Abstract:
Traffic networks around the world, especially in densely populated areas, have experienced a growing rise in commuting time due to increasing levels of traffic. The increased congestion seen in many major cities represents a serious problem for roadway operators and commuters every day. Transportation authorities are looking for new traffic management techniques to better manage traffic and reduce congestion, delay and long queues in urban networks. To this end, a route guidance system that is a part of intelligent transportation systems would provide travelers with information on the best route to their destinations.

In past few years, many concepts had been applied for route guidance. For instance, User equilibrium (UE) model assumes that humans choose a route to minimize his / her travel time and this behavior on the individual level creates equilibrium at the network level. While user equilibrium should satisfy the drivers, it does not necessarily minimize the total travel time in the system, which is defined as the sum of all travelers' travel times. Thus, traffic assignment under UE assumption is not always the same as system optimal traffic assignment. Roughgarden and Tardos (2002) provide examples that show that the total travel time in UE can be arbitrarily large compared to that of the system optimum [1].

In another aspect, Traffic managers look for strategies that guide drivers in order to optimize total network travel times. In other words, they seek for system optimality. System optimality condition is occurring with an illustrative assumption. If this policy directly implemented, it could route some drivers on unacceptably long paths in order to use shorter paths for many other drivers [2]. In fact, the length of a route in the system optimum can be higher than in user equilibrium, even in the simplistic case of a single origin-destination pair [1]. By definition, the system optimality (SO) solution entails some long routes which may imply fewer common routes with the driver-preferred set [3]. If the routing instructions are not strictly followed by all travelers, the compliance is below 100%, reducing the effectiveness of the guidance system. So, the main concern of system optimality is the compliance rate of the network users.
In this research, a route guidance system was proposed in which the system administrator can guide the users by considering their route choice preference and system tolerance level. System tolerance can be defined as the difference between the total cost of the network in system optimality and any other conditions which can be encountered as a result of user route choice. Thus, a new parameter \( \gamma_{rs}^{k,sk} \) can be defined as below:

\[
\gamma_{rs}^{k,sk} = \frac{\sum_{a \in A_{rs}} c_a \times \lambda_{ars}^k}{C_{rs}^{k,sk}}; \forall r \in R, s \in S, k \in K_{rs}, k_{sk} \in SK_{rs}
\]

(1)

Where:

- \( A \): set of links
- \( \gamma_{rs}^{k,sk} \): The overlapping factor between the costs of the route (k) with the same route while network is in system optimality condition (sk) for O-D pair r-s,
- \( \tau_{rs}^k \): The link sets of the route k for O-D pair r-s,
- \( c_a \): The cost of link a, \( a \in A \),
- \( SK_{rs} \): Set of system optimality routes,
- \( K_{rs} \): Set of route between origin r (\( r \in R \)) and destination s (\( s \in S \)),
- \( C_{rs}^{k,sk} \): The cost of the route k in system optimality assignment condition (\( k_{sk} \in SK_{rs} \)) for O-D pair r-s,
- \( \lambda_{ars}^k \): is the link-route incidence matrix for O-D pair r-s.
- \( \lambda_{ars}^k = \begin{cases} 
1 & \text{if link a is on route k between O-D pair r-s} \\
0 & \text{Otherwise}
\end{cases} \)

When the overlapping factor, \( \gamma_{rs}^{k,sk} \), exceeds the System tolerance level, the route will be removed from the route guidance set and other choices will be introduced to users.

System tolerance level (\( \omega_r \)) is considered as a binary variable which defines the possibility of route guidance (\( \beta_{rs}^{k,sk} \)). In this case, at any time period, the overlapping factor of each route (\( \gamma_{rs}^{k,sk} \)) is being compared with System tolerance level. If the overlapping parameter between route (k) and system optimality route (sk) for O-D pair r-s is less than System tolerance level, the users will be guided through the route. The mathematical definition can be expressed as follows:

\[
\beta_{rs}^{k,sk} = \begin{cases} 
1 & \text{if } \gamma_{rs}^{k,sk} \leq \omega \text{ Where } \omega \geq 1 \\
0 & \text{Otherwise}
\end{cases} \]

(2)

Where:

- \( \beta_{rs}^{k,sk} \): The binary variable which is 1 if the overlapping parameter of the route (k) and route (sk) is less than system tolerance level and 0 otherwise
- \( \omega \): System Tolerance Level
In equation 2, when $\beta_{r_i}^{k, i_k}$ is equal to 1, the route guidance is possible. When $\Omega_o$ reaches 1, the route guidance system does not accept any routes except the system optimality ones and in complete obedience assumption of users, network condition reaches near to system optimality. By growing $\Omega_o$, the route guidance system loses his sensitivity and routes can be completely accessible and naturally they select the route with shortest perceived travel time.

In formulation level, we consider a model of reactive route guidance that allows us to work with static flows. As can be seen in Figure 1, formulation is developed in 2 stages. In the first stage, the model runs a system optimality assignment on the network to produce the comparing measures for route guidance. The second model performs the route guidance based on UE. A threshold has been set to make network close to system equilibrium. The threshold has been used to introduce new sets of routes to users in different time periods. User select a route among the introduced route set.

The developed model is tested for simple illustrative network. Results show that the proposed route guidance model has the most effects on total travel time of the network when system tolerance level changes from 1 to 1.4. When the tolerance parameter of the system reaches 1.1, the total travel time of the network is decreased more than 13 percent compare to UE condition. Furthermore, the results show that only a little share of users (15 percent of users in most case) in comparison with UE condition will change their routes.

Keywords: Deterministic route choice, Route Guidance, User equilibrium, System Optimality.

References:


Figure 1 - flowchart for proposed route guidance system

Start

Inputs Entry:
Nodes, Arcs, OD demands, Value of time, etc

Solve system optimality assignment for the network

Find $C_{rs}^{k,sk}$

Start incremental assignment

Define $\omega_o$

Find route $k$ with least cost for O-D pair $r-s$

Assign $q_{rs}$ to route $k$

Find the overlapping parameter

$\gamma_{rs}^{k,sk} = \frac{\sum r_i \times \lambda_{rs,k}}{C_{rs}} \forall r,s,k$

$\gamma_{rs}^{k,sk} \leq \omega_o$

Is $q_{rs}$ Assigned?

No

Stop Assigning to route $k$

Yes

End