

# Self-organizing Route Guidance Systems

Kosuke SEKIYAMA\* and Yasuhiro OHASHI\*

\*Department of Human and Artificial Intelligence Systems, University of Fukui

Email: {sekiyama, ohashi}@dis.his.fukui-u.ac.jp

**Abstract**—This paper proposes Self-organizing Route Guidance Systems which cooperates with self-organizing control of traffic signals. Self-organizing control of traffic signals provides the fully distributed approach to coordinate a number of signals distributed in a wide area according to the local information of the traffic flow so that split and offset between traffic signals are adjusted for the efficient traffic flow. The idea of Self-organizing Route Guidance Systems is to form the efficient route which facilitates the offset adjustment of self-organizing control of traffic signals by self-organizing multi-layered vector fields. Simulation demonstrates the effectiveness of proposed system under the traffic situation dynamically changes.

## I. INTRODUCTION

Recently, heavy traffic congestion at an urban area has brought several social problems such as increasing traffic accidents, environmental pollution, and economic losses. Intelligent Transportation Systems (ITS) is proposed for new transportation systems that is expected to solve these problems. To cope with the problem of traffic congestion, two major approaches are attempted in Japan.

The first approach is traffic flow control by the route guidance system using Vehicle Information and Communication System (VICS)[1][2]. VICS provides drivers with several information such as traveling time, traffic congestion, and route guidance through optical beacons or radio beacons located at roadside[3]. Route guidance system has ability to manage the amount of traffic at an urban area for to improve the traffic congestion problem. In order to provide the information on shortest route, the time-constrained  $A^*$  algorithm is proposed[4]. To minimize the expected travel time, dynamic programming techniques are applied in [5]. Also, to avoid the traffic congestion caused by the concentration of traffic flows, the routing method based on the estimation of the travel time is proposed in [6].

The second approach deals with an optimization of traffic signals in which parameters of traffic signals are adjusted to improve the performance of traffic flow [7][8]. We have proposed self-organizing control of traffic signal network[9]. It is fully distributed approach for improving the efficiency of the traffic flow according to dynamical changes of the environment. Traffic signals are represented by coupled nonlinear oscillators. Offsets between signals and splits are adjusted according to the local information of traffic flows so that the desired signals network offset patterns are self-organized through a mutual synchronization. Self-organizing control of traffic signals has the function for adjustment of the offset between traffic signals. Although, there are some limitations as we describes later. From the perspective of traffic signal control researche, traffic flow is treated as static or dynamic

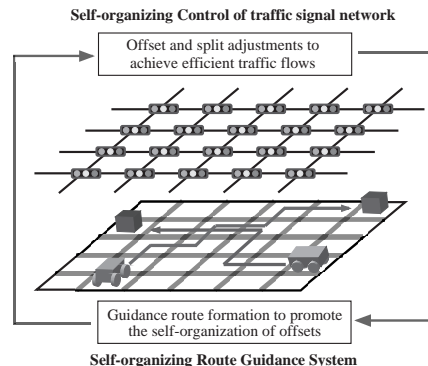


Fig. 1. Concept of Self-organizing Route Guidance Systems

environment that is uncontrollable. On the other hands, from the perspective of traffic flow control researche, traffic signal is not treated explicitly. In this paper, we propose the more efficient traffic system named Self-organizing Route Guidance Systems (SRGS) which cooperate with traffic signal control so that the problem of traffic congestion is improved. Since the traffic situation dynamically changes, it is difficult to manage a number of signals based on surveillances of the global information. In contrast, the self-organizing control of traffic signals provides the fully distributed approach to coordinate a number of signals distributed in a wide area according to the local information of the traffic flow so that the efficient traffic flow is realized. So, we adopt the self-organizing control of traffic signals in propose system. The main idea of SRGS that cooperates with self-organizing control of traffic signals generates group of vehicles as traffic flow which facilitates self-organization of offset pattern between signals network for improving the performance of urban traffic (Fig.1).

## II. EXTENSIONS OF SELF-ORGANIZING CONTROL OF TRAFFIC SIGNAL NETWORK[9]

### A. Urban Traffic Network Model

In this section, we outline the orverview of the theory of self-organizing control of traffic signal network. Despite the route guidance system needs offset adjustment in any route, self-organizing control of taffic signals deals with the traffic flow that pass thorough an intersection straight. Thus, we describe the extension of that theory so that the traffic flow turns at an intersection is dealt with.

A model of an urban traffic signal network used in self-organizing control of signals is described by a bi-directional graph network as depicted in Fig.2(a). Node  $S_i$ , ( $i = 1, \dots, N_S$ ) denotes the intersection and  $S_{ij}$ , ( $j \in 1, 2, \dots, n_i$ ) denotes the neighboring intersection of  $S_i$ .  $\varepsilon_{ij}$  is normalized

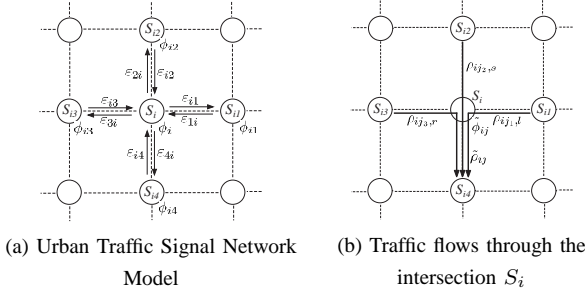


Fig. 2. Traffic inflows and outflows on signal network

amount of traffic flows defined by

$$\varepsilon_{ij} = \frac{1}{\rho_{im} T_i} \int_t^{t+T_i} \rho_{ij}(s) ds. \quad (1)$$

Where  $\rho_{im} = \text{const.}$  is traffic capacity and  $\rho_{ij}$  is traffic density in the road between the intersection  $S_i$  and  $S_{ij}$ .

### B. Modeling of Signal Network by Coupled Phase Oscillators

We describe the traffic signal network using a system of nonlinear oscillators with the nearest coupling. Specifically, we apply the phase description model of coupled oscillators. The following model for the oscillator,

$$\dot{\phi}_i(t) = \omega_i + \frac{K}{n_i} \sum_{j=1}^{n_i} \varepsilon_{ij} \sin(\phi_{ij} - \phi_i), \quad (2)$$

is used in the self-organizing control. By applying the Euler's formula, (2) is reformed as

$$\dot{\phi}_i(t) = \omega_i + \sigma_i K \sin(\bar{\phi}_i - \phi_i). \quad (3)$$

Where  $\bar{\phi}_i$  is the weighted phase average field depending on the amount of traffic flows from the neighboring signals  $S_{ij}$ .

### C. Self-organizing control of Offset Pattern

Self-organization of offset pattern on the signal network is realized through a mutual entrainment in (2). Let  $\Delta\phi_{ij}$  be the phase difference with respect to  $\phi_i$  defined by

$$\Delta\phi_{ij} = \phi_{ij} - \phi_i. \quad (4)$$

Let  $t_{ij}^*$  denote the expected time required for a vehicle to reach the signal  $S_i$  from  $S_{ij}$ , then the desired offset between the signals  $S_{ij}$  and  $S_i$  is given by

$$\Delta\phi_{ij}^* = 2\pi \frac{t_{ij}^*}{T_i} \pmod{2\pi} \quad (5)$$

where  $T_i$  is the cycle time of the signal  $S_i$ . The expected time  $t_{ij}^*$  is given by

$$t_{ij}^* = \frac{l_{ij}}{v_{ij}^{max}} \quad (6)$$

where,  $l_{ij}$  is the distance between the adjacent signal interval ( $S_{ij}, S_i$ ), and  $v_{ij}^{max}$  is the speed limit of the road. We adopt  $\Delta\phi_{ij}^*$  defined in (5) as a desired offset between the oscillators in the east-west direction where  $\phi_i = 0$  corresponds to the start of the green period. However, we need to define the desired relative phase between oscillators in the north-south direction where  $\phi_i = 0$  corresponds to the start of the red period as

$$\Delta\phi_{ij}^* = 2\pi \frac{t_{ij}^* + (T_{i-1,1} - T_{i,1})}{T_i} \pmod{2\pi}, \quad (7)$$

since the offset is defined as the time difference in the start of the green period. In this light, the desired natural frequency of the oscillator  $S_i$  is given by

$$\omega_i^* = \Omega_i - \bar{\sigma}_i^* K \sin \Delta\bar{\phi}_i^*. \quad (8)$$

In order to achieve the desired relative phase,

$$\Delta\bar{\phi}_i^* = \arctan \left( \frac{\sum_{j=1}^{n_i} \varepsilon_{ij} \sin \Delta\phi_{ij}^*}{\sum_{j=1}^{n_i} \varepsilon_{ij} \cos \Delta\phi_{ij}^*} \right), \quad (9)$$

with respect to the weighted phase average field,  $\bar{\phi}_i$ , through mutual entrainment. Furthermore, we consider approximation of the natural frequency of the weighted phase average field  $\bar{\phi}_i$ . The natural frequency of the weighted phase average field is approximated as

$$\Omega_i \cong \frac{1}{1 + \frac{1}{n_i} \sum_{j=1}^{n_i} \frac{\varepsilon_{ij}}{\varepsilon_{ji}}} \left( \omega_i + \frac{1}{n_i} \sum_{j=1}^{n_i} \frac{\varepsilon_{ij}}{\varepsilon_{ji}} (\omega_{ij}) \right). \quad (10)$$

The desired natural frequency  $\omega_i^*$  in (8) is update by

$$\omega_i(\tau_i + 1) = \omega_i(\tau_i) + \alpha(\omega_i^* - \omega_i(\tau_i)) + \beta(\tilde{\omega}_i^* - \omega_i(\tau_i)). \quad (11)$$

Where  $\tau_i$  is a scaled time associated with the period of the signal, and  $\alpha, \beta \in (0, 1)$  is a positive constant. In (11), a base frequency  $\tilde{\omega}_i^*$  for each signal is introduced to prevent the drift of the natural frequency of all the signals.

### D. Split Adjustments

Consider the split settings which are adjusted according to the amount of incoming traffic. Now, define the cycle time of the signal  $S_i$  as

$$T_i(\tau_i) = T_{i1}(\tau_i) + T_{i2}(\tau_i) + 2T_{cl}, \quad (12)$$

where  $T_{cl} = 5(\text{sec})$  is clearance time, and  $T_{i1}(\tau_i)$  and  $T_{i2}(\tau_i)$  are the split time of the signal  $S_i$  in the east-west and north-south direction respectively. The split time  $T_{ik}$  is updated at the beginning of each signal cycle in proportion to the sum of the normalized amount of incoming traffic of the east-west and north-south direction.

$$r_{i1} = \sum_{k=1}^2 \varepsilon_{i,2k-1}, \quad r_{i2} = \sum_{k=1}^2 \varepsilon_{i,2k}. \quad (13)$$

as

$$T_{ik}^* = \frac{r_{ik}(\tau_i)}{r_{i1}(\tau_i) + r_{i2}(\tau_i)} (T_i(\tau_i) - 2T_{cl}), \quad (14)$$

$$T_{ik}(\tau_i + 1) = T_{ik}(\tau_i) + \gamma(T_{ik}^* - T_{ik}(\tau_i)), \quad (k = 1, 2). \quad (15)$$

Where  $\gamma$  is an updating constat.

### E. Offset Adjustment Considering Left-Right Turns

Our previous work of the self-organizing control of signal network did not deal with left-right turns explicitly[9]. In this paper, we incorporate the effect of left-right turns into the previous model. In the previous method, the offset pattern is effectivly self-organized in the case of traffic situation is depicted in Fig.3(a). However, when the traffic situation is as shown in Fig.3(b) where the traffic flow turns left or right at the intersection  $S_i$ , the effective offset pattern is not available. The synchronization between the phase of north-south direction of  $S_{ij}$  and the *reversal* phase of north-south direction of  $S_i$ , which is equivalent to the phase of east-west direction, should

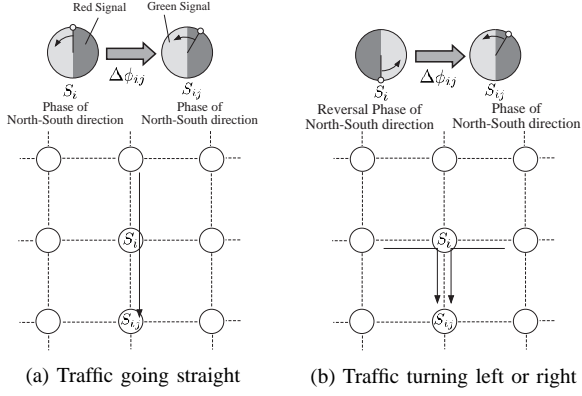


Fig. 3. Offset adjustment between  $S_i$  and  $S_{ij}$

be realized for offset adjustment in the case of Fig.3. To select the synchronization pattern of Fig.3(a) or Fig.3(b), we use the traffic outflow from  $S_i$ . As shown in Fig.2(b), the amount of traffic outflow which turns left through the intersection  $S_i$  is denoted by  $\rho_{ij1,l}(t)$ , the amount of traffic outflow which goes straight denoted by  $\rho_{ij2,s}(t)$  and the amount of traffic outflow which turns right denoted by  $\rho_{ij3,r}(t)$ . Total amount of traffic outflow  $\tilde{\rho}_{ij}(t)$  is given by

$$\tilde{\rho}_{ij}(t) = \rho_{ij1,l}(t) + \rho_{ij2,s}(t) + \rho_{ij3,r}(t). \quad (16)$$

In order to adjust offset between the phase of north-south direction of  $S_{ij}$  and the reversal phase of north-south direction of  $S_i$ , we define the phase considering left-right turns of  $S_i$  denoted by  $\tilde{\phi}_{ij}$ . Turn rate  $\kappa_{ij}$  on  $S_i$  is defined by

$$\kappa_{ij} = \frac{\rho_{ij1,l} + \rho_{ij3,r}}{\tilde{\rho}_{ij}}. \quad (17)$$

Hence, the modified phase considering left-right turns is defined by

$$\tilde{\phi}_{ij} = \phi_i + \xi_{ij} \frac{2\pi T_{i,2}}{T_i}. \quad (18)$$

Where  $\mu, c$  are constants. Also turn sensitivity  $\xi_{ij}$  is defined by

$$\xi_{ij} = \frac{1 + \tanh(\mu \kappa_{ij} - c)}{2}. \quad (19)$$

(19) is introduced to select the synchronization pattern of Fig.3(a) or Fig.3(b).

### III. ARCHITECTURE OF SRGS

#### A. Even Two-way Traffic Problem of Traffic Signal Network

The offset is not adjusted in the certain traffic situation, called even two-way traffic, where the difference between the amounts of traffic flow on the east-west and north-south directions that face toward each other is little (Fig.4). To cope with this problem, we consider the route guidance system which forms the route avoiding the even two-way traffic flow such that effective offset patterns are self-organized.

#### B. Configurations of SRGS

The SRGS is composed of vehicle guidance units depicted in Fig.5. Each intersection  $S_i$  at the guidance area has a vehicle guidance unit  $U_i$ , which has a reward variable corresponding to the destination  $h$ , ( $h = 1, 2, \dots, N_h$ ) denoted by  $R_i^h$ . The

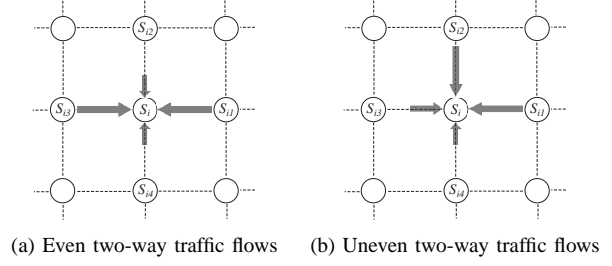


Fig. 4. Traffic flow patterns on signal network

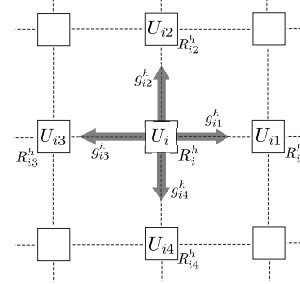


Fig. 5. Reward and guidance vector of  $U_i$  on layer  $h$

SRGS has a multi-layered guidance vector fields corresponding to each destination for the route guidance (Fig.6). The guidance unit decides the route guidance corresponding to the destination of the vehicle based on the guidance vector. The guidance vector of the guidance unit  $U_i$  corresponding to the destination  $h$  denoted by

$$\mathbf{g}_i^h = \{g_{i1}^h, g_{i2}^h, g_{i3}^h, g_{i4}^h\}. \quad (20)$$

Where  $g_{ij}^h \geq 0$ ,  $\sum_{j=1}^4 g_{ij}^h = 1$ . In SRGS, vehicles communicate with a guidance unit at every intersection. Every guidance unit has the local route guidance information for each destination. Vehicle transmits the own destination toward the guidance unit. The guidance unit transmits the route guidance information to the vehicle. Units can observe the local information and self-organize a route corresponding to each destination through a local interaction. The vehicle send its own destination to the guidance unit when it enter the intersection. The guidance unit corresponding to destination receives reward value according to frequency of transmission. Reward value is difused over the guidance unit through the local interaction of guidance units. Guidance vector field corresponding to each destination is formed by the reward field spreading on the urban. The route for the destination which many vehicles select is reinforced. We adopt the distributed and self-organizing strategy because distributed and self-organizing strategy are effective under the dynamical traffic environment.

### IV. SELF-ORGANIZING METHOD OF MULTI-LAYERED GUIDANCE VECTOR FIELDS

#### A. Evaluation of Traffic Flows

In this section, we propose the self-organizing method of multi-layered route guidance vector field (Fig.6, 5). The main theme of SRGS is how does SRGS generate the traffic flow that facilitates a self-organization of an offset pattern of traffic signal network. On the other hand, traffic congestion occur when large amount of vehicles flow into the same

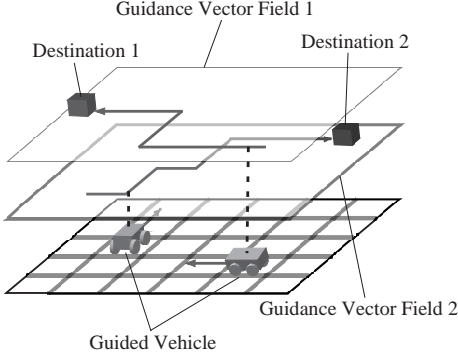


Fig. 6. Multi-layered vector fields of SRGS

intersection. Therefore, we consider the method to facilitate offset adjustment of traffic signals under avoiding traffic jam raises. Multi-layered guidance vector field is formed by reward difference between intersections. We describe the evaluation of traffic, the method of spreading the reward field and deciding guidance vectors in following.

Consider evaluation of traffic flow pattern which inflow to the intersection  $S_i$ . Self-organizing control of signal network is able to adjust the offset between signals on the condition that uneven two-way traffic. Although, the offset adjustment does not function on the condition that even two-way traffic. Therefore, uneven two-way traffic is needed to facilitate the offset adjustment.

In order to evaluate the traffic pattern of  $S_i$ , let  $E_{i,1}, E_{i,2}$  denote unevenness of east-west and north-south two-way traffic define by as follows,

$$E_{i,1}(\tau_i) = \frac{1}{2} \left| \frac{\varepsilon_{i1}(\tau_i) - \varepsilon_{i3}(\tau_i)}{\varepsilon_{i1}(\tau_i) + \varepsilon_{i3}(\tau_i)} \right|, \quad (21a)$$

$$E_{i,2}(\tau_i) = \frac{1}{2} \left| \frac{\varepsilon_{i2}(\tau_i) - \varepsilon_{i4}(\tau_i)}{\varepsilon_{i2}(\tau_i) + \varepsilon_{i4}(\tau_i)} \right|. \quad (21b)$$

Where,  $\tau_i$  is discrete time based on the period of  $S_i$ . The higher value of  $E_{i,1}$  and  $E_{i,2}$  imply that the unevenness of  $S_i$  is large. Large amount of traffic flow cause the traffic congestion even though the traffic flow pattern is facilitating the offset adjustment. Thus, we consider the congestion degree denoted by  $F_i$ .  $F_i$  is defined by

$$F_i(\tau_i) = \prod_{j=1}^{n_i} (1.0 - \varepsilon_{ij}(\tau_i)). \quad (22)$$

Note that the traffic congestion degree  $F_i$  will go to 0 when the traffic congestion occurs. Also, we define the traffic evaluation value  $E_i$  based on  $E_{i,1}$ ,  $E_{i,2}$  and  $F_i$ , as follows.

$$E_i(\tau_i) = F_i(\tau_i) (E_{i,1}(\tau_i) + E_{i,2}(\tau_i)). \quad (23)$$

Weighted evaluation  $\bar{E}_i$  of recent time steps  $T$  is given by

$$\bar{E}_i(\tau_i) = \sum_{n=0}^L \lambda^{n-L} E_i(\tau_i - n), \quad (24)$$

where  $\lambda \in (0, 1)$  is decay rate.  $\bar{E}_i$  will take a large value in the condition that the traffic flow pattern facilitates an effective offset adjustment for the traffic flow.

## B. Update Law of Reward Variable and Guidance Vector

The reward  $R_i^h$  is decided by the evaluation of traffic and a neighboring reward  $R_{ij}$ . The route between  $S_i$  and  $S_{ij}$  is reinforced when traffic flow from  $S_i$  to  $S_{ij}$ . Now, normalized traffic density  $\varepsilon_{ij}$  is divided into the traffic density of every layer according to the destination  $h$ , which is denoted by  $\varepsilon_{ij}^h$ . Where the relation between  $\varepsilon_{ij}$  and  $\varepsilon_{ij}^h$  holds as follows,

$$\varepsilon_{ij} = \sum_{k=1}^{N_h} \varepsilon_{ij}^k. \quad (25)$$

The in-out relation of traffic flow between  $S_i$  and  $S_{ij}$  is defined by

$$e_{ij}^h = \frac{\varepsilon_{ji}^h + (1.0 - \varepsilon_{ij}^h)}{2}, \quad (0 \leq e_{ij}^h \leq 1). \quad (26)$$

When the amount of traffic flow from  $S_i$  to  $S_{ij}$  is larger and the amount of traffic flow from  $S_{ij}$  to  $S_i$  is smaller, a in-out relation  $e_{ij}^h$  approaches to 1. The reward  $R_i^h$  is updated with (26) and (24) as follows,

$$R_i^h(\tau_i + 1) = \bar{E}_i(\tau_i) \left( R_i^h(\tau_i) + \frac{\eta}{n_i} \sum_{j=1}^{n_i} e_{ij}^h \Delta R_{ij}^h(\tau_i) \right). \quad (27)$$

Where  $\eta \in (0, 1)$  is updating constant,  $\Delta R_{ij}^h$  is the difference of a reward  $R_i^h$  and a neighboring reward  $R_{ij}^h$  is given by

$$\Delta R_{ij}^h(\tau_i) = R_{ij}^h(\tau_i) - R_i^h(\tau_i). \quad (28)$$

By this updating law, a reward spread to intersections having high evaluation value of traffic. Then, the guidance vector is updated by following law.

$$g_{ij}^h(\tau_i + 1) = \frac{g_{ij}^h(\tau_i) + \theta \Delta R_{ij}^h(\tau_i)}{\sum_{j=1}^{n_i} (g_{ij}^h(\tau_i) + \theta \Delta R_{ij}^h(\tau_i))}. \quad (29)$$

Where,  $\theta \in (0, 1)$  is an updating constant.

## C. Route Selection Method

The route guidance information is transmitted to the vehicle is selected by a probability  $p_{ij}^h$  corresponding to the destination  $h$ . Let  $u_j$  be the coefficient to restrict a U-turn defined by

$$u = \{u_1, u_2, u_3, u_4\} = \begin{cases} \{0, 1, 1, 1\} & \text{if vehicle's position is east} \\ \{1, 0, 1, 1\} & \text{if vehicle's position is north} \\ \{1, 1, 0, 1\} & \text{if vehicle's position is west} \\ \{1, 1, 1, 0\} & \text{if vehicle's position is south} \end{cases}. \quad (30)$$

Then, the route selection probability  $p_{ij}^h$  is defined with  $u_j$  as follows,

$$p_{ij}^h = \frac{u_j \exp(Dg_{ij}^h)}{\sum_{j=1}^{n_i} u_j \exp(Dg_{ij}^h)}. \quad (31)$$

Where  $D$  is a constatnt.

TABLE I  
SIMULATION PARAMETERS

$v_{max}$	maximum velocity	50 km/h(13.89 m/s)
$a_{max}^+$	acceleration	1.5 m/s <sup>2</sup>
$a_{max}^-$	deceleration	-3.0 m/s <sup>2</sup>
$K$	coupling constant	0.1
$\alpha, \beta$	updating rate of $\omega$	0.5, 0.05
$\tilde{\omega}_i^*$	base natural frequency	0.10147
$\rho_{im}$	road capacity	80 vehicles
$\mu, c$	turn sensitivity constant	30, 15
$\lambda$	decay rate of $E$	0.8
$\theta$	updating rate of $g$	0.05
$D$	decision constant	25

## V. SIMULATIONS

### A. Vehicle Model

We use the vehicle model and the road model depicted in Fig.7. The road has a two-lane. One is a straight line, and the other is a right-turn lane. In the vehicle model, the desired velocity of the vehicle  $m$  is determined by

$$v_m^* = \begin{cases} v_{max} & \text{if } x_m > \frac{(v_{max})^2}{2|a_{max}^-|} \\ \sqrt{2|a_{max}^-|x_m} & \text{if } x_m \leq \frac{(v_{max})^2}{2|a_{max}^-|}, \end{cases} \quad (32)$$

according to the distance between the preceding vehicle  $m + 1$ , and each vehicle follows this desired velocity with the acceleration/deceleration of  $a_{max}^\pm$ , where  $v_{max}$  is the speed limit.

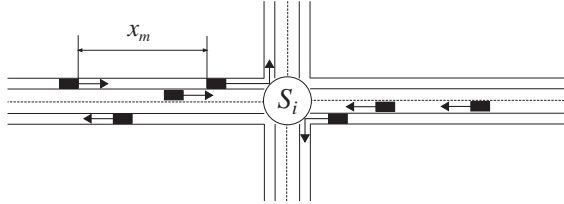


Fig. 7. Traffic flow model

### B. Comparative Method

Now, to observe the effectiveness of the proposed system, we introduce a simple route selection method. The route selection probability  $\tilde{p}_{ij}^h$  is decided by a normalized direction vector  $d_{ij}^h$  from the intersection  $S_i$  to the destination  $h$ .

$$d_i^h = \{d_{i1}^h, d_{i2}^h, d_{i3}^h, d_{i4}^h\}, \quad (33)$$

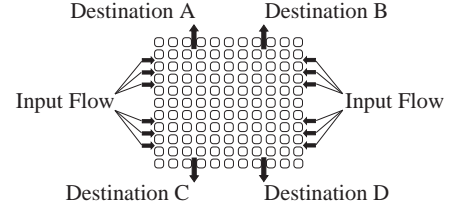
where  $d_{ij}^h \geq 0$ ,  $\sum_{j=1}^4 d_{ij}^h = 1$ . The probability  $\tilde{p}_{ij}^h$  is defined with  $d_{ij}^h$  and  $u_j$  in (30) as follows,

$$\tilde{p}_{ij}^h = \frac{u_j d_{ij}^h}{\sum_{j=1}^4 u_j d_{ij}^h}. \quad (34)$$

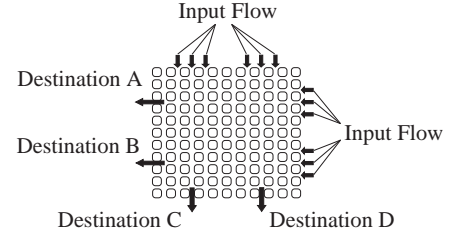
By using this simple route selection method, a vehicle avoids the roundabout route, and the traffic flow is dispersed in a measure. We call this method basic direction route guidance.

TABLE II  
SIMULATION ENVIRONMENT

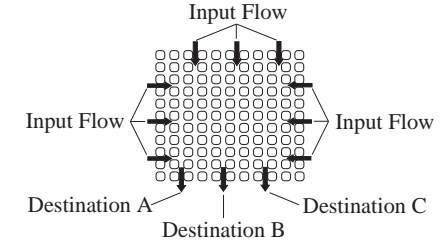
$t = 0 \sim 20000$	Traffic situation 1 (Fig.8(a))
$t = 20000 \sim 40000$	Traffic situation 2 (Fig.8(b))
$t = 40000 \sim 60000$	Traffic situation 3 (Fig.8(c))



(a) Traffic situation 1



(b) Traffic situation 2



(c) Traffic situation 3

Fig. 8. Traffic situations used in simulations

### C. Simulation settings

Consider the traffic signal network with the  $10 \times 10$  grid. We examine the proposed system in several traffic situations. Vehicles flow into the signal network from the specified intersections called inflow intersection. Then, each vehicle chooses the destination randomly with the same probability. Following two cases are examined.

- self-organizing route guidance
- basic direction route guidance

The amount of inflow is set to 5 vehicles/min on each inflow intersection. The initial natural frequency (rad/s) and phase (rad) of each signal is given randomly as  $0.08 \pm 0.02$  rad/s and  $0 \sim 2\pi$  rad respectively. The initial route guidance vectors are  $g_i^h = \{0.25, 0.25, 0.25, 0.25\}$ . The distance between signals is 200 meters, and the other parameters are listed in Table.I

### D. Simulation Results

Consider the traffic situation as shown in Fig.8(a) - 8(c). We simulate the dynamical environment where the traffic situation changes according to Table.II.



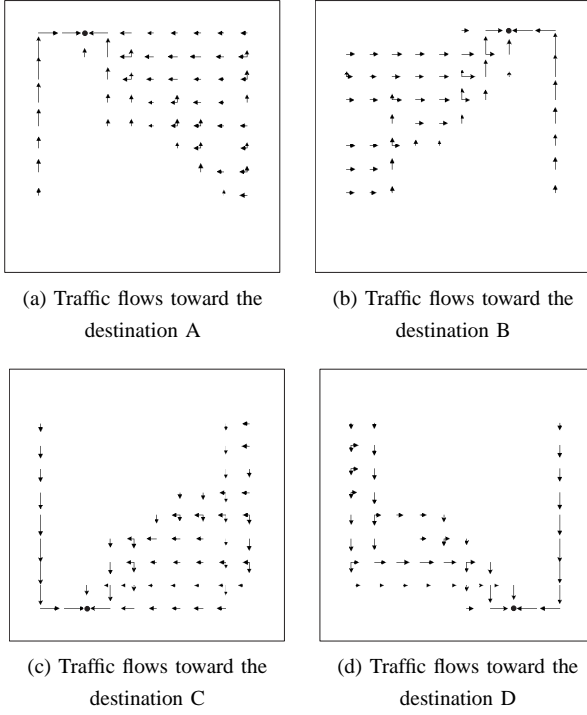


Fig. 10. Average traffic flows toward each destination during  $t = 0 \sim 20000$

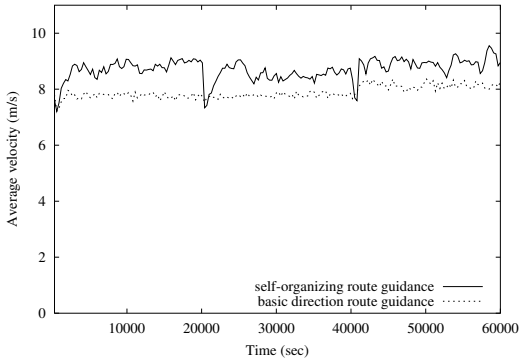


Fig. 9. The average velocity of all vehicles. The traffic simulation is changed at  $t = 20000$  and  $t = 40000$ .

The simulation result, the average velocity of all vehicles, is shown in Fig.9. The solid line shows the result of the case where self-organizing route guidance is used, and the dotted line shows the result of the case where basic direction route guidance is used. In the case of self-organizing route guidance is used, the average velocity of all vehicles during  $t = 0 \sim 20000$  is  $8.72 \text{ m/s} = 31.4 \text{ km/h}$ . In contrast, when we use basic direction route guidance, the average velocity of all vehicles during  $t = 0 \sim 20000$  is  $7.75 \text{ m/s} = 27.9 \text{ km/h}$ . The proposed system achieves the effective traffic flows. The average traffic flow corresponding to each destination during  $t = 0 \sim 20000$  is as shown in Fig.10. Guidance vector fields avoiding even two-way traffic flow patterns are self-organized by proposed method even though the even two-way traffic flow pattern occurs frequently in this traffic situation.

Then, we discuss the traffic situation dynamically changes. In this simulation, the traffic situation is changed at  $t = 20000$  and  $t = 40000$ . Though the average velocity comes down at  $t = 20000$  and  $t = 40000$ , the proposed system adapts to changes of the traffic situation immediately. In the case of self-organizing route guidance is used, the average velocity of all vehicles during  $t = 20000 \sim 40000$  is  $8.50 \text{ m/s} = 30.6 \text{ km/h}$ . On the other hand, the average velocity of all vehicles during  $t = 20000 \sim 40000$  is  $7.77 \text{ m/s} = 28.0 \text{ km/h}$  when we use basic direction route guidance. By the same, the average velocity during  $t = 40000 \sim 60000$  is  $8.91 \text{ m/s} = 32.1 \text{ km/h}$  when self-organizing route guidance is used, and the average velocity during  $t = 40000 \sim 60000$  is  $8.13 \text{ m/s} = 29.1 \text{ km/h}$  when basic direction route guidance is used. The simulation result illustrates effectiveness of the proposed system under the dynamic traffic situation.

## VI. CONCLUSIONS

In this paper, we proposed Self-organizing Route Guidance Systems (SRGS), which integrates the self-organizing control of signal network and the route guidance systems. The route guidance system navigates the traffic flow with the same destination to a specific path so that the self-organization of signal network functions more efficiently. In the simulation, the guidance vector fields, which avoid the traffic congestion and the traffic flow pattern preventing the offset adjustment, are generated. The simulation results show the effectiveness of this unified approach in the dynamical traffic situation. In general, since each vehicle has its own destination, some technique should be introduced to put together similar destinations into roughly defined destination. This is a future work.

## REFERENCES

- [1] S. Tsugawa, M. Aoki, A. Hosaka, and K. Seki. A survey of present ivhs activities in japan. *Control Engineering Practice*, 5(11):1591–1597, Nov. 1997.
- [2] H. Kojima. Radio communication technology on vics and next emerging systems in its. *IEEE Conference on Intelligent Transportation System*, pages 26–32, Nov. 1997.
- [3] M. Yamaguchi, T. Kitamura, S. Jinno, and T. Tajima. The interactive cdrg using infrared beacons. *IEEE International Conference on Intelligent Transportation Systems*, pages 278–283, Oct. 1999.
- [4] H. Hiraishi, H. Ohwada, and F. Mizoguchi. Time-constrained heuristic search for practical route finding. *5th Pacific Rim International Conference on Artificial Intelligence*, pages 389–398, Nov. 1998.
- [5] L. Fu. An adaptive routing algorithm for in-vehicle route guidance systems with real-time information. *Transportation Research Part B*, 35(8):749–765, Sep. 2001.
- [6] F.P. Defflorio. Evaluation of a reactive dynamic route guidance strategy. *Transportation Research Part C*, 11(5):375–388, Jul. 2003.
- [7] I. Porche, M. Sampath, R. Sengupta, Y.-L. Chen, and S. Lafortune. A decentralized scheme for real-time optimization of traffic signals. *IEEE International Conference on Control Applications*, pages 582–589, Sept. 1996.
- [8] T. Misawa, H. Kimura, S. Hirose, and N. Osato. Multiagent-based traffic signal control with reinforcement learning. *IEICE Transactions*, J83-D-I(5):478–486, May 2000.
- [9] K. Sekiyama, J. Nakanishi, I. Takagawa, T. Higashi, and T. Fukuda. Self-organizing control of urban traffic signal network. *IEEE International Conference on Systems, Man, and Cybernetics*, pages 2481–2486, Oct. 2001.