

#### A Game-Theoretic Method to Efficiently Assess the Vulnerability of a Dynamic Transportation Network

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# Outline

Motivation

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- Game Theoretic Vulnerability Assessment
- Deterministic Vulnerability Assessment
- Example scenario

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- Conclusion
- Future Work

### Motivation

• Continuing increase in city populations

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- Expect criticality of transportation infrastructure to increase
- Disaster planning, response, and recovery decision support systems
  - Often assume transportation network completely available
  - Unrealistic assumption may lead to suboptimal strategy



### Static Traffic Assignment

- Previous transportation network vulnerability research performed in context of static traffic models
- Simplifies assumptions
  - Travel times of each link on route added together to compute travel time
  - Inflow and Outflow of link equal
  - Congestion occurs if Volume-to-Capacity ratio (V/ C) > 1.0



## Dynamic Traffic Assignment

- Travel demand function of time
- Explicit modeling of traffic flow dynamics
  - Ensures direct link between travel time and congestion
- Application of dynamic transportation models
  - Congestion and vulnerability assessment



#### Framework



# **Deterministic Method**

- For every edge *e* in network *G*
- For every time interval  $\Delta t_i$  in  $\Delta T$ 
  - Disable edge *e* during interval  $\Delta t_i$
  - Record travel times of vehicles
  - Calculate ratio of disrupted travel time with undisrupted travel time
- **Problem**: Method is not scalable

### Game Theory

- The analysis of competitive situations (or situations of conflicts) using mathematical models
- Involves one or more **players**

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- Actions taken by players called **moves**
- A set of **outcomes** for each move
- An amount received for each outcome called payoff

# Approach

- Two-player mixed strategy stochastic game
- Router vs Tester

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- Router Seeks strategy to distribute traffic over roads to minimize risk
- Tester Develops attack strategy to maximally disrupt smooth flow of traffic
- Perfect knowledge Strategy of adversary is immediately known

#### Simulation setup

- Transportation network represented as a graph *G*(*V*, *E*), with *V* vertices and *E* edges
- Trips are characterized by demand  $D_{|V| \times |V|}(t)$
- Simulation divided into k time intervals  $\Delta T = \langle \Delta t_1 \dots \Delta t_k \rangle$

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• Disrupting a link renders it unavailable for interval  $\Delta t_i$  and is fully restored at  $\Delta t_{i+1}$ 

#### **Mini-max Formulation**

$$\min_{\gamma} \max_{\rho} \mu^{n}(\gamma, \rho) = \sum_{i \in \Delta T} \sum_{e \in E} \gamma^{n}_{e,i} \rho^{n}_{e,i} \tau^{n}_{e,i}$$

 $\mu^n$  system vulnerability in the  $n^{th}$  iteration  $\gamma_{e,i}^n$  usage probability of edge e in interval i and iteration n  $\rho_{e,i}^n$  link attack probability  $\tau_{e,i}^n$  heuristic link travel cost

> Product is summed over all edges and intervals to quantify system vulnerability

#### Probabilities

Link usage probability

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$$\gamma_{e,i}^{n} = \frac{f_{e,i}^{n}}{\sum_{i \in \Delta T} \sum_{e \in E} f_{e,i}^{n}}$$

Tester attack probability

$$\rho_{e,i}^{n} = \frac{\tau_{e,i}^{n} \times \gamma_{e,i}^{n}}{\sum_{i \in \Delta T} \sum_{e \in E} (\tau_{e,i}^{n} \times \gamma_{e,i}^{n})}$$

 $f_{e,i}^{n}$  traffic on edge *e*, interval *i*, in  $n^{th}$  iteration

Link cost

$$C_e^n = \{ \begin{matrix} C_e^- & \text{if } \rho_e^n = 0 \\ C_e^+ = \beta \times |E| \times C_e^- & \text{if } \rho_e^n > 0 \end{matrix} \right.$$

S-expected link cost

$$S_{e,i}^{n+1} = \left( \left( 1 - \rho_{e,i}^n \right) \times C_e^- \right) + \left( \rho_{e,i}^n \times C_e^+ \right)$$

S-expected link costs with Method of Successive Averages (MSA)

$$\tau_{e,i}^{n+1} = \frac{1}{n^{\alpha}} S_{e,i}^{n+1} + \left(1 - \frac{1}{n^{\alpha}}\right) \tau_{e,i}^{n}$$

 $\alpha > 1.0$  rate of convergence

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# **Algorithm Initialization**

**Require:** Road network *G* with *v* vertices and *e* edges **Require:** Dynamic traffic demand data profile  $D_{|V| \times |V|}(t)$  **Require:** Array of time intervals  $\Delta T$  **Require:** Maximum iterations  $N_{max}$ 1: Initialize iteration n = 02: Initialize system vulnerability  $\mu^0 = 0$ 3: for i = 1 to k do 4: for e = 1 to |E| do 5:  $\tau_{e,i}^1 = C_e^-$ 6: end for 7: end for

Heuristic travel cost is initialized to free flow travel time

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# Algorithm

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8: <b>do</b>								
9:	n = n + 1							
10:	$f_{e,i}^n = Simulate(G, \tau_{e,i}^n)$							
11:	for $i = 1 to k do$							
12:	for $e = 1$ to $ E $ do							
13:	Calculate usage probability $\gamma_{e,i}^n$							
14:	Calculate attack probability $\rho_{e,i}^n$							
15:	Calculate link vulnerability							
	$\mu_{e,i}^n = \gamma_{e,i}^n \times \rho_{e,i}^n \times \tau_{e,i}^n$							
16:	Update system vulnerability $\mu^n = \mu^n + \mu_{e,i}^n$							
17:	Update s-Expected link cost $S_{e,i}^n$							
18:	$\tau_{e,i}^{n+1} = MSA(S_{e,i}^{n+1}, \tau_{e,i}^{n})$							
19:	end for							
20:	end for							
21: while $( \mu^n - \mu^{n-1}  > \varepsilon)$ or $(n < N_{max})$								
Algorithm terminates if convergence criterion is met								

or number of iterations exceeds  $N_{max}$ 

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#### Illustration





#### First three iterations for $\Delta t_1$

Link		Iteration 1	l	I	teration 2		Iteration 3			
Name	$\tau^1_{e,1}$	$\gamma_{e,1}^1$	$ ho_{e,1}^1$	$\tau_{e,1}^2$	$\gamma_{e,1}^2$	$ ho_{e,1}^2$	$\tau^3_{e,1}$	$\gamma_{e,1}^3$	$ ho_{e,1}^3$	
L0,1	73.5399	0.0499	0.0528	120.1637	0.1141	0.1380	138.9500	0.0188	0.0231	
L0,2	73.5399	0.0717	0.0758	140.4437	0.0000	0.0000	123.7100	0.1017	0.1116	
<i>L</i> 1,0	73.5399	0.0000	0.0000	73.5399	0.0000	0.0000	73.5400	0.0000	0.0000	
L1,2	88.5347	0.0000	0.0000	88.2923	0.0000	0.0000	88.2900	0.0000	0.0000	
<i>L</i> 1,3	59.1946	0.0374	0.0319	81.8295	0.0794	0.0654	87.7800	0.0220	0.0171	
L2,0	73.5399	0.0000	0.0000	73.5399	0.0000	0.0000	73.5400	0.0000	0.0000	
L2,4	59.1946	0.0443	0.0377	85.9972	0.0034	0.0030	79.8300	0.0733	0.0520	
L3,4	88.5347	0.0000	0.0000	88.5347	0.0000	0.0000	88.5300	0.0000	0.0000	
L3,5	73.7957	0.0458	0.0486	116.8651	0.0892	0.1049	129.3300	0.0451	0.0518	
L4,2	59.1946	0.0000	0.0000	59.1946	0.0000	0.0000	59.1900	0.0000	0.0000	
L4,3	88.5347	0.0000	0.0000	88.5347	0.0000	0.0000	88.5300	0.0000	0.0000	
L4,5	73.7957	0.0403	0.0428	111.6604	0.0172	0.0193	106.4700	0.0749	0.0708	
L5,3	73.7957	0.0000	0.0000	73.7957	0.0000	0.0000	73.8000	0.0000	0.0000	



#### Network Vulnerability (µ)



Variations in vulnerability decrease after 80<sup>th</sup> iteration



#### Change in vulnerability ( $\Delta \mu$ )



MSA places less emphasis on later strategies, forcing convergence



#### Link vulnerability in interval $\Delta t_1$



Link level vulnerability oscillates until a stable solution is found



#### MSA in interval $\Delta t_1$



Crossing link cost at convergence explains oscillation



# Comparison of game-theoretic and deterministic methods

Link			part with relationship ID rld2 was				part with relationship ID rld2 was				part with relationship ID rld2 was	
Name	The image part with relationship ID rld2 was not found in the file.	The image part with relationship ID rld2 was not found in the file.	TT	TT rank	The image part with relationship ID rid2 was not found in the file.	The image part with relationship ID rld2 was not found in the file.	TT	TT rank	The image part with relationship ID rld2 was not found in the file.	The image part with relationship ID rld2 was not found in the file.	TT	TT rank
L <sub>0.1</sub>	0.0639	15	1665	11	0.1394	12	1658	14	0.0817	14	1669	8
L <sub>0.2</sub>	1.4447	1	1672	7	1.0896	5	1664	12	1.2558	2	1668	9
$L_{1.0}$	0.0000	20	1638	16	0.0000	20	1638	16	0.0000	20	1638	16
L <sub>1.2</sub>	0.0000	20	1638	16	0.0000	20	1638	16	0.0000	20	1638	16
$L_{1.3}$	0.0071	19	1672	7	0.0380	17	1667	10	0.0390	16	1769	2
$L_{2.0}$	0.0000	20	1638	16	0.0000	20	1638	16	0.0000	20	1638	16
L <sub>2.4</sub>	0.4149	7	1702	5	0.3081	9	1738	3	0.3085	8	1817	1
$L_{3.4}$	0.0000	20	1638	16	0.0000	20	1638	16	0.0000	20	1638	16
L <sub>3.5</sub>	0.1323	13	1645	15	0.1475	11	1659	13	0.1489	10	1735	4
$L_{4.2}$	0.0000	20	1638	16	0.0000	20	1638	16	0.0000	20	1638	16
$L_{4.3}$	0.0306	18	1638	16	0.0000	20	1638	16	0.0000	20	1638	16
L <sub>4.5</sub>	0.8295	6	1668	9	1.1028	4	1681	6	1.1038	3	1735	4
L <sub>5.3</sub>	0.0000	20	1638	16	0.0000	20	1638	16	0	20	1638	16



# Comparison of game-theoretic and deterministic methods (2)

- Spearman's rank correlation
  - Correlation  $r_s = 0.8882$  at convergence
  - p-value =  $4.63 \times 10^{-14}$ 
    - Strong correlation between approaches
- As size of network increases, number of simulations will decrease



# Comparison of game-theoretic and deterministic methods (3)



Correlation never below  $r_s = 0.8$  and trend increases

#### Conclusion

 Presented a game-theoretic approach to assess dynamic vulnerability of transportation network

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- Considers relative vulnerability of all links and time intervals in parallel
- Results indicate that game-theoretic approach achieves strong correlation to slower deterministic method

#### Future Research

 Address performance and accuracy challenges to scale game- theoretic approach to larger networks

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 Utilize game-theoretic dynamic transportation network vulnerability approach to allocate limited defensive resources to links at specified times to mitigate vulnerability most effectively