

Real-Time route guidance in stochastic networks

Hai L. Vu

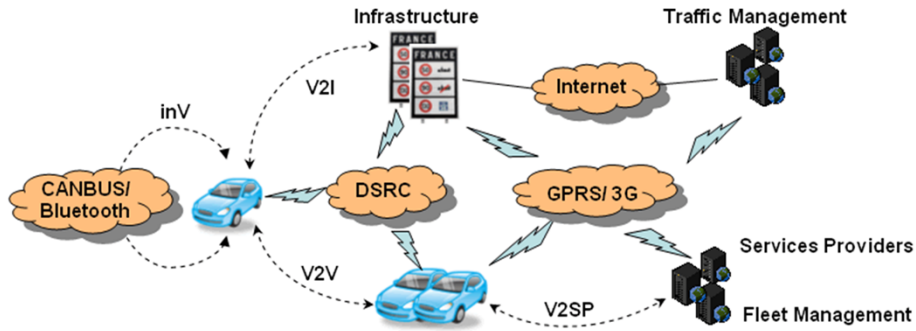
Wei Dong, Bao Vo, Yoni Nazarathy and Serge Hoogendoorn



Intelligent Transport Systems (ITS)



ITS Architecture



DSRC: Dedicated short range communications



SWINBURNE
UNIVERSITY OF
TECHNOLOGY

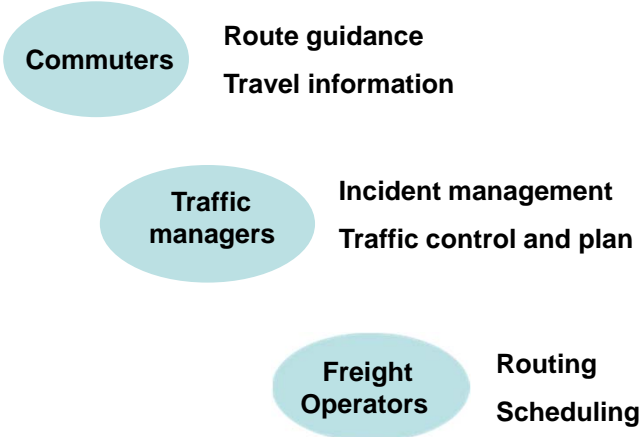
Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/>

hvu@swin.edu.au

Page 3

ITS Apps



SWINBURNE
UNIVERSITY OF
TECHNOLOGY

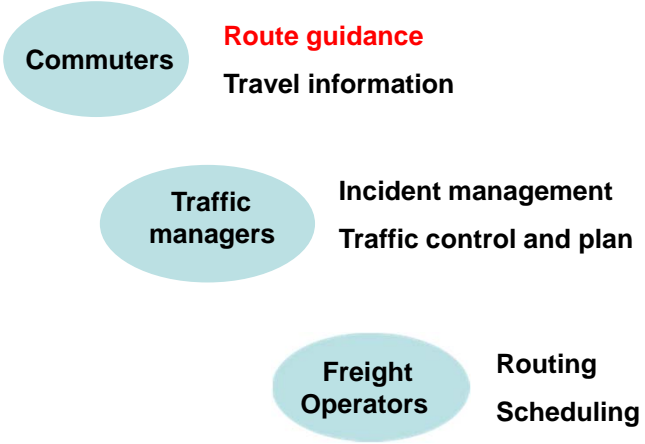
Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/>

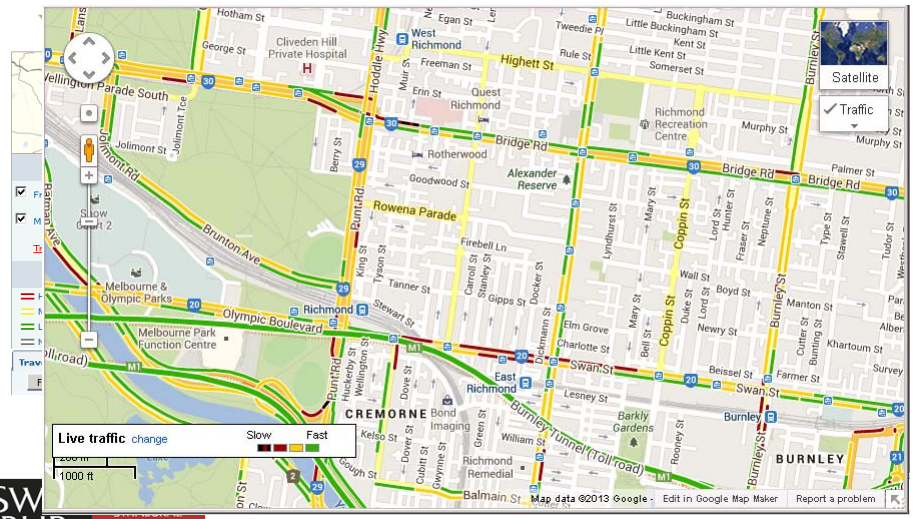
hvu@swin.edu.au

Page 4

ITS Apps



Example: Almost real time info



Outline



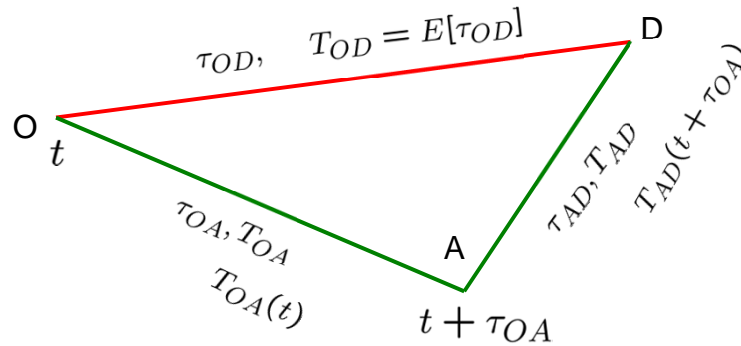
- Shortest path: Real-time application
- Framework for shortest path in stochastic time-dependent networks
- Numerical results
- Conclusion

Outline



- Shortest path: Real-time application
- Framework for shortest path in stochastic time-dependent networks
- Numerical results
- Conclusion

Shortest Path: Real time application



Shortest path (least expected travel time): $T_{OA} + T_{AD} < T_{OD}$

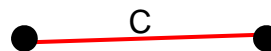
Congestion level (network state)



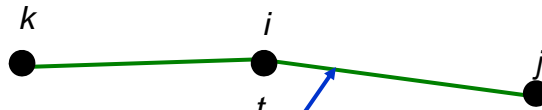
Time-dependent network state: $T_{OA}(t)$ $T_{AD}(t + \tau_{OA})$

Link is congested if $DS > 0.95$ and $V_k/V_o > 2.4$

V_o : # cars passed the detector (V_o), DS : degree of saturation value,
 V_k : estimated traffic count having the same DS under free flow conditions

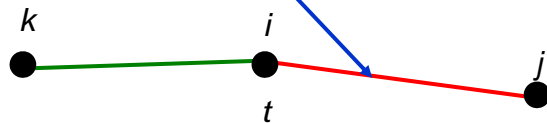


Spatial dependency



Real Time
LET

$$u_{ki}(t) = \min_{j \neq i} [\alpha_{kij}(U_{ij}(t) + u_{ij}(t + U_{ij}(t))) + (1 - \alpha_{kij})(C_{ij}(t) + c_{ij}(t + C_{ij}(t)))]$$



SWINBURNE
UNIVERSITY OF
TECHNOLOGY

LET: Least Expected Travel time

Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/> hvu@swin.edu.au

Page 11

Outline



- Shortest path: Real-time application
- Framework for shortest path in stochastic time-dependent networks
- Numerical results
- Conclusion



SWINBURNE
UNIVERSITY OF
TECHNOLOGY

Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/> hvu@swin.edu.au

Page 12

The Framework



Bellman's optimality equation:

$$u_{ki}^s(t) = \min_{j \neq i} \left[\sum_{r=1}^M p_{kij}^{sr} (U_{ij}^r(t) + u_{ij}^r(t + U_{ij}^r(t))) \right]$$

$$u_{kD}^s(t) = 0 \quad \bullet \quad u_{ki}^s(t): \text{ LET between node } i \text{ and the destination node } D \text{ at time } t \text{ if the incoming link } (k, i) \text{ is in state } s$$

time dependent

$U_{ij}^r(t)$: expected travel time between nodes i, j at time $t \in [\tau_h, \tau_{h+1})$ under link state r

time independent

p_{kij}^{sr} : the prob. that link (i, j) is in state r if the incoming link (k, i) is in state s



SWINBURNE
UNIVERSITY OF
TECHNOLOGY

Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/>

hvu@swin.edu.au

Page 13

Framework Cont.



$$u_{ki}^s(t) = \min_{j \neq i} \left[\sum_{r=1}^M p_{kij}^{sr} (U_{ij}^r(t) + u_{ij}^r(t + U_{ij}^r(t))) \right]$$

$$p_{kij}^{sr}(t) = \int_{t_{h(i,j)}^{r-1}}^{t_{h(i,j)}^r} P_{ij}(t, \xi) d\xi \approx p_{kij}^{sr} \quad \sum_{r=1}^M p_{kij}^{sr} = 1$$

$P_{ij}(t, \xi) d\xi$ is the prob. traveling from node i to j requires time between ξ and $\xi + d\xi$,
 $t \in [\tau_h, \tau_{h+1})$ given (k, i) is in state s

$t_{h(i,j)}^r$ is the travel time threshold for link (i, j) to be in state r



SWINBURNE
UNIVERSITY OF
TECHNOLOGY

Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/>

hvu@swin.edu.au

Page 14

Temporal Dependency



$$U_{ij}^r(t) = \tau_{h+\eta} - t + \frac{l_{ij} - l_{ij}^{\eta-1}}{v_{h+\eta}^r(i,j)} \quad l_{ij} \text{ length of link } (i,j)$$

$\eta \geq 1$ number of time zones a car crosses while traveling link (i,j)

$$l_{ij}^{\eta-1} = l_{ij}^{\eta-2} + v_{h+\eta-1}^r(i,j) (\tau_{h+\eta} - \tau_{h+\eta-1})$$

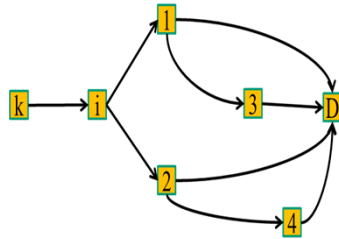
$$l_{ij}^{-1} = 0$$

Outline



- Shortest path: Real-time application
- Framework for shortest path in stochastic time-dependent networks
- Numerical results
- Conclusion

Simple Network



Time interval	Link	w1 (0.5)	w2 (0.3)	w3 (0.2)
0	(i, 1)	4	4	5
	(i, 2)	5	5	6
	(1, D)	5	5	4
	(1, 3)	4	4	3
	(3, D)	3	3	5
	(2, D)	4	4	5
	(2, 4)	4	4	5
	(4, D)	4	4	5
1	(i, 1)	15	15	16
	(i, 2)	14	16	14
	(1, D)	15	15	17
	(1, 3)	16	15	15
	(3, D)	15	16	14
	(2, D)	14	16	16
	(2, 4)	15	16	16
	(4, D)	14	16	15

Results 1



	Route	Actual travel time (w_2)	Percentage difference
Method 1	Node i-1-D	19	0
Method 2	Node i-1-D	19	0
Method 3	Node i-2-D	21	0.11

Method 1 utilizes both temporal and spatial dependences (our framework)

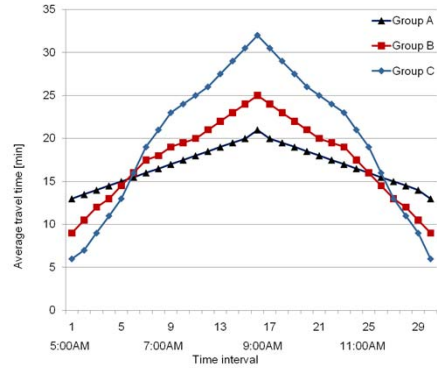
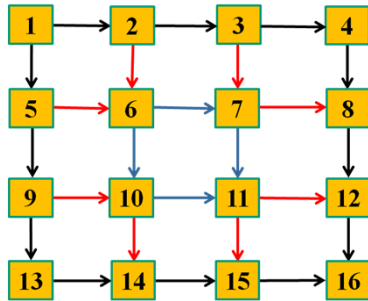
Method 2 uses full travel time distribution [1]

Method 3 utilizes spatial travel time correlation only and is time-independent [2]

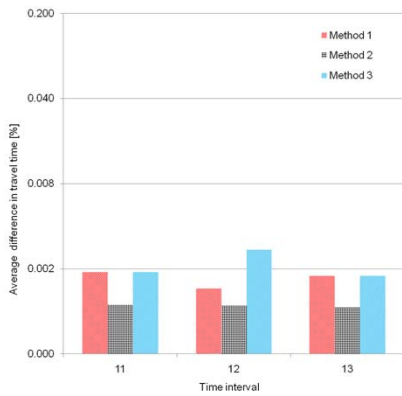
[1] Gao, S. and I. Chabini, "Optimal routing policy problems in stochastic time-dependent networks," Transportation Research Part B: Methodological, Vol. 40, No. 2, 2006, pp. 93-122.

[2] Fan, Y. Y., R. E. Kalaba, and J. E., "Moore, Shortest paths in stochastic networks with correlated link costs," Computers and Mathematics with Applications, Vol. 49, No. 9-10, 2005, pp.1549-1564.

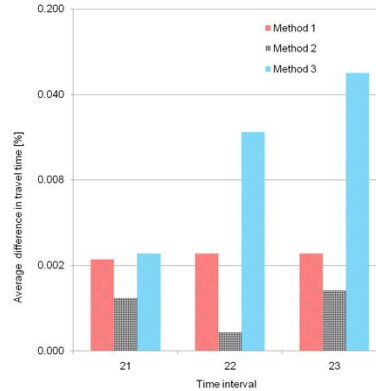
Grid Network



Results 2



Average difference in travel time using different methods (middle 10 time intervals)



Average difference in travel time using different methods (last 10 time intervals)



Outline



- Shortest path: Real-time application
- Framework for shortest path in stochastic time-dependent networks
- Numerical results
- Conclusion

Conclusion



- The proposed framework can achieve similar accuracy with a much smaller set of parameters compared to the case when the full joint distribution of network travel times is required.
- Both temporal and spatial correlations are handled while the framework remains simple.
- Improvements in optimal route choice decisions are shown via examples.



THANK YOU

SWIN
BUR
NE

SWINBURNE
UNIVERSITY OF
TECHNOLOGY

Swinburne Intelligent Transport Systems Lab

<http://caia.swin.edu.au/cv/hvu/> hvu@swin.edu.au

Page 23