

Feedback Based Dynamic Congestion Pricing

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Abstract

5 This paper presents a mathematical model for dynamic congestion pricing at a toll where alternate lane or
6 routes are available. The model developed is based on traffic conservation law and queueing, and moreover
7 uses fundamental macroscopic relationships for its derivation. The modeling uses a Logit model for the
8 price and driver choice behavior relationship. We use this nominal mathematical model for the dynamics
9 to derive a feedback control law that uses real-time information to come up with tolling price. Simulation
10 results show the performance of this dynamic feedback congestion pricing algorithm.

11 INTRODUCTION

12 Congestion pricing is defined as charging motorists during peak hours to encourage them to either switch
 13 their travel times or to use alternative routes which are not congested at peak hours. The theory behind road
 14 pricing suggests that, in order to reach social optimum, a toll needs to be charged which is equal to the
 15 difference between social marginal costs (which include external costs that users impose on each other on a
 16 congested road) and private average costs of users (travel delays, fuel, maintenance etc.).

17 In recent years, with the help of technological developments such as electronic toll collection sys-
 18 tem, pricing can also be done dynamically, that is, tolls can be set real-time varied according to the traffic
 19 conditions. Although the continuously time-varying optimal tolls suggest a fair system for the users, it is
 20 also debatable whether smoothly-varying toll rate will be appreciated by drivers. Therefore, in real world dy-
 21 namic pricing applications step (piecewise constant) tolls are mainly used. Examples from US are depicted
 22 in Table 1.

TABLE 1 Dynamic Pricing Applications in US

FACILITY	TOLLING SYSTEM	WEBSITE
San Diego I-15 FasTrak	The toll schedule varies dynamically every 6 minutes <i>footnote</i>	http://argo.sandag.org/fastrak
Orange County, CA SR-91	Toll varies every hour depending on traffic conditions	www.91expresslanes.com
Minnesota I-394	Tolls can be varied as frequently as 3 minutes <i>footnote</i>	http://www.mnpass.org/394/index.html

23 BACKGROUND AND MOTIVATION

24 As mentioned in the previous section, dynamic congestion pricing application currently make use of step
 25 functions that dynamically adjust toll rates based on the prevailing traffic conditions on the toll road. Clearly,
 26 this approach has several shortcomings including the lack of theoretical basis for the determination and
 27 implementation of tolls. Moreover, such an approach can cause unpredicted fluctuations in travel times and
 28 overall sub-optimal results in terms of users as well as the system.

29 In this paper, we will propose a theoretically sound feedback based congestion pricing model that
 30 will be attempt to:

- 31 1. Achieve the pre-set objective such as system optimal or allowable user-equilibrium.
- 32 2. Develop a control law that is robust against uncertainties within a set.
- 33 3. Assure that the developed control law is stable and does not fluctuate in an implementable way.
 34 For example, if the dynamic toll prices changes from \$5 to \$20 and then to \$5 in a very short
 35 period of time, say 5 minutes, the, this will create unexpected results and low compliance rates.
 36 Thus, dynamic toll prices should increase and decrease in a relatively smooth way.
- 37 4. Incorporate bounds for maximum and minimum tolls to ensure equity.

38 LITERATURE REVIEW

39 Congestion pricing has been one of the most important research topics in traffic engineering throughout the
 40 last few decades. Several studies were conducted on the theoretical aspects of pricing models ((13), (5),
 41 (3), Byung-Wook Wie and Tobin, 1998, (2)). Lindsey (2003) reviewed road pricing applications in the US

¹http://ops.fhwa.dot.gov/tolling_pricing/value_pricing/projects/involving_tolls/priced_lanes/hot_lanes/ca_hotlanes_i15sd.htm

²<http://www.tollroadsnews.com/node/1197>

42 and Canada by comparing the implementations in Europe and applicability of the different categories of
 43 congestion pricing to US roads (11). Other international congestion pricing experiences such as Singapore,
 44 Norway and United Kingdom were explained and lessons learned from these implementations were also
 45 analyzed by different authors ((10), (7), (12)). In practice, congestion pricing is performed generally by
 46 1) HOT (High-occupancy toll lanes) lanes, which are the lanes reserved for vehicles that meet minimum
 47 occupant requirements or vehicles that pay tolls, 2) Cordon pricing, which is charging vehicles to access a
 48 zone (e.g. highly congested part of a metropolitan city). Some of the major road pricing applications in the
 49 US are summarized in Table 2

TABLE 2 Major Road Pricing Initiatives in the US

(a) Dynamic Pricing Applications

Project/Location	Initiation Date	For More Information
I-15 San Diego, CA	1998	http://www.sandag.org/index.asp?projectid = 34&useaction = projects.detail .
I-394 Minneapolis, MN	2005	http://www.mnpass.org/phase2.html
SR 167 Puget Sound Region, WA	2008	http://www.wsdot.wa.gov/Projects/SR167/HOTLanes/
SR 91 Orange County, CA	1995	http://ceenvve.calpoly.edu/sullivan/sr91/sr91.htm
I-95 Miami, FL	2008	http://ops.fhwa.dot.gov/tolling_pricing/value_pricing/projects/involving_tolls/priced_lanes/hot_lanes/fl_hotlanes_i95miami.htm

(b) Time-of-Day Pricing Applications

Project/Location	Initiation Date	For More Information
New Jersey Turnpike Variable Tolls, NJ	2000	http://knowledge.fhwa.dot.gov/cops/hcx.nsf/384aeffe48229e85256a71004b24e0/ba2414ce1eac182685256dc500674090?OpenDocument .
Variable Pricing of Bridges, Lee Co., FL	1998	http://leewayinfo.com/
Variable Tolls on N.Y. Hudson River Crossings	2001	http://www.panynj.gov/tbt/tbtframe.HTM
I-25/US 36 Denver	2006	http://www.its.dot.gov/jpodocs/repts_te/13668_files/chapter_7.htm

50 Although congestion pricing is generally studied for simple settings such as static networks and
 51 homogeneous users, there are also several studies concerning real world conditions. Holguin-Veras and
 52 Cetin (2009) studied optimal tolls for multi-class traffic and generated analytical formulations (6), De Palma
 53 et al. (2005) analyzed time varying tolls considering departure time, route choice, mode split in a dynamic
 54 network equilibrium (1), Chen and Berstein (2004) conducted a study tolling for different user types (9).
 55 Ozbay and Yanmaz-Tuzel (2008) conducted a study on the valuation of travel time and departure time choice
 56 under congestion pricing, considering the New Jersey Turnpike's value pricing implementation (8). There
 57 have also been some recent attempts for developing real-time dynamic congestion pricing algorithms. Zhang
 58 et al. (2008) created a feedback-based tolling algorithm for high-occupancy toll lane facilities. In their
 59 algorithm, the feedback control is obtained by a step-wise function monitoring the speeds of general purpose
 60 lanes and HOT lanes and toll rates are estimated by backward calculation using logit model. Simulation
 61 results of the model showed that overall traffic conditions were improved significantly (4).

SYSTEM MODEL

63 Most of the real-world dynamic toll pricing projects in the US is based on the existence of a toll road and
 64 a toll free road as an alternative. Commuters are this expected to make a decision about choosing the toll
 65 road at a decision point where the prevailing traffic conditions on both roadways as well as the current tolls
 66 are communicated to them mainly through variable message sigs. Thus, each traveler makes a decision as
 67 to whether or not to pay the toll and use the toll road or to simply continue to drive on the free alternative.

68 A modified version of feedback routing model developed by Kachroo and Ozbay (2005, 1998a, b)
 69 with some modifications to its route cost functions, can be used as a mechanism to regulate traffic coming
 70 to the single decision model. However, it is important to first analyze that model in terms of the above

71 requirements specific to the congestion pricing problem.

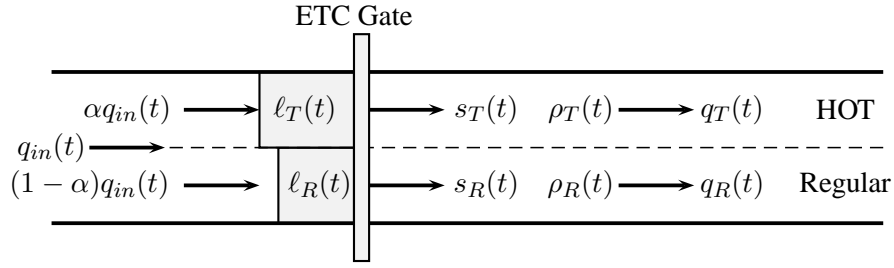


FIGURE 1 System Configuration

72 The system dynamics are given by:

$$\begin{aligned}
 \dot{\ell}_T &= \alpha q_{in}(t) - s_T(t) \\
 \dot{\rho}_T &= s_T(t) - v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m} \right) \\
 \dot{\ell}_R &= (1 - \alpha) q_{in}(t) - s_R(t) \\
 \dot{\rho}_R &= s_R(t) - v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m} \right)
 \end{aligned} \tag{1}$$

73 The symbols for different variables are shown in Table 3.

TABLE 3 Symbols used in the Mathematical Formulation

SYMBOL	MEANING
q_{in}	Traffic in-flow
α	Percent flow using toll
ℓ_T	Queue length for toll lane
ℓ_R	Queue length for regular lane
s_T	Service rate for toll lane
s_R	Service rate for regular lane
ρ_T	Traffic density in toll lane
ρ_R	Traffic density in regular lane
q_T	Traffic outflow from toll lane
q_R	Traffic outflow from regular lane
\bar{L}_T	Length of the toll lane
\bar{L}_R	Length of the regular lane
β	Fraction of toll in-flow using RF-tags
T_T	Travel time through toll lane
T_R	Travel time through regular lane

74 Using Greenshield's fundamental relationship, which we have already used to derive equation 1, we

75 have:

$$\begin{aligned} q_T(t) &= v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m} \right) \\ q_R(t) &= v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m} \right) \end{aligned} \quad (2)$$

76 There are various modifications of the basic model in equation 1 that we can use based on the actual
77 implementation of the tolling scheme. For instance if the tolling is done automatically for everyone using
78 RF-tags, then there will be no queues in the system and there will be no ETC gate. Hence, the dynamics for
79 that implementation will have equations that are shown in equation 3

$$\begin{aligned} \dot{\rho}_T &= \alpha q_{in}(t) - v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m} \right) \\ \dot{\rho}_R &= (1 - \alpha) q_{in}(t) - v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m} \right) \end{aligned} \quad (3)$$

80 If only the toll lane has a gate and no gate for regular lane, then the model has equations that are
81 shown in equation 4

$$\begin{aligned} \dot{\ell}_T &= \alpha q_{in}(t) - s_T(t) \\ \dot{\rho}_T &= s_T(t) - v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m} \right) \\ \dot{\rho}_R &= (1 - \alpha) q_{in}(t) - v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m} \right) \end{aligned} \quad (4)$$

82 If β is the fraction of toll vehicles that use the RF-tages as depicted in Figure 2, then the dynamics
83 will be given by equation 5

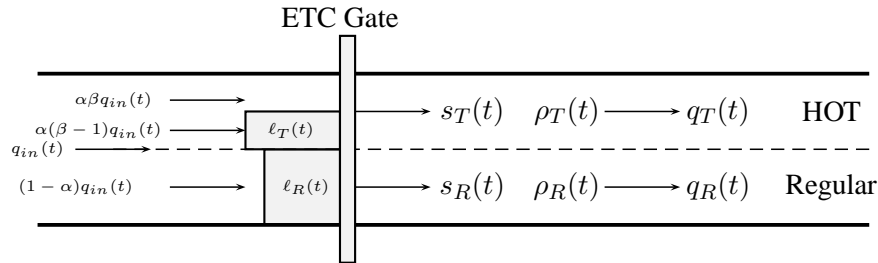


FIGURE 2 System Configuration

$$\begin{aligned} \dot{\ell}_T &= \alpha(1 - \beta)q_{in}(t) - s_T(t) \\ \dot{\rho}_T &= s_T(t) + \alpha\beta q_{in}(t) - v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m} \right) \\ \dot{\ell}_R &= (1 - \alpha)q_{in}(t) - s_R(t) \\ \dot{\rho}_R &= s_R(t) - v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m} \right) \end{aligned} \quad (5)$$

84 The most general model in our setting would allow queueing in every lane and would have the
 85 structure shown in equation 6. In this model, the queue length for tagged vehicles is given by ℓ_{RF} and the
 86 service rate as $s_{RF}(t)$.

$$\begin{aligned}
 \dot{\ell}_T &= \alpha(1 - \beta)q_{in}(t) - s_T(t) \\
 \dot{\ell}_{RF} &= \alpha\beta q_{in}(t) - s_{RF}(t) \\
 \dot{\rho}_T &= s_T(t) + s_{RF}(t) - v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m}\right) \\
 \dot{\ell}_R &= (1 - \alpha)q_{in}(t) - s_R(t) \\
 \dot{\rho}_R &= s_R(t) - v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m}\right)
 \end{aligned} \tag{6}$$

87 In this model, however, if the lanes are wide enough so that no queues are formed for the tagged
 88 and regular vehicles, then we will obtain the model given by equation 7. This is the model we have used in
 89 the simulation studies presented in this paper.

$$\begin{aligned}
 \dot{\ell}_T &= \alpha(1 - \beta)q_{in}(t) - s_T(t) \\
 \dot{\rho}_T &= s_T(t) + \alpha\beta q_{in}(t) - v_f \rho_T \left(1 - \frac{\rho_T(t)}{\rho_m}\right) \\
 \dot{\rho}_R &= (1 - \alpha)q_{in}(t) - v_f \rho_R \left(1 - \frac{\rho_R(t)}{\rho_m}\right)
 \end{aligned} \tag{7}$$

90 FEEDBACK CONTROL LAW FOR THE ROBUST CONGESTION PRICING

91 In order to derive a feedback control law for allowable user-equilibrium, we will use the formula 8 for
 92 travel time through tolled and regular lanes. Drivers would not pay for a facility that would give the same
 93 performance as a free facility. Therefore, the travel times of the two facilities can not be the same. However,
 94 we can obtain an ‘‘allowable’’ user-equilibrium by maintaining some scaled version of travel time in the
 95 tolled lane equal to the regular lane. We use the symbol γ for that factor.

$$\begin{aligned}
 T_T(t) &= \frac{\ell_T(t)}{s_T(t)} + \frac{L_T}{v_f \left(1 - \frac{\rho_T(t)}{\rho_m}\right)} \\
 T_R(t) &= \frac{\ell_R(t)}{s_R(t)} + \frac{L_R}{v_f \left(1 - \frac{\rho_R(t)}{\rho_m}\right)}
 \end{aligned} \tag{8}$$

96 We would like to make the error defined by equation 9 to have closed loop dynamics that is asymp-
 97 totically stable in the sense of Lyapunov.

$$\begin{aligned}
 e(t) &= \gamma T_T(t) - T_R(t) = \\
 &\gamma \left(\frac{\ell_T(t)}{s_T(t)} + \frac{L_T}{v_f \left(1 - \frac{\rho_T(t)}{\rho_m}\right)} \right) - \left(\frac{\ell_R(t)}{s_R(t)} + \frac{L_R}{v_f \left(1 - \frac{\rho_R(t)}{\rho_m}\right)} \right)
 \end{aligned} \tag{9}$$

98 We will use feedback linearization technique to derive the feedback control law. For that design
 99 we need to differentiate the error term with respect to time. Hence, differentiating equation 9 gives us the
 100 dynamics 10

$$\begin{aligned}
\dot{e}(t) &= \gamma \dot{T}_T(t) - \dot{T}_R(t) \\
&= \gamma \left(-\frac{\ell_T(t)}{s_T^2(t)} \dot{s}_T(t) + \frac{\dot{\ell}_T(t)}{s_T(t)} + \frac{L_T}{v_f \rho_m \left(1 - \frac{\rho_T(t)}{\rho_m}\right)^2} \dot{\rho}_T(t) \right) \\
&\quad - \left(-\frac{\ell_R(t)}{s_R^2(t)} \dot{s}_R(t) + \frac{\dot{\ell}_R(t)}{s_R(t)} + \frac{L_R}{v_f \rho_m \left(1 - \frac{\rho_R(t)}{\rho_m}\right)^2} \dot{\rho}_R(t) \right)
\end{aligned} \tag{10}$$

101

Expanding just one term in these error dynamics using system dynamics 5, we get

$$\begin{aligned}
\dot{T}_T(t) &= -\frac{\ell_T(t)}{s_T^2(t)} \dot{s}_T(t) + \frac{\dot{\ell}_T(t)}{s_T(t)} + \frac{L_T}{v_f \rho_m \left(1 - \frac{\rho_T(t)}{\rho_m}\right)^2} \dot{\rho}_T(t) \\
&= -\frac{\ell_T(t)}{s_T^2(t)} \dot{s}_T(t) + \frac{1}{s_T(t)} [\alpha(1 - \beta)q_{in}(t) - s_T(t)] \\
&\quad + \frac{L_T}{v_f \rho_m \left(1 - \frac{\rho_T(t)}{\rho_m}\right)^2} \left[s_T(t) + \alpha\beta q_{in}(t) - v_f \rho_T(t) \left(1 - \frac{\rho_T(t)}{\rho_m}\right) \right]
\end{aligned} \tag{11}$$

102

Similarly,

$$\begin{aligned}
\dot{T}_R(t) &= -\frac{\ell_R(t)}{s_R^2(t)} \dot{s}_R(t) + \frac{\dot{\ell}_R(t)}{s_R(t)} + \frac{L_R}{v_f \rho_m \left(1 - \frac{\rho_R(t)}{\rho_m}\right)^2} \dot{\rho}_R(t) \\
&= -\frac{\ell_R(t)}{s_R^2(t)} \dot{s}_R(t) + \frac{1}{s_R(t)} [(1 - \alpha)q_{in}(t) - s_R(t)] \\
&\quad + \frac{L_R}{v_f \rho_m \left(1 - \frac{\rho_R(t)}{\rho_m}\right)^2} \left[s_R(t) - v_f \rho_R(t) \left(1 - \frac{\rho_R(t)}{\rho_m}\right) \right]
\end{aligned} \tag{12}$$

103

Substituting equations 11 and 14 into equation 10 gives us

$$\dot{e}(t) = f + g\alpha \tag{13}$$

104

where f and g are state dependent terms whose exact formula can be extracted from using 11 and 14 with equation 10.

105

Now, we can design the feedback control law for α as

106

$$\alpha = g^{-1}(-f - ke(t)) \tag{14}$$

107

Using this control law in dynamic equation 14 shows the asymptotic stability of the error as:

$$\lim_{t \rightarrow \infty} e(t) = 0 \tag{15}$$

108

Although we have obtained the closed-loop desired behavior, we still have to come up with the actual toll rate that we must charge. To come up with the functional form for that, we choose a Logit

109

110 model to formulate the driver decision to choose between the tolled and regular lanes. We use the following
111 relationship:

$$\alpha = \frac{1}{1 + \exp(a_1(T_T(t) - T_R(t)) + a_2 p(t) + a_3)} \quad (16)$$

112 Here, $p(t)$ is the toll rate, a_1 is the marginal effect factor of the travel time difference to the driver's
113 utility, a_2 is the marginal effect factor of the toll rate to the same utility, and finally, a_3 covers other factors
114 in the driver choice. From equation 16, we can obtain the deployable toll rate in terms of computed α as

$$p(t) = \frac{1}{a_2} (\ln(\alpha - 1) - a_1(T_T(t) - T_R(t)) - a_3) \quad (17)$$

115 SIMULATION BASED EVALUATION OF ROBUST CONGESTION PRICING

116 We use Scilab software to perform simulations for the dynamics for the system given by equation 7. We use
117 the control law given by equation 14. Now, since we are using no queues for the tagged and regular lanes,
118 there will be no terms for the queue lengths in the controller. Moreover, for the sake of simulation we will
119 assume that the service rate for tolling is fixed. Based on these conditions we get

$$f = \gamma \left(\frac{L_T}{v_f \rho_m \left(1 - \frac{\rho_T(t)}{\rho_m}\right)^2} \left[s_T - v_f \rho_T(t) \left(1 - \frac{\rho_T(t)}{\rho_m}\right) \right] - 1 \right) - \frac{L_R \rho_R(t)}{\rho_m \left(1 - \frac{\rho_R(t)}{\rho_m}\right)} \quad (18)$$

$$g = \gamma \left(\frac{1}{s_T} [(1 - \beta)q_{in}(t)] + \frac{L_T}{v_f \rho_m \left(1 - \frac{\rho_T(t)}{\rho_m}\right)^2} [\beta q_{in}(t)] \right) \quad (19)$$

120 The control law for the simulation is

$$\alpha = g^{-1}(-f - ke(t)) \quad (20)$$

121 where

$$e(t) = \gamma \left(\frac{\ell_T(t)}{s_T} + \frac{L_T}{v_f \left(1 - \frac{\rho_T(t)}{\rho_m}\right)} \right) - \left(\frac{L_R}{v_f \left(1 - \frac{\rho_R(t)}{\rho_m}\right)} \right) \quad (21)$$

122 The simulation parameters are given in Table 4.

123 We use a variable inflow rate to see how the system would evolve. The feedback control law tries to
124 keep the error rate low. In the simulation we also make sure that the implemented value of the split is between
125 zero and one, and also that the queue length and all other state variables always remain non-negative. The
126 simulation results are shown in Figure 3.

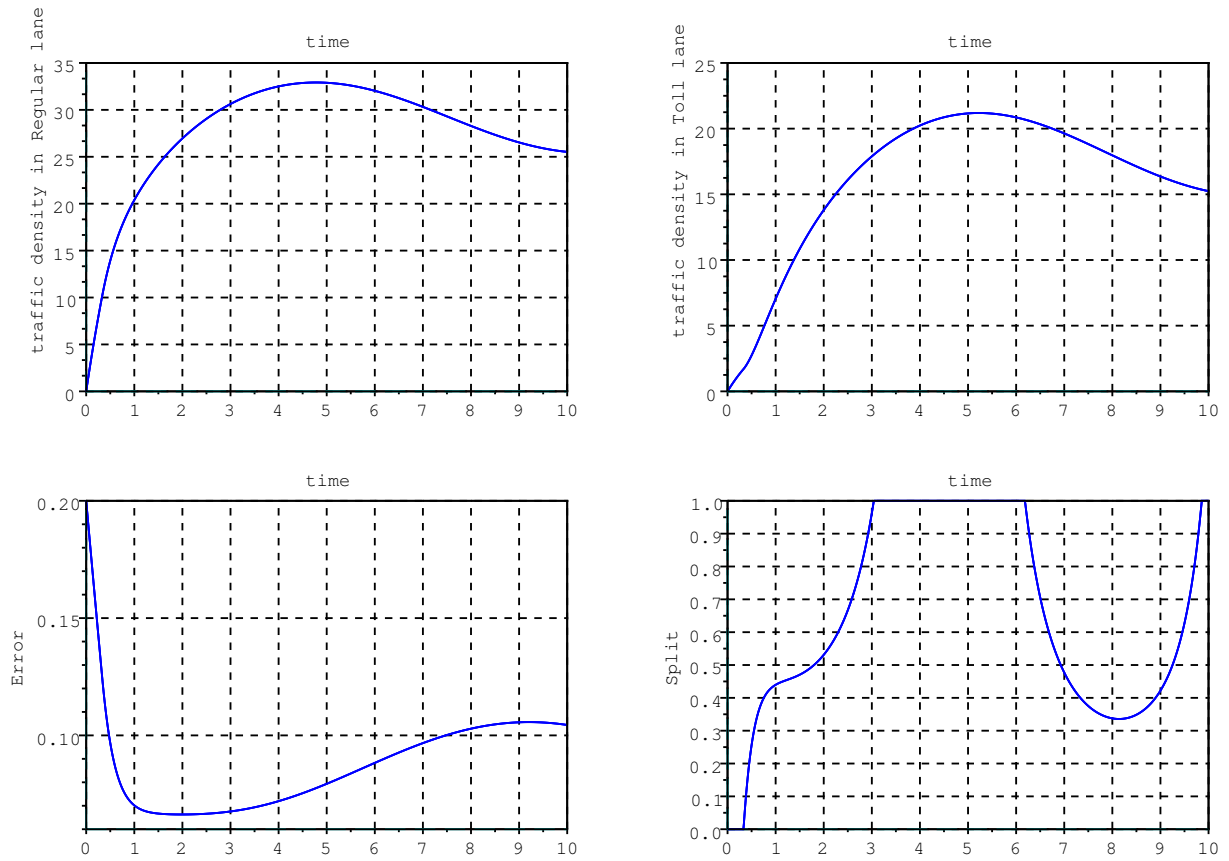


FIGURE 3 Feedback Tolling Results

TABLE 4 Simulation Parameters

SYMBOL	VALUE
β	0.75
S_T	6
ρ_m	120
v_f	1
γ	1.2
L_T	1
L_R	1

127 CONCLUSIONS

128 In this paper we formulated the mathematical models for different scenarios of tolling. The models allowed
 129 the flexibility for having RF tags based lanes and also regular lanes. The paper showed how to build models
 130 in a modular fashion to include the needed features. The models were then used to design real-time feedback
 131 controller using feedback linearization technique to regulate the traffic in the different lanes (or routes).
 132 Simulation software was developed using Scilab to show the robustness and performance of the algorithm,
 133 and it provided the validation for the control design.

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