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A Three-objective User Equilibrium Model: Time Surplus Maximisation under Uncertainty

Judith Y.T. Wang

School of Civil Engineering & Institute for Transport Studies

University of Leeds, Woodhouse Lane, Leeds LS2 9JT, UK

email: j.y.t.wang@leeds.ac.uk

Matthias Ehrgott*

Department of Management Science, Lancaster University Management School

Bailrigg, Lancaster LA1 4YX, UK

email: m.ehrgott@lancaster.ac.uk

Abstract

In this paper, we propose a user equilibrium model considering the three most important factors influencing route choice behaviour in a road network, namely, travel time, travel time reliability and monetary cost. We further develop the time surplus maximisation bi-objective user equilibrium (TSmaxBUE) model and incorporate the concept of travel time budget to model how users might react to uncertainty induced by day-to-day variability in travel time caused by traffic incidents. This results in a three-objective user equilibrium model, which has a possibly infinite set of equilibrium flows. To compute equilibrium flows, we introduce time budget surplus (TBS) defined as the maximum travel time a user is willing to spend minus the actual time budget required for a desired level of travel time reliability. At equilibrium, for each origin-destination (O-D) pair, all individuals are travelling on the path with the highest TBS value among all the efficient paths between this O-D pair. This becomes a time budget surplus maximisation three-objective user equilibrium model (TBSmaxTUE). We show that the TBSmaxTUE model is a special case of three-objective user equilibrium considering minimisation of expected travel time, travel time variance and toll (monetary cost) as objectives. We illustrate the model and our results on a small network.

Keywords: Traffic assignment; route choice; equilibrium problem; multi-objective optimisation; travel time reliability.

1 Introduction

It is well known that the three most important criteria affecting route choice behaviour are travel time, travel time reliability and monetary cost. This observation is drawn from numerous studies in the

*Corresponding author. Tel.: +44 113 343 3259 ; Fax: +44 113 343 3265.

literature. From a route choice survey, Abdel-Aty *et al.* (1995) identified the three most important factors to be: (1) shorter travel time (ranked as the first reason by 40% of respondents); (2) travel time reliability (32%); and (3) shorter distance (31%). Note that although the effect of monetary cost was not considered explicitly in this study, the third most important factor, i.e. distance, is directly related to vehicle operating cost for the trip. From numerous road pricing studies in the US (e.g. Lam & Small 2001, Liu *et al.* 2004, Brownstone & Small 2005) and in Australia (Li *et al.* 2010), it is also clear that the value of reliability (VOR) estimates, when travel time is represented by its median and unreliability of travel time as the difference between the 90th percentile and the median of travel time, can be as high as the value of time (VOT). For some user groups, VOR was found to be even higher than VOT, e.g. the VOR for women is 39.5% higher than the VOT in the best fitted model in Lam & Small (2001). In order to model route choice behaviour realistically, there is no doubt we should consider the influence of all these three factors.

Network equilibrium models have been studied for decades since the discovery of Wardrop's principles. Combinations of some of the three most important route choice factors mentioned above have also been considered in many studies (e.g. Florian 2006, Lo *et al.* 2006, Watling 2006). Florian (2006) summarised the two approaches employed in practice, when toll and travel time are considered: (1) models based on generalised cost path choice; and (2) models based on explicit choice of tolled facilities. These two approaches follow the principles of the two classical traffic assignment models in the literature, namely, the user equilibrium (UE) model and the stochastic user equilibrium (SUE) model.

Lo *et al.* (2006) formulated a multi-class equilibrium model by considering travel time and travel time reliability, combined in a single objective as minimising travel time budget, which is defined as the expected travel time plus a travel time margin (or buffer time), with the travel time margin being dependent on the level of risk aversion of each user class. Watling (2006) proposed a late arrival penalised UE (LAPUE) which assumes users minimise a composite path disutility, incorporating the generalised cost plus a late arrival penalty. Watling (2006) assumes that travellers make their route choice decision with a longest possible travel time in mind for their journey. If this is exceeded, the inconvenience incurred will be modelled by the penalty component of the disutility function. In principle, Watling's and Lo *et al.*'s models can be applied in a manner to consider all three factors, with toll/monetary cost and travel time considered as *generalised travel time*, and travel time reliability in the penalty component or travel time and travel time reliability in the time budget plus the time equivalent of monetary cost, respectively.

All existing network equilibrium models share one common feature, i.e. all the factors considered are combined into a single objective. As discussed in Wang & Ehrgott (2013), the linear combination of time and cost, for instance, implies that a certain class of *efficient* paths will have zero flow at the equilibrium solution, and hence might not be able to model route choice behaviour realistically. Similarly, Wang *et al.* (2014) show that Lo *et al.*'s and Watling's models might miss out some rational route choices in a similar manner.

To address this issue, Wang & Ehrgott (2013) propose a bi-objective approach to overcome the

problem, without assuming the two objectives of travel time and travel time reliability being additive as in Dial (1996). Wang & Ehrgott introduce the concept of time surplus maximisation to model route choice behaviour in a tolled road network considering two objectives: (1) minimise travel time; and (2) minimise toll cost. Time surplus is defined as the maximum time users are willing to spend minus the actual travel time. All users would want to maximise *time surplus* based on the two objectives. This becomes the time surplus maximisation (bi-objective) equilibrium model (TSmaxBUE). At equilibrium, only the paths with the maximal time surplus will have positive flow. Differences in preferences among travellers between each origin-destination (O-D) pair are modelled by indifference curves, which represent the maximum time a user is willing to spend for a given toll. The TSmaxBUE model also satisfies the condition for bi-objective user equilibrium as defined in Wang *et al.* (2010), whereby no user can decrease travel time, or toll, or both without worsening the other and vice versa. As demonstrated in Wang & Ehrgott (2013), the TSmaxBUE model is a proper generalisation of the UE model with additive generalised cost functions.

On the other hand, Wang *et al.* (2014) consider travel time and travel time reliability with a bi-objective approach. Wang *et al.* propose a general network equilibrium model, travel time reliability bi-objective user equilibrium (TTR-BUE) model, considering two objectives: (1) minimise travel time; and (2) minimise travel time variability. Wang *et al.* prove that existing travel time reliability models, including Lo *et al.*'s and Watling's models, as well as their bi-objective counterparts are special cases of TTR-BUE. For example, the bi-objective version of Lo *et al.*'s model has the two objectives (1) minimise expected travel time and (2) minimise travel time budget. This is a special case of TTR-BUE.

In this paper, we propose the first multi-objective network equilibrium model considering all the three most important objectives affecting route choice. We further develop the TSmaxBUE model from Wang & Ehrgott (2013), incorporating findings from Wang *et al.* (2014), to include the consideration of travel time reliability. We note that multi-objective user equilibrium models in general admit an infinite number of equilibrium flows. Despite the drawback of losing uniqueness of equilibrium solutions, this allows the verification that observed flows are consistent with assumptions on the user objectives, independent of any parameters used to convert the multiple objectives into a single cost function. In order to be able to compute the set of equilibrium flows, we must, however, first address the question of computing single equilibrium flows. In this paper we propose a model that makes this possible.

The rest of the paper is organised as follows. Section 2 introduces notation and presents the notion of user equilibrium models of travel assignment. We review the concepts of generalised cost, travel time reliability and travel time budget user equilibrium. We also state a small three-link example that we use throughout the paper to illustrate results. In Section 3, we give a general definition of multi-objective user equilibrium and prove some general results. Section 4 is dedicated to a summary of the travel time surplus maximisation bi-objective user equilibrium model. In Section 5, the main results of the paper on a generalisation of the TSmaxBUE model and the incorporation of travel time reliability through the travel time budget equilibrium model are discussed and the relationships of this

time budget surplus maximisation model to multi-objective user equilibrium models are investigated. We also present an NCP formulation of the time budget surplus maximisation model. Section 6 explores the new model on the three-link example in detail and Section 7 concludes the paper with some general comments and suggested avenues for future research.

2 User Equilibrium

In this section, we briefly review equilibrium models of traffic assignment and introduce notation, as well as an example, that are used throughout the paper.

2.1 Traditional Traffic Assignment Models

Let $G = (N, A)$ denote a network, where N is a finite set of $|N|$ nodes and $A \subset N \times N$ is a set of $|A|$ arcs or links. Moreover, let $Z \subset N \times N$ be a set of origin-destination pairs (O-D pairs) and for all $p \in Z$, let D_p denote the demand for travel between the origin and destination of O-D pair p . Equilibrium models attempt to determine the amount of traffic f_a on all links $a \in A$ under some assumptions on the behaviour of road users. One of these assumptions is that road users choose a path k^* between their origin and destination that minimises a non-negative cost function C_k :

$$k^* \in \operatorname{argmin}\{C_k : k \in K_p\},$$

where K_p is the set of all simple paths from the origin of O-D pair p to its destination.

Let δ_a^k be an indicator with $\delta_a^k = 1$ if and only if link a is contained in path k and 0 otherwise. Then $f_a = \sum_{p \in Z} \sum_{k \in K_p} \delta_a^k F_k$, where F_k is the flow on path $k \in K_p$. The cost $C_k(\mathbf{F})$ of path k may depend on the entire vector $\mathbf{F} = (F_1, \dots, F_{|K|})$ of flows on all paths $k \in K := \cup_{p \in Z} K_p$. The user equilibrium condition of Wardrop's first principle states that the costs of all used paths are equal and not greater than the cost which would be experienced by a single user on any unused route.

Let $U_p := \min_{k \in K_p} C_k(\mathbf{F})$ denote the minimum cost of any path for O-D pair $p \in Z$. Then, following e.g. Florian & Hearn (1995), the user equilibrium condition can be written mathematically as follows: Path flow vector \mathbf{F}^* is an *equilibrium flow* if \mathbf{F}^* satisfies conditions (1) – (5):

$$F_k^*(C_k(\mathbf{F}^*) - U_p) = 0 \quad \text{for all } k \in K_p \text{ and all } p \in Z, \quad (1)$$

$$C_k(\mathbf{F}^*) - U_p \geq 0 \quad \text{for all } k \in K_p \text{ and all } p \in Z, \quad (2)$$

$$\sum_{k \in K_p} F_k^* - D_p = 0 \quad \text{for all } p \in Z, \quad (3)$$

$$F_k^* \geq 0 \quad \text{for all } k \in K, \quad (4)$$

$$U_p \geq 0 \quad \text{for all } p \in Z. \quad (5)$$

Equation (1) states that if flow on path k is positive then the cost $C_k(\mathbf{F}^*)$ has to be minimal, whereas if $C_k(\mathbf{F}^*) > U_p$ then the flow on path k must be 0. Equation (2) says that the costs of all paths in K_p are greater than or equal to the minimum U_p . Equation (3) guarantees that demand is

satisfied, whereas equations (4) and (5) postulate non-negativity of flow and cost. Note that For future use, let us introduce

$$\Omega := \{\mathbf{F} : \mathbf{F} \text{ satisfies (3) – (4)}\} \quad (6)$$

to denote the set of all feasible path flow vectors \mathbf{F} .

Existence of a solution of the network equilibrium model (1) – (5) is guaranteed if the path cost functions $C_k(\mathbf{F})$ are all positive and continuous. If, in addition, the vector of path cost functions $C : \mathbb{R}_{|K|} \rightarrow \mathbb{R}_{|K|}$ is strictly monotone, then equilibrium link flows are unique, see e.g. Smith (1979) for additive cost functions and Gabriel & Bernstein (1997) for non-additive cost functions.

In this paper, we will consider equilibrium problems with multiple user classes. We use an index $m \in M$ to indicate user class. Equilibrium models with multiple user classes split the demand D_p for each OD pair p into separate demands D_{pm} for each user class, hence $D_p = \sum_{m \in M} D_{pm}$. Consequently, flow vector \mathbf{F} records flow F_{km} for each path k and each user class m , U_{pm} denotes the minimum cost of a path between OD pair p and user class m and path cost functions C_{km} will differ by user class. The equilibrium conditions (1) – (5) need to be satisfied for each user class.

2.2 Travel Time, Generalised Cost and Generalised Time

The most important cost function is travel time. In this paper, we use the common Bureau of Public Roads (1964) function to model the relation between travel time and traffic flow on any link $a \in A$, i.e.

$$t_a(f_a) = t_a^0 \left[1 + \beta \left(\frac{f_a}{C_a} \right)^n \right], \quad (7)$$

where t_a^0 is the free-flow travel time on link a , C_a is the practical capacity of link a in vehicles per time unit, and β, n are function parameters. If the cost function C_k considered in (1) – (5) is path travel time, then

$$C_k(\mathbf{F}) = T_k(\mathbf{F}) := \sum_{a \in k} t_a(f_a) \quad (8)$$

for all $k \in K$.

The best known algorithm used to solve conventional traffic assignment is the Frank-Wolfe algorithm (Frank & Wolfe 1956) but many others such as path equilibration (Dafermos & Sparrow 1969), gradient projection (Jayakrishnan *et al.* 1994), projected gradient (Florian *et al.* 2009), and other methods (e.g. Bar-Gera (2010)) have been proposed, see e.g. Perederieieva *et al.* (2015).

Many researchers have suggested more general cost functions than travel time, see e.g. Chen *et al.* (2010), Larsson *et al.* (2002). Most often $C_k(\mathbf{F})$ takes the form of a *generalised cost function* that incorporates a linear combination of travel time and a monetary component (Dial 1996, Leurent 1993). An additive generalised cost function is of the form

$$C_k(\mathbf{F}) = M_k(\mathbf{F}) + \alpha T_k(\mathbf{F}), \quad (9)$$

where $M_k(\mathbf{F})$ is the monetary cost associated with path k . This may be composed of different factors such as toll cost and vehicle operating costs. In addition, α is a value of time, i.e. it converts the

travel time $T_k(\mathbf{F})$ into a monetary value. To solve traffic assignment problems with generalised cost function (9), one can apply the same algorithms as for conventional traffic assignment, depending on the properties of function $M_k(\mathbf{F})$. We note that some authors also consider (9) with a continuous distribution of α , (Dial 1996, Leurent 1993).

Larsson *et al.* (2002) on the other hand considers a nonlinear “generalised time” function of the form

$$C_k(\mathbf{F}) = T_k(\mathbf{F}) + G_k(M_k(\mathbf{F})), \quad (10)$$

where $G_k : \mathbb{R} \rightarrow \mathbb{R}$ is a non-negative increasing function, the time-equivalent of money.

2.3 Travel Time Reliability and Travel Time Budget Equilibrium

To model the influence of travel time and travel time reliability on route choice behaviour, there are two main theoretical frameworks, as categorised in Li *et al.* (2010) and Carrion & Levinson (2012), namely, the mean-variance model (Jackson & Jucker 1982) and the scheduling model (Small 1982). Reliability based UE models were further developed based on these two concepts by Lo *et al.* (2006) and Watling (2006), respectively.

In this paper, we are interested in travel time variability due to variations in link capacity caused by day-to-day incidents. We assume that link capacity distributions are independent. To model the influence of such variability of travel time, Lo *et al.* (2006) formulated a multi-class equilibrium model considering the minimisation of travel time budget, defined as the expected travel time plus a travel time margin (or buffer time), with the travel time margin being dependent on the level of risk aversion of each user class, as shown in Equation (11),

$$B_{km}(\mathbf{F}) = E(T_k(\mathbf{F})) + \lambda_m \sigma_{T_k}(\mathbf{F}) \quad (11)$$

for all $k \in K_p, m \in M$, and $p \in Z$, where $B_{km}(\mathbf{F})$ is the travel time budget; $T_k(\mathbf{F})$ is the random variable of travel time on route k between origin and destination of OD pair p ; $E(T_k(\mathbf{F}))$ and $\sigma_{T_k}(\mathbf{F})$, respectively, are the mean and standard deviation of $T_k(\mathbf{F})$. λ_m is a parameter associated with the level of risk aversion of user class m . We adopt the concept of travel time budget as a mechanism to deal with travel time variability in this paper, because we focus on modelling the travel behaviour of road users in response to the day-to-day variations in travel time induced by disruptions on a minor scale, caused by traffic incidents. We, therefore, adopt the results from Lo & Tung (2003), summarised as follows for completeness.

Assuming that link capacity follows a uniform distribution, defined by an upper bound (the design capacity) and a lower bound (the worst-degraded capacity), which is a fraction, ϕ_a , of the design capacity, \bar{c}_a , i.e.

$$C_a \sim U(\phi_a \bar{c}_a, \bar{c}_a), \quad (12)$$

Lo & Tung (2003) have derived that, due to the central limit theorem for paths with many links, the

path travel time is normally distributed with mean and standard deviation that can be written as

$$E(T_k) = \sum_a \left[\delta_a^k \cdot E(t_a) \right], \quad (13)$$

$$\sigma_{T_k} = \sqrt{\sum_a [\delta_a^k \cdot \text{var}(t_a)]}. \quad (14)$$

By applying the assumption of uniformly distributed arc capacity as expressed in Equation (12), the mean and standard deviation of the route travel time distribution are

$$E(T_k) = \sum_a \left\{ \delta_a^k \cdot \left[t_a^0 + \beta t_a^0 f_a^n \frac{1 - \phi_a^{1-n}}{\bar{c}_a^n (1 - \phi_a) (1 - n)} \right] \right\}, \quad (15)$$

$$\sigma_{T_k} = \sqrt{\sum_a \left[\delta_a^k \cdot \beta^2 (t_a^0)^2 f_a^{2n} \left\{ \frac{1 - \phi_a^{1-2n}}{\bar{c}_a^{2n} (1 - \phi_a) (1 - 2n)} - \left[\frac{1 - \phi_a^{1-n}}{\bar{c}_a^n (1 - \phi_a) (1 - n)} \right]^2 \right\} \right]}. \quad (16)$$

The travel time budget model of Lo *et al.* (2006) is a multi-user class equilibrium model which considers both the expected travel time $E(T_k)$ and the variability of travel time, as measured by σ_{T_k} with users in class m minimising their travel time budget $B_k = E(T_k) + \lambda_m \sigma_{T_k}$. Mathematically, λ_m can be related to the probability ρ_m that a trip arrives within the travel time budget,

$$P \{ T_k \leq B_k = E(T_k) + \lambda_m \sigma_{T_k} \} = \rho_m. \quad (17)$$

After rearranging (17), we have

$$P \left(S_{T_k} = \frac{T_k - E(T_k)}{\sigma_{T_k}} \leq \lambda_m \right) = \rho_m. \quad (18)$$

Note that S_{T_k} on the left hand side of Equation (18) is the standard normal variate of T_k , $S_{T_k} \sim N(0, 1)$.

To summarise, the path cost functions that we shall use in this paper are

$$\begin{aligned} T_k(\mathbf{F}) &:= \sum_{a \in k} t_a(f_a); \text{ travel time on path } k; \\ E(T_k(\mathbf{F})) &\text{ defined by (15); expected path travel time on path } k; \\ \sigma_{T_k(\mathbf{F})} &\text{ defined by (16); standard deviation of path travel time on path } k; \\ B_{km}(\mathbf{F}) &:= E(T_k(\mathbf{F})) + \lambda_m \sigma_{T_k(\mathbf{F})}; \text{ path travel time budget of user class } m \text{ on path } k; \\ \tau_k(\mathbf{F}) &:= \sum_{a \in k} \tau_a; \text{ toll on path } k; \\ M_k(\mathbf{F}) &:= \sum_{a \in k} m_a; \text{ monetary cost on path } k. \end{aligned}$$

Notice that we assume that toll and monetary cost in general are independent of flows and that only the time budget B_{km} varies by user class through the risk aversion parameter λ_m .

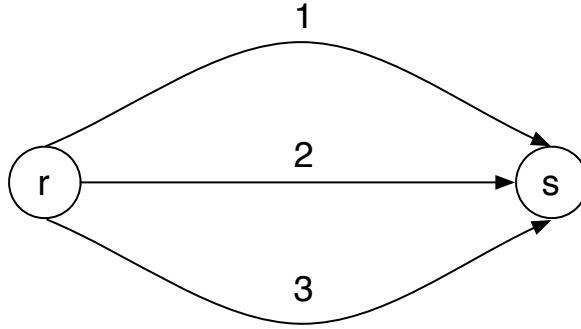


Figure 1: A three-link example network.

2.4 A Three-link Example

We will demonstrate the results of our research using a small three-link network, which is shown in Figure 1, where the link parameters are specified in Table 1. The parameters of the travel time function, Equation (7), are $\beta = 0.15$ and $n = 4$. The total demand is assumed to be fixed at 15,000 vehicles per hour. The free flow travel time t_a^0 , capacity C_a , reliability parameter ϕ_a and two values of tolls τ_a^1 and τ_a^2 that we use later are summarised in Table 1. We note that we shall use toll as the only monetary cost. Further parameters will be specified as necessary in the context of the examples illustrating specific results.

Table 1: Route characteristics of the three-link network.

Route	Type	Distance	t_a^0	C_a	ϕ_a	τ_a^1	τ_a^2
a		(km)	(mins)	(veh/hr)			
1	Expressway	20	12	4000	0.5	40	20
2	Highway	50	30	5400	0.7	20	40
3	Arterial	40	40	4800	0.9	0	0

3 Multi-objective User Equilibrium

In Sections 2.2 and 2.3, we have seen that equilibrium models considering the minimisation of generalised cost such as Equation (9) or travel time budget such as in Equation (11) do in fact consider the minimisation of the weighted sum of several objective functions, namely travel time $T_k(\mathbf{F})$ and monetary cost $M_k(\mathbf{F})$ in generalised cost user equilibrium (9), and expected travel time $E(T_k(\mathbf{F}))$ and standard deviation of travel time $\sigma_{T_k(\mathbf{F})}$ in travel time budget user equilibrium (11). As pointed out in Section 1, in our previous work (Wang & Ehrgott 2013, Wang *et al.* 2014) we have shown that this model assumption of a linear combination of objective functions raises the possibility that

some routes that are rational choices must necessarily have zero flow in an equilibrium solution. To overcome this limitation, we have studied bi-objective user equilibrium models regarding (toll) cost and travel time in Wang & Ehrgott (2013) and expected value and standard deviation of travel time in Wang *et al.* (2014), respectively. In this paper, we extend our analysis to multi-objective equilibrium models considering expected value and standard deviation of travel time as well as monetary (toll) cost. In this section, we present general definitions and basic results on multi-objective network equilibrium models. For ease of notation, we assume one user class. Generalisations to multiple user classes are straightforward.

First, we repeat the definitions of efficient paths and (weak) multi-objective user equilibrium from Wang & Ehrgott (2013). In the remainder of this section, let $G = (N, A)$ be a network, $Z \subset N \times N$ be a set of O-D pairs and for all $p \in Z$, let D_p be the demand of O-D pair p . Let $C_k^{(i)}(\mathbf{F}), i = 1, \dots, r$ be r cost functions of path k and let $\mathbf{C}_k(\mathbf{F})$ denote the cost vector of path k . We assume that all cost functions $C_k^{(i)}(\mathbf{F}), i = 1, \dots, r$ are positive and continuous.

Definition 1 Let $\mathbf{F} \in \Omega$ be a feasible path flow vector.

1. Path k is efficient, if there is no path $k' \in K_p$ such that $C_{k'}^{(i)}(\mathbf{F}) \leq C_k^{(i)}(\mathbf{F})$ for all $i = 1, \dots, r$ and $C_{k'}^{(j)}(\mathbf{F}) < C_k^{(j)}(\mathbf{F})$ for at least one $j \in \{1, \dots, r\}$. If such a path k' exists then path k' dominates path k and path cost vector $\mathbf{C}_{k'}(\mathbf{F})$ dominates path cost vector $\mathbf{C}_k(\mathbf{F})$.
2. Path k is weakly efficient, if there is no path $k' \in K_p$ such that $C_{k'}^{(i)}(\mathbf{F}) < C_k^{(i)}(\mathbf{F})$ for all $i = 1, \dots, r$. If such a path k' exists then path k' strictly dominates path k and path cost vector $\mathbf{C}_{k'}(\mathbf{F})$ strictly dominates path cost vector $\mathbf{C}_k(\mathbf{F})$.

Definition 2 Let $\mathbf{F}^* \in \Omega$ be a feasible path flow vector.

1. Flow vector $\mathbf{F}^* \in \Omega$ is a multi-objective equilibrium flow, if whenever $\mathbf{C}_k(\mathbf{F}^*)$ dominates $\mathbf{C}_{k'}(\mathbf{F}^*)$ for $k, k' \in K_p$ for any $p \in Z$ then $F_{k'} = 0$.
2. Flow vector $\mathbf{F}^* \in \Omega$ is a weak multi-objective equilibrium flow, if whenever $\mathbf{C}_k(\mathbf{F}^*)$ strictly dominates $\mathbf{C}_{k'}(\mathbf{F}^*)$ for $k, k' \in K_p$ for any $p \in Z$ then $F_{k'} = 0$.

We present some basic general results concerning multi-objective equilibrium flows.

Proposition 1 A multi-objective equilibrium flow exists.

Proof: Because of the assumption of positive and continuous path cost functions, we know that the generalised path cost functions

$$G_k(\mathbf{F}) := \sum_{i=1}^r C_k^{(i)}(\mathbf{F})$$

are positive and continuous. Hence an equilibrium flow \mathbf{F}^* with respect to G_k exists. We show that this equilibrium flow \mathbf{F}^* is a multi-objective equilibrium flow.

Assume to the contrary that there is an O-D pair p and two paths $k, k' \in K_p$ with positive flow such that $\mathbf{C}_{k'}(\mathbf{F}^*)$ dominates $\mathbf{C}_k(\mathbf{F}^*)$. Then $G_{k'}(\mathbf{F}^*) < G_k(\mathbf{F}^*)$, contradicting the fact that \mathbf{F}^* is an equilibrium flow with respect to G_k . \square

In user equilibrium models, the question of uniqueness of equilibrium flows is important. We shall see in Example 1 that in multi-objective user equilibrium, even if all path cost functions are strictly monotone, there are no uniqueness results. We, therefore, proceed to another general result on multi-objective user equilibrium. It states that equilibrium flows remain (at least weak) equilibrium flows as the number of objectives increases.

Theorem 1 *Let $\emptyset \neq R \subset \{1, \dots, r\}$ and let $\mathbf{F}^* \in \Omega$ be a multi-objective equilibrium flow with respect to the cost functions $C_k^{(i)}$ for $i \in R$. Then \mathbf{F}^* is a weak multi-objective equilibrium flow with respect to cost functions $C_k^{(i)}, i = 1, \dots, r$.*

Proof: Assume to the contrary that \mathbf{F}^* is not a weak multi-objective equilibrium flow. Then there exists some OD pair p and two paths k and k' in K_p with positive flow such that $C_{k'}^{(i)}(\mathbf{F}^*) < C_k^{(i)}(\mathbf{F}^*)$ for all $i = 1, \dots, r$. In particular, this implies that $C_{k'}^{(i)}(\mathbf{F}^*) < C_k^{(i)}(\mathbf{F}^*)$ for all $i \in R$. Hence, in the equilibrium problem considering objectives $C_k^{(i)}$ for $i \in R$, cost vector $\mathbf{C}_k(\mathbf{F}^*)$ strictly dominates cost vector $\mathbf{C}_{k'}(\mathbf{F}^*)$ and both $F_k^* > 0$ and $F_{k'}^* > 0$, contradicting the property of multi-objective equilibrium flow for the equilibrium problem considering the objectives in R . \square

Because of the non-uniqueness of multi-objective user equilibrium flows special attention needs to be paid to ways of computing these flows. Theorem 2 provides a general method of scalarisation to compute multi-objective equilibrium flows.

Theorem 2 *Let $s : \mathbb{R}^r \rightarrow \mathbb{R}$ be a strictly monotone and continuous function and let \mathbf{F}^* be an equilibrium flow with respect to the path cost function $s(\mathbf{C}_k(\mathbf{F}^*))$. Then \mathbf{F}^* is a multi-objective equilibrium flow with respect to path cost functions $C_k^{(i)}(\mathbf{F}^*), i = 1, \dots, r$.*

Proof: Let \mathbf{F}^* be an equilibrium flow with respect to the path cost function $s(\mathbf{C}_k(\mathbf{F}^*))$ and assume that it is not multi-objective equilibrium flow with respect to path cost functions $C_k^{(i)}(\mathbf{F}^*), i = 1, \dots, r$. Then there exists some OD pair p and two paths k and k' in K_p with positive flow such that $\mathbf{C}_{k'}(\mathbf{F}^*)$ dominates $\mathbf{C}_k(\mathbf{F}^*)$. Because s is a strictly decreasing function, this implies $s(\mathbf{C}_{k'}(\mathbf{F}^*)) < s(\mathbf{C}_k(\mathbf{F}^*))$, contradicting the assumption. \square

In fact, the basic result of Theorem 2 shows that equilibrium models using generalised cost as in Equation (9), but also the nonlinear model of Larsson *et al.* (2002) using path cost function (10) compute bi-objective user equilibria with respect to the path cost functions $T_k(\mathbf{F})$ and $M_k(\mathbf{F})$.

Example 1 *In the three-link example of Section 2.4, we will consider single- and multi-objective equilibrium flows with respect to the three cost functions toll $\tau_k := \tau_a^1$, expected travel time $E(T_k)$ and standard deviation of travel time σ_{t_k} as defined in (15) and (16). All equilibrium flows are illustrated in Figure 2, which shows Ω represented in \mathbb{R}^2 by values of f_1 and f_2 .*

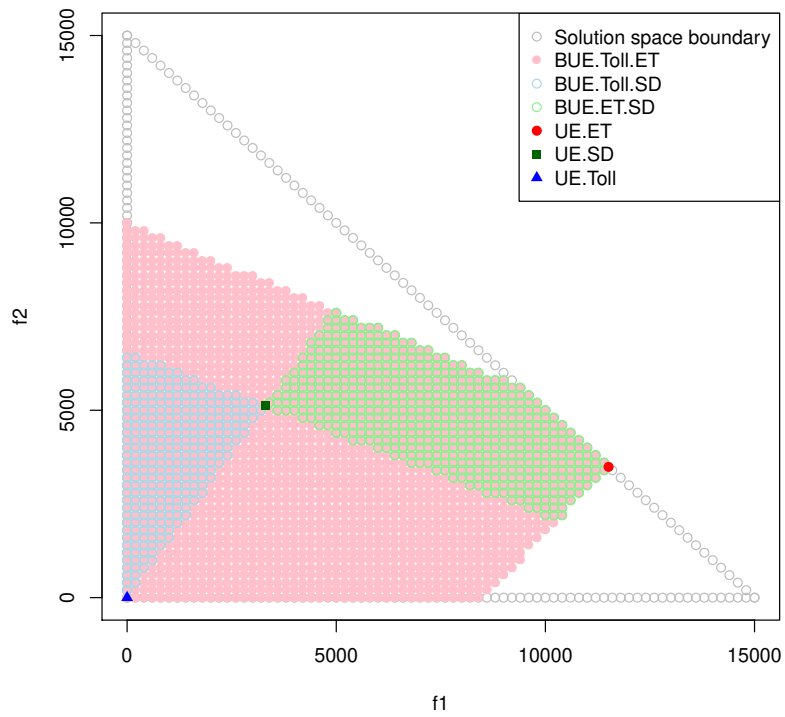


Figure 2: Equilibrium flows with one, two and three objectives for $\tau_1 = 40, \tau_2 = 20, \tau_3 = 0$.

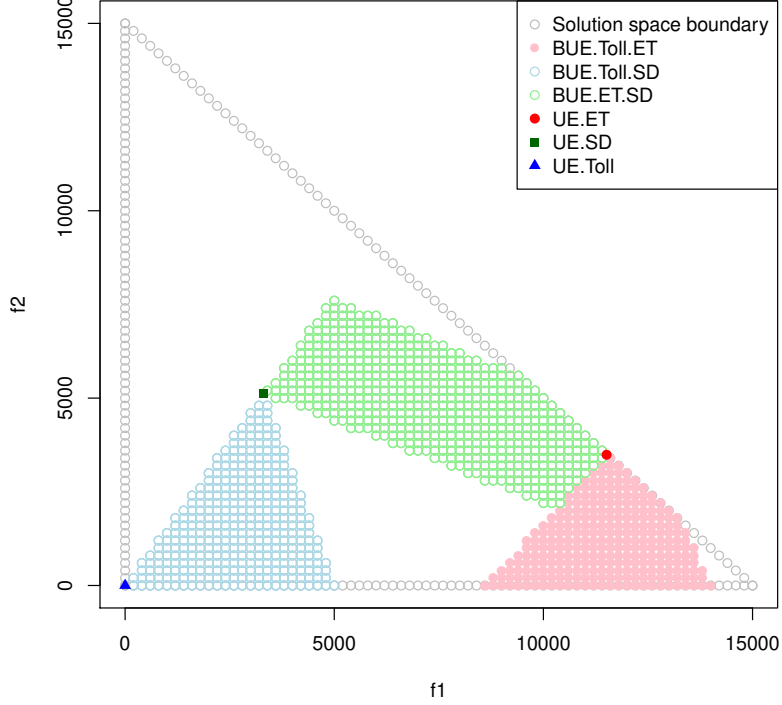


Figure 3: Equilibrium flows with one, two and three objectives for $\tau_1 = 20, \tau_2 = 40, \tau_3 = 0$.

The (unique) single-objective equilibrium flows are shown by the blue triangle (toll, $UE.Toll$), green square (standard deviation of travel time, $UE.SD$), and red circle (expected travel time, $UE.ET$), respectively. All three combinations of two objectives result in regions of bi-objective user equilibrium flows, indicated by pale red ($BUE.Toll.ET$), green ($BUE.Toll.SD$) and blue ($BUE.ET.SD$) circles. Finally, in this particular example, the region of three-objective user equilibrium flows coincides with the bi-objective toll versus expected travel time user equilibrium flows. Because toll ranks the three routes in the opposite way of expected travel time, addition of the third objective does not add any further equilibrium flows. The inclusions stated in Theorem 1 are clearly verified. Notice that the inclusions postulated may or may not be strict.

In a second example, we change the tolls to τ_a^2 , so that $\tau_1 = 20, \tau_2 = 40, \tau_3 = 0$. Here, the toll for the faster but less reliable expressway is lower than the toll for the slower but more reliable highway. This affects the regions of bi-objective and three-objective user equilibrium flows. In Figure 3, the three-objective user equilibrium flows form the convex hull of all the bi-objective user equilibrium regions.

4 The Time Surplus Maximisation Model

In previous work (Wang & Ehrgott 2013, Wang *et al.* 2014) we have investigated bi-objective user equilibrium models considering the path cost functions $T_k(\mathbf{F})$, $\tau_k(\mathbf{F})$, $E(T_k(\mathbf{F}))$, and $\sigma_{T_k}(\mathbf{F})$. Specifically, in Wang & Ehrgott (2013), we introduced the time surplus maximisation equilibrium model as a new model for route choice behaviour in tolled road networks considering the two objective functions travel time $C_k^{(1)}(\mathbf{F}) = T_k(\mathbf{f}) = \sum_{a \in k} t_a(f_a)$, where $t_a(f_a)$ is the travel time function (7) and toll $C_k^{(2)}(\mathbf{F}) = M_k(\mathbf{f}) = \tau_k = \sum_{a \in k} \tau_a$, with exogenously defined link tolls τ_a .

For every O-D pair $p \in Z$, users of class m have an indifference function between toll and time, a function $T_{pm}^{max} : \mathbb{R} \rightarrow \mathbb{R}$ that is strictly decreasing, i.e. $T_{pm}^{max}(\tau_k^1) < T_{pm}^{max}(\tau_k^2)$ if $\tau_k^1 > \tau_k^2$, where $T_{pm}^{max}(\tau)$ is interpreted as the maximum time they would be willing to spend to travel from the origin to the destination of O-D pair p if the toll is τ . The strict decreasing property of T_{pm}^{max} takes into account that users would expect to spend less time in traffic if they need to pay a higher toll.

Time surplus is defined as the time that the user would be willing to spend minus the actual travel time, i.e. given the indifference curves T_{pm}^{max} for all $p \in Z$ and $m \in M$, time surplus for path $k \in K_p$ and user class m is

$$TS_{km}(\mathbf{F}) := T_{pm}^{max}(\tau_k) - T_k(\mathbf{f}) = T_{pm}^{max}\left(\sum_{a \in k} \tau_a\right) - \sum_{a \in k} t_a(f_a). \quad (19)$$

Assuming that users choose the path k^* with maximum time surplus, i.e.

$$k^*(m) = \operatorname{argmax}\{TS_{km}(\mathbf{F}) : k \in K_p\}, \quad (20)$$

we obtain the time surplus maximisation (bi-objective) user equilibrium (TSmaxBUE) condition by using $\Delta_{pm} - TS_{km}(\mathbf{F})$ for $C_{km}(\mathbf{F})$ in the user equilibrium definition of Equations (1) – (5), where Δ_{pm} is a sufficiently large constant, e.g. $\Delta_{pm} = \max_{k \in K_p} T_k(D_p)$, to ensure positivity of $T_{km}(\mathbf{F})$. Notice that it is natural to assume that T_{pm}^{max} is bounded by $\max_{k \in K_p} T_k(D_p)$, the longest travel time if all demand for O-D pair p is on a single path.

In Wang & Ehrgott (2013) we have shown that time surplus maximisation bi-objective user equilibrium characterises all bi-objective user equilibrium flows with respect to the objectives travel time and toll, which we state as Theorem 3. Moreover, in Wang & Ehrgott (2013) we show that generalised cost user equilibrium with objective function (9) is a special case of time surplus maximisation bi-objective user equilibrium. Note that this result also follows from Theorem 2 in Section 3.

Theorem 3 \mathbf{F}^* is a bi-objective equilibrium flow, with respect to objectives $C_k^{(1)}(\mathbf{F}) = T_k(\mathbf{F})$ and $C_k^{(2)}(\mathbf{F}) = \tau_k$ if and only if there exist indifference functions T_{pm}^{max} for all $p \in Z$ and $m \in M$ such that \mathbf{F}^* is also a TSmaxBUE flow.

In this paper, we are interested in a single-objective user equilibrium model that allows to compute three objective user equilibrium flows. In Section 5 we proceed to extend the TSmaxBUE concept to the case of travel time uncertainty using the concept of travel time budget.

5 Time Surplus Maximisation under Travel Time Uncertainty

Let us consider how to apply the time surplus concept to model individual route choices with different views towards travel time reliability.

We first need to introduce variability of travel time in the time surplus function defined in Wang & Ehrgott (2013) as in (19). Here we apply the concept of travel time budget in Lo *et al.*'s model. We define ρ as the desired level of reliability, or equivalently, $1 - \rho$ as the critical level of risk. ρ equals the desired percentage of times that an individual can get to the destination within the allotted travel time. A risk neutral user will have a ρ -value of 50% while a risk averse user might have a ρ -value of 95%. As in Wang *et al.* (2014), we assume that users are all either risk neutral or risk averse, i.e. $\rho \in [0.5, 1)$.

Now we define travel time budget surplus as in Equation (21),

$$TBS_{km}(\mathbf{F}) = T_{pm}^{max}(\tau_k) - T_{km}^\rho(\mathbf{F}), \quad (21)$$

where

$$P(T_k \leq T_{km}^\rho) = \rho. \quad (22)$$

For risk neutral users, T_{km}^ρ will be the median value of the travel time on path k , while for risk averse users, T_{km}^ρ will be higher than the median, depending on the level of risk aversion ρ .

Now let us consider a risk neutral user with a convex indifference curve as shown in Figure 4. The time budget surplus of this user can be determined by comparing the indifference curve with the median travel time curve. In this case, Route 1 will be the optimal choice for this individual as this route has the highest time budget surplus. On the other hand, a risk averse user with the same indifference curve and $\rho = 95\%$, as shown in Figure 5, will consider Route 2 an optimal choice, as this route provides the least negative time budget surplus.

Similarly, we can consider users with a different attitude towards risk, but with a concave indifference curve as shown in Figures 6 and 7. Again, their choices would be different even though they have the same indifference curve. A risk neutral user will find Route 1, with the largest positive time budget surplus, the most attractive, while a risk averse user will choose Route 2 instead as it has the least negative time budget surplus value.

As discussed in Section 2.3, we know that we can guarantee the travel time reliability stated in Equation (22), if we set T_{km}^ρ equal to the time budget $E(T_k) + \lambda_m \sigma_{T_k}$, as presented in Equations (17) and (18). Therefore, to model user equilibrium considering toll, travel time, and travel time reliability we propose a user equilibrium model considering the path *benefit* function

$$TBS_{km}(\mathbf{F}) = T_{pm}^{max}(\tau_k) - E(T_k) - \lambda_m \sigma_{T_k}. \quad (23)$$

Notice that Equation (23) is obtained from Equation (21) using the time budget formula of Equation (11). We consider multiple user classes $m \in M$, where each user class m has its specific indifference curve $T_{pm}^{max}(\tau_k)$ and attitude towards risk, quantified in parameter ρ_m and the corresponding

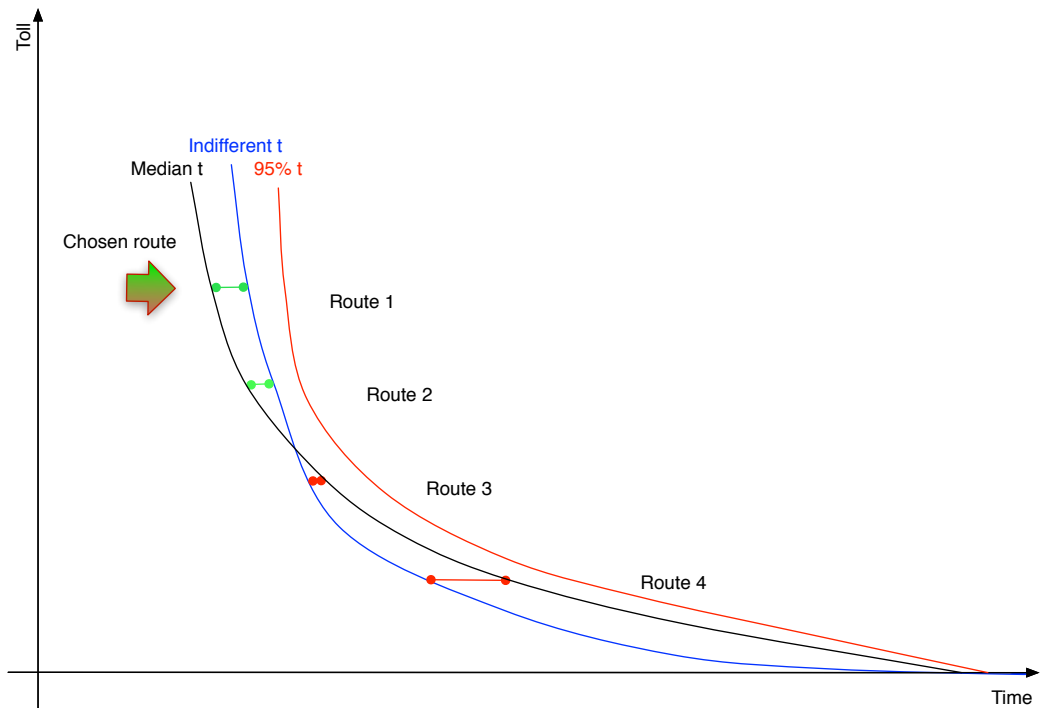


Figure 4: Travel time budget surplus for risk neutral users with convex indifference curve.

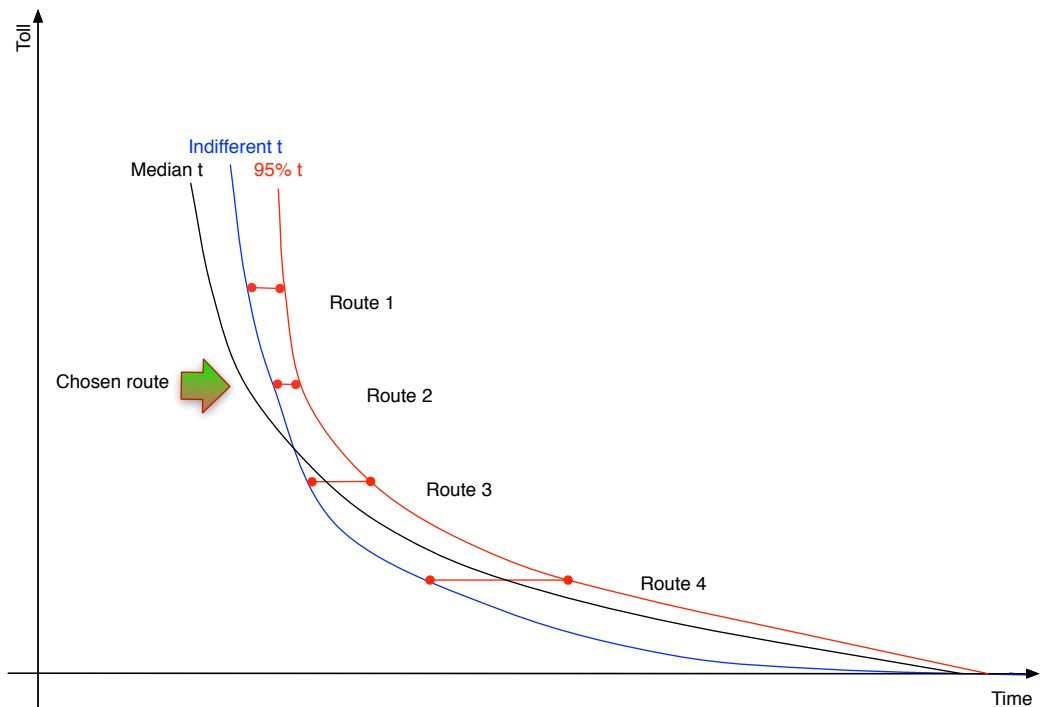


Figure 5: RTravel time budget surplus for risk averse users with convex indifference curve.

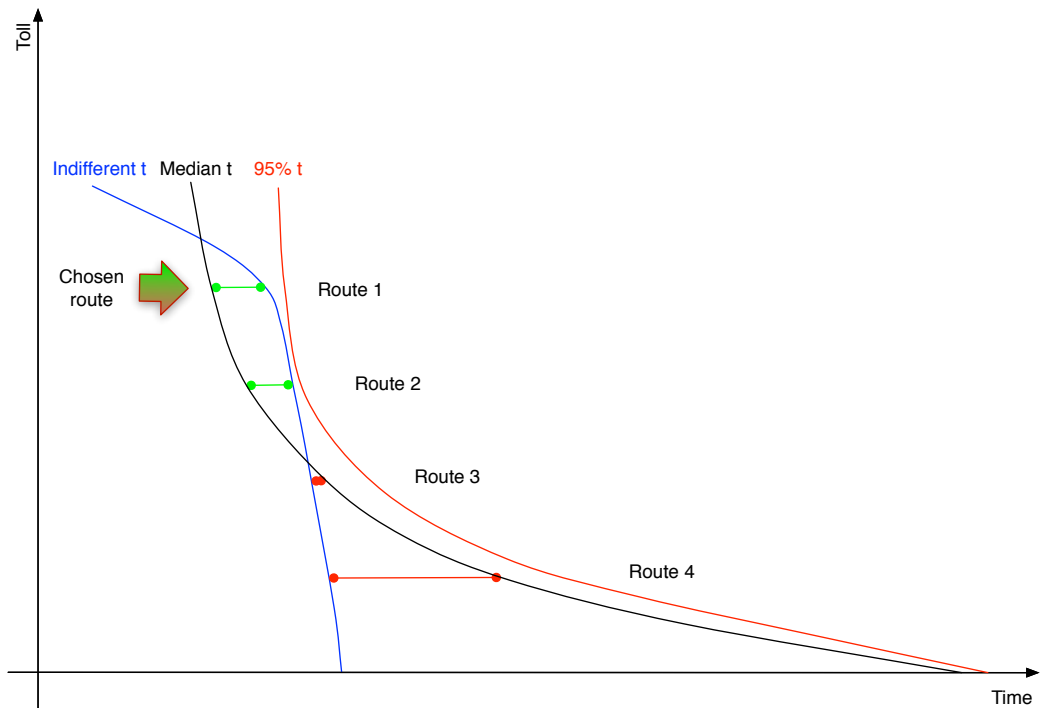


Figure 6: Travel time budget surplus for risk neutral users with concave indifference curve.

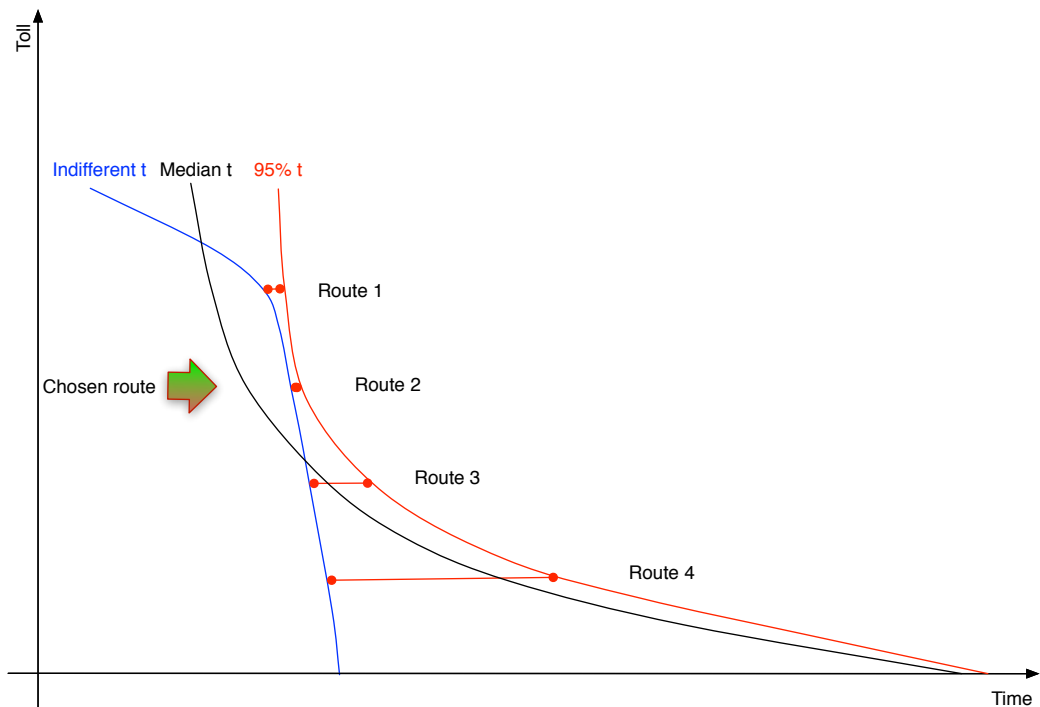


Figure 7: Travel time budget surplus for risk averse users with concave indifference curve.

value λ_m . Hence, our multi-user class equilibrium model incorporating users' trade-offs between travel time and toll as well as attitude towards risk (measured by standard deviation of travel time) is based on the class-specific path benefit function (23), which is turned into a cost function by subtracting it from a sufficiently large constant such as Δ_{pm} as in Section 4, but considering the least capacity $\rho_a C_a$ in its definition. We call this model the *Time Budget Surplus Maximisation (Three-Objective) User Equilibrium* (TBSmaxTUE) model. We call a feasible flow vector $\mathbf{F}^* \in \Omega$ that satisfies the (multi-user) class equilibrium conditions with path cost function $\Delta_{pm} - TBS_{km}(\mathbf{F})$ a TBSmaxTUE flow. The name is justified by Theorems 4 and 5 below, which show that a TBSmaxTUE equilibrium flow is indeed a three-objective use equilibrium flow with respect to the three constituent objectives.

Replicating the result of Theorem 3 we can immediately state the equivalence between time budget surplus user equilibrium and bi-objective user equilibrium considering path cost functions τ_k (toll) and B_{km} (travel time budget).

Theorem 4 *Path flow vector $\mathbf{F}^* \in \Omega$ is a bi-objective equilibrium flow with respect to objectives $C_{km}^{(1)}(\mathbf{F}) = B_{km}(\mathbf{F})$ and $C_{km}^{(2)}(\mathbf{F}) = \tau_k$ if and only if there exist indifference functions T_{pm}^{max} such that \mathbf{F}^* is also a TBSmaxTUE flow.*

We will further link travel time budget surplus equilibrium with the general framework of multi-objective user equilibrium discussed in Section 3. That is, we consider in particular three-objective user equilibrium with cost functions toll, expected travel time, and standard deviation of travel time.

Theorem 5 *Let $\mathbf{F}^* \in \Omega$ be a bi-objective equilibrium flow with respect to path cost functions $C_{km}^{(1)}(\mathbf{F}) = E(T_k(\mathbf{F})) + \lambda_m \sigma_{T_k}(\mathbf{F})$ and $C_{km}^{(2)}(\mathbf{F}) = \tau_k$. Then \mathbf{F}^* is also a three-objective user equilibrium flow with respect to toll, expected travel time, and standard deviation of travel time.*

Proof: Assuming the contrary, there would be an OD pair p , a user class m and two paths $k, k' \in K_p$ with positive flow such that

$$\begin{aligned} E(T_k) &\leq E(T_{k'}), \\ \sigma_{T_k} &\leq \sigma_{T_{k'}}. \\ \tau_k &\leq \tau_{k'} \end{aligned}$$

with at least one of the inequalities being strict. If the strict inequality occurs in the expected travel time or standard deviation component, then we have $E(T_k) + \lambda \sigma_{T_k} < E(T_{k'}) + \lambda \sigma_{T_{k'}}$ (notice that $\lambda > 0$). Together with $\tau_k \leq \tau_{k'}$, the cost vector of path k dominates the cost vector of path k' when considering toll and time budget, contradicting the property of multi-objective equilibrium flow of \mathbf{F}^* with respect to time budget and toll.

The case of strict inequality with toll is completely analogous. □

Example 2 *Once again, we illustrate the results of Theorems 4 and 5 using the three-link example of Section 2.4. Here we consider only a single user class and $\rho = 0.9$. Figures 8 and 9 show the*

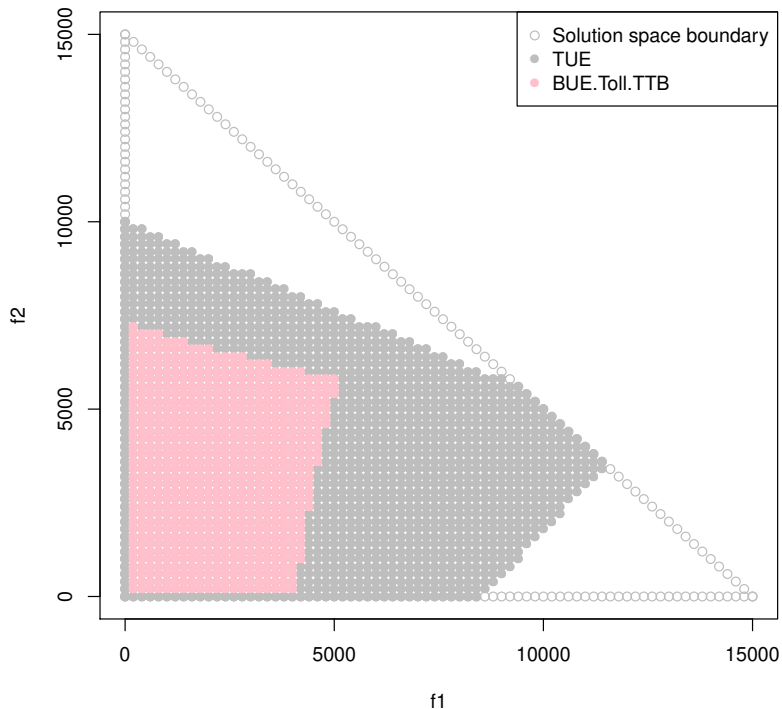


Figure 8: Equilibrium flows for toll versus travel time budget and three objectives, $\tau_1 = 40, \tau_2 = 20, \tau_3 = 0$.

region of three-objective user equilibrium (grey, TUE) and bi-objective user equilibrium for toll versus travel time budget (pale red, BUE.TollTTB) for the cases of toll as in Table 1 (Figure 8) and for $\tau_1 = 20, \tau_2 = 40, \tau_3 = 0$ (Figure 9), see also Example 1.

Figures 8 and 9 clearly show that for fixed ρ and indifference curve, the bi-objective equilibrium flows are (proper) subsets of three-objective equilibrium flows. The reverse, whether for any three-objective user equilibrium flow \mathbf{F}^* there exists an indifference curve and ρ such that \mathbf{F}^* is a bi-objective user equilibrium with cost functions toll and travel time budget, remains an open question for now.

5.1 An NCP Formulation of the Time Budget Surplus Maximisation Model

Recall that the time budget surplus function $TBS_{km}(\mathbf{F})$ defined in Equation (23) is strictly increasing, continuous and positive. Therefore, an equilibrium flow with respect to $TBS_{km}(\mathbf{F})$ can be computed by minimising a gap function related to the nonlinear complementarity formulation associated with the equilibrium problem. We use the function $\Phi(a, b) = \sqrt{a^2 + b^2} - (a + b)$ that was also used by Lo & Chen (2000). Let

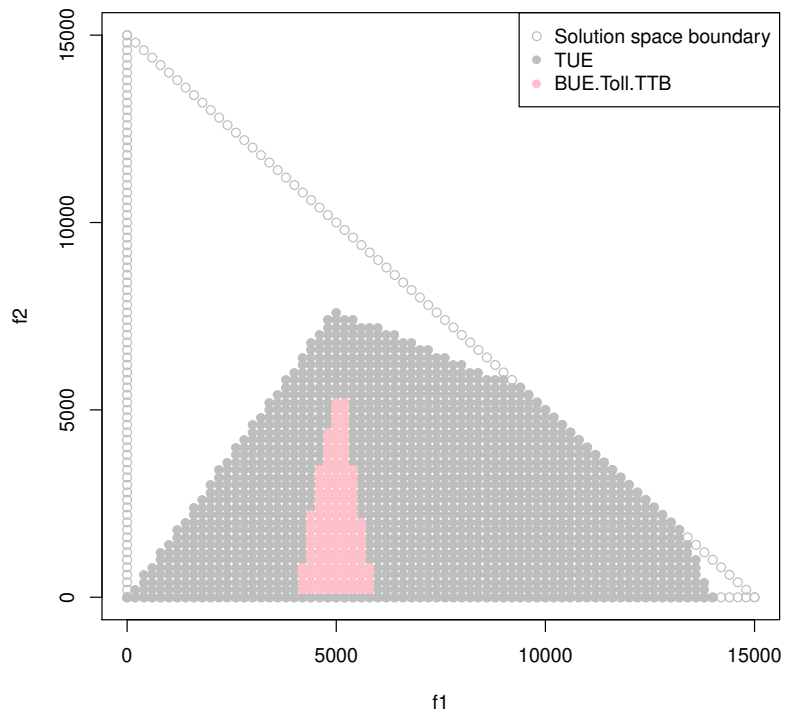


Figure 9: Equilibrium flows for toll versus travel time budget and three objectives, $\tau_1 = 20, \tau_2 = 40, \tau_3 = 0$.

- user class m be defined by a specific indifference curve T_{pm}^{max} and risk attitude ρ_m ;
- D_{pm} be the demand for OD pair p and user class m ;
- f_{kpm} be the flow of user class m on path k for OD pair p ;
- π_{pm} be a variable for the minimal path cost of user class m for OD pair p ; and
- $\eta_{kpm} := \Delta_{km} - TBS_{km}$ be the “cost” of path $k \in K_p$ for user class m , where Δ_{km} is a sufficiently large number.

Then the gap function $G(\mathbf{F})$ defined by Φ is

$$G(\mathbf{F}) := \sum_{m \in M} \sum_{p \in Z} \sum_{k \in K_p} \frac{1}{2} \Phi^2(f_{kpm}, \eta_{kpm} - \pi_{pm}) + \sum_{m \in M} \sum_{p \in Z} \frac{1}{2} \Phi^2\left(\pi_{pm}, \sum_{k \in K_p} -D_{pm}\right) \quad (24)$$

and it holds that \mathbf{F} is an equilibrium flow if and only if $G(\mathbf{F}) = 0$. The variables in the NCP model are the path flows for each OD pair, path, and user class as well as the minimum path costs for each OD pair and user class.

To summarise our results, we have suggested a three-objective user equilibrium model as a general framework to model the influence of the three most important factors influencing route choice behaviour in tolled road networks, namely expected travel time, travel time reliability, and monetary cost. Because such a multi-objective equilibrium model has in general infinitely many equilibrium flows, we have proposed a time surplus maximisation model under uncertainty to operationalise the concept, i.e. to compute three objective user equilibrium solutions. This model combines the time-surplus maximisation equilibrium for route choice modelling in tolled road networks (Wang & Ehrgott 2013) with the time budget concept to model travel time reliability (Lo *et al.* 2006).

6 Numerical Example

In this section, we apply the travel time budget surplus maximisation model to the three-link example of Section 2.4. Since there is only one OD pair, we omit the index p hereafter. We consider six user classes altogether. They are characterised by three indifference curves T_1^{max} , T_2^{max} and T_3^{max} as shown in Table 2. For each indifference curve, we have two user classes, one representing risk neutral users with $\rho_m = 0.50$, and another representing risk averse users with $\rho_m = 0.95$. The tolls are as specified for τ_a^1 in Table 1 and routes are identified by their toll in figures below. The problem has been solved in MATLAB using the NCP formulation described in Section 5.1. The tolls are as specified for τ_a^1 in Table 1 and routes are identified by their toll in figures.

6.1 Equilibrium Flows

The equilibrium flows for the six different user classes are shown in Figure 10. All risk neutral users will travel on the toll-free route. For risk averse users, those with lower maximum time willing to

Table 2: Maximum time willing to spend.

Route k	Indifference curve		
	T_1^{max}	T_2^{max}	T_3^{max}
1	12.5	17.5	22.5
2	32.5	37.5	42.5
3	65.0	75.0	85.0

spend (indifference curves T_1^{max} and T_2^{max}) will choose to pay, while all risk averse users with the highest maximum time willing to spend (indifference curve T_3^{max}) will travel on the toll-free route. This is consistent with our expectations that it would be those who are risk averse and more willing to pay who would travel on the tolled (and more reliable) routes.

6.2 TBSmaxTUE Conditions

To check that the TBSmaxBUE conditions are satisfied, the time budget surplus values for all six user classes are compared at equilibrium flows. For risk neutral users, Figure 11 shows that the toll-free route has the highest time budget surplus for users with all three indifference curves and thus is the only route with positive flow. For risk averse users, on the other hand, Figure 12 shows that the tolled routes for users with indifference curves T_1^{max} and T_2^{max} have the least negative time budget surplus and thus have positive flows. The toll-free route has the highest positive time budget surplus for users with indifference curve T_3^{max} and thus is the only route with positive flow. Note that as expected the time budgets for risk averse users are higher than those for the risk neutral ones.

7 Conclusion and Outlook

In this paper, we propose a novel approach to incorporate the consideration of the three most important factors influencing route choice behaviour in a road network, namely, travel time, travel time reliability and monetary cost. Our proposed model is based on the concept of three-objective user equilibrium, assuming that all users aim to (1) minimise expected travel time; (2) minimise travel time variability; and (3) minimise toll (monetary cost).

We further develop the TSmaxBUE model introduced in Wang & Ehrgott (2013), where we introduced the concept of time surplus maximisation to model user preferences in a general form rather than using a VOT value. We assume that for any given OD pair and a specific toll value, there is a limit on the time that an individual would be willing to spend (forming an indifference curve between toll and time for the individual). Each individual can have his/her own preference represented by this indifference curve between toll and time. Time surplus is defined as the time that an individual is willing to spend minus the actual travel time. Given a set of routes, the one with the highest (or least

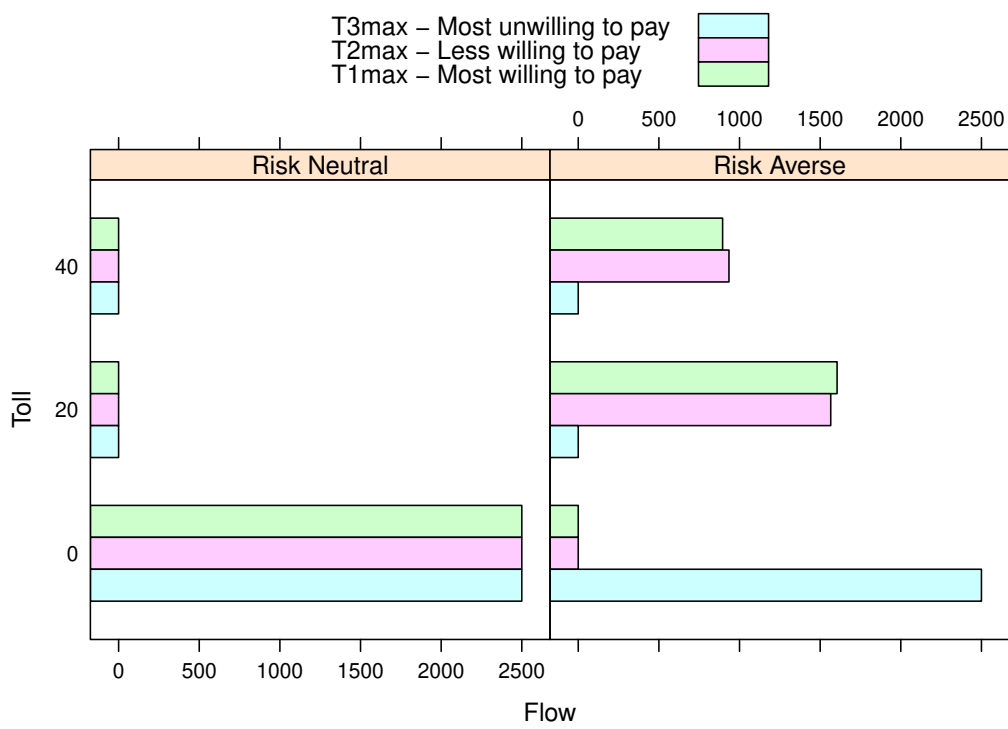


Figure 10: Equilibrium flows.

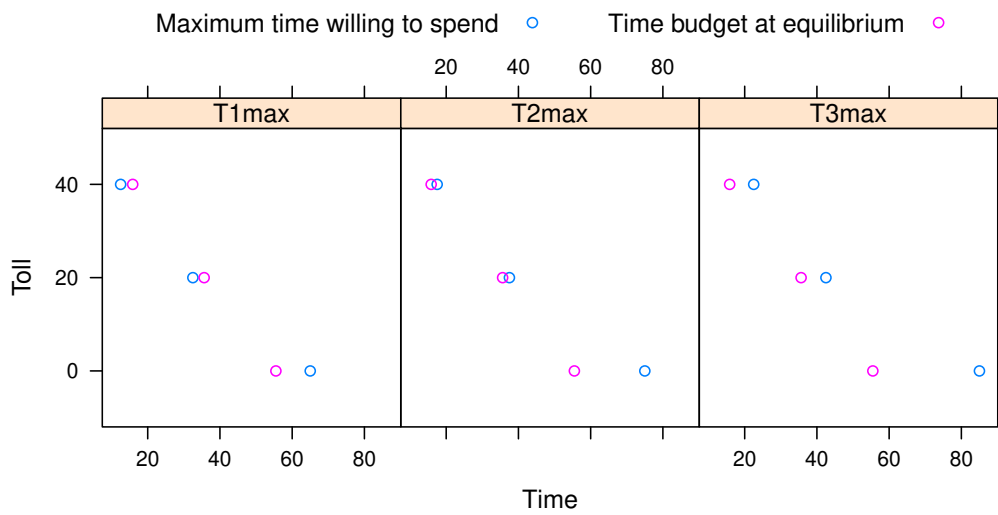


Figure 11: Time budget surplus of risk neutral users.

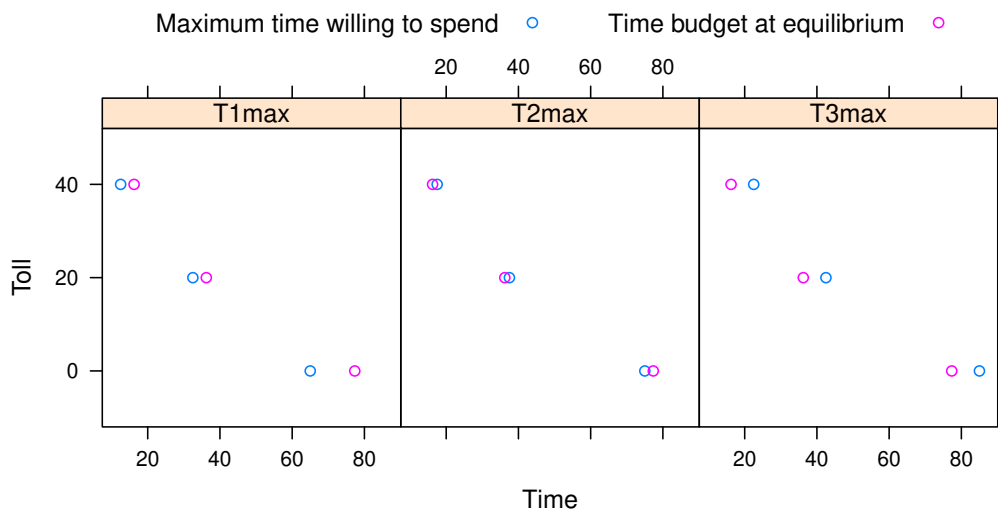


Figure 12: Time budget surplus of risk averse users.

negative) time surplus will be the preferred route for the individual.

We then incorporate the concept of travel time budget (Lo *et al.* 2006) into the time surplus maximisation concept. For a given toll, a user will have two limits, one on the maximum travel time a user is willing to spend, and one on the minimum level of travel time reliability. Here, travel time reliability is defined as the probability of arriving at the destination within the allocated travel time budget. Based on the desired level of reliability, one can determine the travel time budget required on a particular route that will satisfy this level of reliability. Whether this route is attractive or not to an individual would depend on how much this required travel time budget differs from the maximum time the user is willing to spend given a particular toll. We define time budget surplus as the maximum time a user is willing to spend minus the travel time budget required for a desired level of travel time reliability, which is effectively a measure of the level of attractiveness. The route with the highest (or least negative) time budget surplus would be the most attractive option. We assume that each individual can be characterised by (1) an indifference curve between toll and maximum time willing to spend; and (2) the desired minimum level of reliability. In this way, we can determine the choice of each individual based on his/her own preference, modelled by the two characteristics. This becomes our proposed TBSmaxTUE model.

This concept can also be applied in a toll-free road network. As mentioned earlier, Abdel-Aty *et al.* (1995) identified that the third important factor affecting route choice is distance, which can be considered as a proxy variable of vehicle operating cost. One would expect that users will not choose a longer route unless the travel time on this route is expected to be shorter or the travel time reliability is higher. By replacing toll with distance, we can characterise each individual with (1) an indifference curve between distance and maximum time willing to spend for the desired travel time reliability; and (2) the desired minimum level of reliability.

We can further represent user groups by identifying a set of toll-time or distance-time indifference curves and their desired minimum level of reliability within each group represented by each curve. Each user class is identified by the combination of the two characteristics. We can then deduce the time budget surplus maximisation user equilibrium condition, whereby all individuals are travelling on the route with the highest time budget surplus value among all the efficient routes (based on the trade off between toll and time or distance and time) between each O-D pair.

Further research topics include consideration of elastic demand and other factors that might have strong influence on route choice behaviour. For example, users might not see the use of credits in emission trading schemes as a monetary cost. It is worthwhile to look into the possibility of modelling the effect of credits not as part of a multi-objective path cost function.

8 Acknowledgments

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