indicate that it behaves in a u-shape manner, see Table 5. This could be explained by the fact that low income people's (less than \$50,000) daily activities have an intense dependency on their strict work and personal schedules (e.g. they cannot afford to be late for work and risk losing it). On the other hand, although high income people (more than \$150,000) may not have a fixed schedule, their daily activities may have a higher monetary value. It should be noticed that the estimated WTP for income level 2 is not significant at the 5% confidence level; hence it should not be considered as accurate.

| Level | HH Income       | WTP (\$/hour) | [95% Conf. Inte | erval] (\$/hour) |
|-------|-----------------|---------------|-----------------|------------------|
| 0)    | Less than 50K   | 16.03         | 19.61           | 7.22             |
| 1)    | 50 K – 100 K    | 11.03         | 12.17           | 8.21             |
| 2)    | 100 K – 150 K   | 5.59          | 8.31            | (1.11)           |
| 3)    | More than 150 K | 17.67         | 16.70           | 20.05            |

Table 5 Willingness to Pay by Income Level

Model 2 result is shown in Table 6. Toll cost and the interaction between time and income levels have expected signs and are significant at the 1% level, except for income level 2, which is significant at the 10% level. Moreover, all variables within the model are significant at the 5% level or less and have expected sign. As expected, travelers that carpool show preference to HOV lanes. The results indicate that travelers with longer trips would prefer HOT lanes, although it should be noticed that the difference between alternative's coefficients is relatively small. Travelers with trip purpose home based office are the most likely to use HOT lanes. In addition, as age increases HOV preference monotonically decreases. On the other hand, as education level increases HOV preference increases as well. Self employed travelers prefer to use HOT lanes. This could be explained by assuming that self employed travelers have very unique and personal schedules that are difficult to match, compared to other occupations. In contrast, private employees prefer HOV lanes, which could be explain by the previous reasoning. As an overall, based on the constant, HOT is the least preferred alternative.

| Scenario        | Coef.       | P >  z   | ]        |     |       |          |     |        |  |
|-----------------|-------------|----------|----------|-----|-------|----------|-----|--------|--|
| TimeInc0        | -0.0600 *** | 0.008    |          |     |       |          |     |        |  |
| TimeInc1        | -0.0413 *** | 0.001    |          |     |       |          |     |        |  |
| TimeInc2        | -0.0209 *   | 0.079    |          |     |       |          |     |        |  |
| TimeInc3        | -0.0662 *** | 0.000    |          |     |       |          |     |        |  |
| Tc_             | -0.2247 *** | 0.000    |          |     |       |          |     |        |  |
| Choice          | Normal I    | Lane     | E        | НОТ |       |          | HOV |        |  |
|                 | Coef.       | P> z     | Coef.    |     | P> z  | Coef.    |     | P >  z |  |
| Carpool         |             |          | -1.3632  | *** | 0.002 | 4.6624   | *** | 0.000  |  |
| Distance        |             |          | 0.0181   | *** | 0.007 | 0.0134   | *   | 0.056  |  |
| Hbw_sch         |             |          | 0.8956   |     | 0.104 | -1.6864  | **  | 0.010  |  |
| Hbshop          |             |          | 1.6402   | **  | 0.018 | -15.8597 |     | 0.995  |  |
| Hbo             |             |          | 2.0191   | *** | 0.001 | -1.0094  |     | 0.177  |  |
| AgeLess25       |             |          | 1.2346   | **  | 0.043 | 1.8416   | *   | 0.051  |  |
| Age25to35       |             |          | 0.8777   | **  | 0.044 | -1.0536  |     | 0.218  |  |
| Age36to45       |             |          | -0.1377  |     | 0.763 | -2.6457  | **  | 0.012  |  |
| Age46to60       |             |          | 0.8177   | **  | 0.048 | -4.2467  | *** | 0.000  |  |
| Ass_degr        | Rasa altar  | notivo   | 0.5587   |     | 0.717 | -16.9022 |     | 0.993  |  |
| Bach_degr       | Dase allen  | lative   | 0.0694   |     | 0.893 | -3.5681  | *** | 0.001  |  |
| Gradprof_degr   |             |          | 0.7716   |     | 0.134 | -2.8665  | *** | 0.004  |  |
| Occ_other       |             |          | -0.7082  |     | 0.580 | 5.7006   | *** | 0.000  |  |
| Occ_facsch      |             |          | 0.1477   |     | 0.789 | -15.4865 |     | 0.997  |  |
| Occ_priv        |             |          | 0.2417   |     | 0.252 | 2.7924   | *** | 0.000  |  |
| Occ_self        |             |          | 0.5596   |     | 0.228 | -2.7975  | *   | 0.075  |  |
| Occ_ret         |             |          | 4.0022   | *** | 0.001 | -15.8858 |     | 0.994  |  |
| Occ_stud        |             |          | -0.3708  |     | 0.552 | 0.2612   |     | 0.845  |  |
| Occ_unemp       |             |          | -16.6141 |     | 0.998 | 1.4388   |     | 0.395  |  |
| _cons           |             |          | -3.3891  | *** | 0.000 | -0.3547  |     | 0.806  |  |
| Log Likelihood: | -482.913    | # obs:   | 2535     |     |       |          |     |        |  |
| Wald Chi2(43):  | 201.19      | # cases: | 845      |     |       |          |     |        |  |
| Prob > Chi2:    | 0.0000      | # Alt:   | 3        |     |       |          |     |        |  |

Table 6 Model 2 Result

# **5.3 Model 3: Willingness to Pay (WTP) by Trip Purpose**

The third model is estimated to calculate willingness to pay by trip purpose. The results are significant at the 5% level, except for willingness to pay for shopping trips, which is not significant. Hence, its value should not be taken into account. The resulting WTP's are presented

in Table 7. Work and school related trips presented the lower WTP, whereas 'Other' trips presented the highest WTP. This result, although counterintuitive, is not uncommon. Several studies, such as Cirillo and Axhausen (2006) and Finkleman, Casello, and Fu (2011), have found that high urgency trips and shopping and leisure trips have higher WTP compared to work and school related trips.

| Purpose        | Variable | WTP (\$/hour) | [95% Conf. Interval] (\$/ho |         |
|----------------|----------|---------------|-----------------------------|---------|
| Shopping       | HBShop   | 10.35         | 18.64                       | (12.05) |
| Work or School | HBW_Sch  | 11.17         | 10.94                       | 11.81   |
| Other          | HBO      | 22.95         | 24.20                       | 19.56   |
| Social         | HBSoc    | 14.00         | 18.28                       | 2.46    |

Table 7 Willingness to Pay by Trip Purpose

Model 3 result is shown in Table 8. Toll cost and the interaction between trip purposes and time have expected signs and are significant at the 1% level, except for social related trips, which is significant at the 5% level. Two new variables are included in this model, cars per household and income, both significant at the 5% level. The higher the number of vehicles in a household the lower their likelihood to select HOT lanes. This could be explained by the unwillingness of households to pay toll for each individual vehicle. In addition, travelers with high income prefer HOV lanes. Finally, all remaining variables follow the same pattern as the previous model.

| Scenario        | Coef.       | P> z     |          |     |        |          |     |        |
|-----------------|-------------|----------|----------|-----|--------|----------|-----|--------|
| Hbshop_time     | -0.0362     | 0.229    |          |     |        |          |     |        |
| Hbw_sch_time    | -0.0391 *** | 0.000    |          |     |        |          |     |        |
| Hbo_time        | -0.0803 *** | 0.000    |          |     |        |          |     |        |
| Hbsoc_time      | -0.0490 **  | 0.030    |          |     |        |          |     |        |
| Tc_             | -0.2101 *** | 0.000    |          |     |        |          |     |        |
| Choice          | Normal      | Lane     | HOT E    |     |        | IOV      |     |        |
|                 | Coef.       | P> z     | Coef.    |     | P >  z | Coef.    |     | P >  z |
| Carpool         |             |          | -1.3776  | **  | 0.000  | 5.4282   | *** | 0.000  |
| Distance        |             |          | 0.0146   | **  | 0.029  | 0.0083   |     | 0.227  |
| AgeLess25       |             |          | 1.7300   | *** | 0.006  | 3.0939   | *** | 0.005  |
| Age25to35       |             |          | 0.8626   | *   | 0.055  | -0.9282  |     | 0.310  |
| Age36to45       |             |          | -0.3380  |     | 0.469  | -3.1690  | *** | 0.003  |
| Age46to60       |             |          | 0.7293   | *   | 0.089  | -4.9329  | *** | 0.000  |
| Ass_degr        |             |          | 0.9419   |     | 0.535  | -20.5178 |     | 0.996  |
| Bach_degr       |             |          | 0.5144   |     | 0.324  | -4.9062  | *** | 0.000  |
| Gradprof_degr   |             |          | 1.0487   | **  | 0.042  | -4.4373  | *** | 0.000  |
| Occ_other       | Base alte   | rnative  | 0.2340   |     | 0.853  | 7.6714   | *** | 0.000  |
| Occ_facsch      |             |          | 0.6741   |     | 0.215  | -11.7820 |     | 0.991  |
| Occ_priv        |             |          | 0.4453   | **  | 0.035  | 3.2580   | *** | 0.000  |
| Occ_self        |             |          | 0.2687   |     | 0.531  | -3.4877  | **  | 0.036  |
| Occ_ret         |             |          | 5.6679   | *** | 0.000  | -12.6054 |     | 0.985  |
| Occ_stud        |             |          | -0.0431  |     | 0.946  | 1.0704   |     | 0.430  |
| Occ_unemp       |             |          | -13.4859 |     | 0.995  | 2.0227   |     | 0.227  |
| Cars_per_hh     |             |          | -0.3110  | **  | 0.010  | 0.2589   |     | 0.320  |
| Income          |             |          | 0.5068   | *** | 0.000  | 0.8817   | *** | 0.004  |
| _cons           |             |          | -3.0976  | *** | 0.000  | -2.8879  | **  | 0.044  |
| Log Likelihood: | -483.647    | # obs:   | 2535     |     |        |          |     |        |
| Wald Chi2(41):  | 199.22      | # cases: | 845      |     |        |          |     |        |
| Prob > Chi2:    | 0.0000      | # Alt:   | 3        |     |        |          |     |        |

Table 8 Model 3 Result

### 5.4 Model 4: Willingness to Pay (WTP) by Congestion Period

The fourth model is estimated to calculate willingness to pay by congestion period (i.e. peak hour and off-peak hour). The results indicate that WTP for peak hour is less than for off-peak hours, see Table 9. As indicated in the previous models, non work and school related trips have higher willingness to pay. It is reasonable to assume that these trips take place during off-peak hours, hence the higher WTP for off-peak hour trips.

| Period        | WTP (\$/hour) | [95% Conf. Interval] (\$/hour) |         |  |  |  |
|---------------|---------------|--------------------------------|---------|--|--|--|
| Peak hour     | \$ 11.70      | \$ 11.33                       | \$ 0.21 |  |  |  |
| Off-peak hour | \$ 13.62      | \$ 5.01                        | \$ 0.21 |  |  |  |

Table 9 Willingness to Pay by Congestion Period

Model 4 result is shown in Table 10. Toll cost and the interaction between congestion period and time have expected signs and are significant at the 1% level. All remaining variables follow the same pattern as the previous models.

| Scenario        | Coef.      | P> z     |           |          |           |           |  |
|-----------------|------------|----------|-----------|----------|-----------|-----------|--|
| TimePeakHr      | -0.0441 ** | * 0.000  |           |          |           |           |  |
| TimeOffPea~r    | -0.0513 ** | * 0.000  |           |          |           |           |  |
| Tc_             | -0.2260 ** | * 0.000  |           |          |           |           |  |
| Choice          | Normal     | Lane     | НОТ       |          | HOV       |           |  |
|                 | Coef.      | P> z     | Coef.     | P> z     | Coef.     | P> z      |  |
| Carpool         |            |          | -1.4486 * | ** 0.001 | 4.5737 *  | *** 0.000 |  |
| Distance        |            |          | 0.0150 *  | * 0.028  | 0.0100 *  | ° 0.158   |  |
| Hbw_sch         |            |          | 0.7155    | 0.173    | -1.4686 * | ** 0.029  |  |
| Hbshop          |            |          | 1.4166 *  | * 0.035  | -13.9082  | 0.991     |  |
| Hbo             |            |          | 1.8882 *  | ** 0.001 | -0.6460   | 0.406     |  |
| AgeLess25       |            |          | 2.2025 *  | ** 0.001 | 2.8351 *  | ** 0.010  |  |
| Age25to35       |            |          | 0.9784 *  | * 0.028  | -0.8992   | 0.316     |  |
| Age36to45       |            |          | -0.3358   | 0.469    | -2.7133 * | *** 0.009 |  |
| Age46to60       |            |          | 0.6102    | 0.149    | -4.6046 * | *** 0.000 |  |
| Ass_degr        |            |          | 0.8032    | 0.601    | -15.9549  | 0.985     |  |
| Bach_degr       | Base alter | mative   | 0.1935    | 0.714    | -4.0866 * | *** 0.000 |  |
| Gradprof_degr   |            |          | 0.7962    | 0.127    | -3.4276 * | *** 0.001 |  |
| Income          |            |          | 0.5403 *  | ** 0.000 | 0.5777 *  | ° 0.066   |  |
| Occ_other       |            |          | 0.1653    | 0.899    | 6.7651 *  | *** 0.000 |  |
| Occ_facsch      |            |          | 0.6020    | 0.305    | -13.5461  | 0.995     |  |
| Occ_priv        |            |          | 0.4289 *  | * 0.045  | 2.9825 *  | *** 0.000 |  |
| Occ_self        |            |          | 0.5501    | 0.235    | -2.9623 * | · 0.069   |  |
| Occ_ret         |            |          | 4.5479 *  | ** 0.000 | -14.1206  | 0.989     |  |
| Occ_stud        |            |          | 0.1353    | 0.832    | 0.2559    | 0.855     |  |
| Occ_unemp       |            |          | -14.4932  | 0.995    | 1.8987    | 0.269     |  |
| _cons           |            |          | -4.3586 * | ** 0.000 | -1.1913   | 0.449     |  |
| Log Likelihood: | -476.432   | # obs:   | 2535      |          |           |           |  |
| Wald Chi2(43):  | 209.39     | # cases: | 845       |          |           |           |  |
| Prob > Chi2:    | 0.0000     | # Alt:   | 3         |          |           |           |  |

Table 10 Model 4 Result

#### 6. CONCLUSIONS AND FUTURE RESEARCH DIRECTIONS

In this study, we propose the user response model to dynamic toll pricing. The modeling framework provides the willingness to pay toll to reduce travel time which varies across user socioeconomics and trip related characteristics such as time of day, and trip purpose. The data used for the analysis are based on the dedicated stated preference survey on the lane choice where attributes are travel time, toll cost, and fuel cost, all of which account for traffic condition by time of day.

Four models are proposed to calculate the willingness to pay toll to reduce travel time. The model considers the user choice of lane among 1) Normal lane, 2) HOT lane, and 3) HOV lane in the presence of dynamic toll pricing. Given that departure time is incorporated in the choice attributes, the model can determine not only change in lane choice in response to dynamic toll pricing, but also temporal shift toward less congested period. The model estimation results are in line with the expectation based on the previous studies. Apart from estimating the generic willingness to pay (WTP) for toll, the approach enables to account for different WTP by income level, trip purpose, and congestion period. Based on the results, the generic willingness to pay toll for travel time saving is 12.93 \$/hour. The willingness to pay by income level ranges from 5.59 \$/hour to 17.67 \$/hour for household with \$100,000-\$149,999 income and greater than \$150,000 income respectively. The willingness to pay by trip purpose ranges from 10.35 \$/hour to 22.95 \$/hour for the home based shopping and home based others respectively. Finally, the willingness to pay by congestion period is 11.70 \$/hour in the peak hour and 13.62 \$/hour in the off peak hour.

Future research extensions are suggested based on this modeling capability. For instance, the model can be integrated into a solution framework to find an optimal dynamic pricing scheme so as to alleviate congestion and social benefit. In addition, the data collected from this survey can incorporate stochastic travel condition based on travel time uncertainty. More importantly, the model calibrated from the data collected at the ICC facility when the system is fully operational will provide more representative behavioral characteristics of the ICC users.

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