Extending the I-95 Rule-based Incident Duration System with an Automated Knowledge Transferability Model by Yen-Lin Huang, Yi-Ting Lin, and Gang-Len Chang The University of Maryland, College Park Knowledge Transferability Analysis (KTA) Knowledge-based System Abstract

- The rule-based incident duration prediction model (IDPM), covering I-95, I-495, and I-695, has been adopted by MDOT SHA in daily responses to non-recurrent congestion.
- In light of its effectiveness and robustness in practice, expanding such a system to all other highway emerges as a desirable but challenging task due to:
- □ The need of integrating field operators' expertise in generating prediction rules
- □ The **dependence of sufficient incident records** for key parameter calibration
- To circumvent such a data-demanding and time-consuming process for extending IDPM's spatial coverage, this study has proposed a Knowledge Transferability Analysis (KTA) method, featuring its automated process to assess, select, and transfer existing prediction rules to perform incident duration estimate for the new target highway
- Evaluation of the proposed KTA with the incident records from Maryland I-70 reveals that it can achieve the accuracy of 87% with the training dataset (i.e., 2016-2018) and of 82% with the test dataset (i.e., 2019), comparable to the current system's performance but demanding much fewer incident records for model calibration and significantly fewer efforts for system expansion



I-70 in Maryland was selected for the case study. The 2016-2018 incident records from the CHART II Database were for model calibration, and those from 2019 served for performance evaluation





Transferability Analysis





		Cate
Rules of CPI1	<30 minutes ≥30 minutes	Cate
Rules of CPI2	<60 minutes ≥60 minutes	(# of Cates (First
Rules of CPI3 → ···	<120 minutes ≥120 minutes	Cate (Veh
Rules of CPD1 → ···		Cates (Pave
Rules of CPD2 → ···		Cates (Lane
		Cates (Ope
Rules of CPD3		Cates (Time

ategory	Description
ategory-1 # of responders)	the number of different responders at the incident scene
ategory-2 First arrived responder)	type of the first-arriving responders
ategory-3 /ehicle status)	the number and the type of vehicles involving in incidents and their damage levels
ategory-4 Pavement conditions)	indicators for the pavement conditions
ategory-5 .ane blockage)	indicators to denote the lane- blockage conditions
ategory-6 Operation center)	indicators reflecting different incident response centers
ategory-7 Time)	temporal-related indicators associated with an incident

Transferability Ranking Analysis

Transferability Ranking-I Analysis: All rules based on their logic structure and Using **Random forest** to identify the relative target incident types are assigned with a impacts of seven categories of factors on customized score with following regulations the resulting incident durations revealed in • Take optimized rankings as grades the target new system's incident records • The rule combined by "AND", sum up the • Transferability Ranking-II Analysis: qrade Using Data Envelopment Analysis (DEA) to • The rule combined by "OR", take the rank the importance associated with minimum combination and add 100

categories of factors from the perspective of how often they have been used in the existing prediction rules and the resulting effectiveness

• Transferability Ranking Integration Employing the method for rank aggregation to produce final optimized ranking list for all categories

 δ_i^* is the optimal rank for category *i*; r_i^1 and r_i^2 are the resulting ranks from Rank-I and Rank-II test for category i; and w_i stands for the number of existing rules using one or more factors from category *i*

An example of an Application

Collision, CPI2	Transferring Priority Grade	Accuracy	Transferable or not]		EVall	aleu by gi	oups of me		as	
[Tow service arrived] ≥30 mins	1	86%(24/28)	Yes			CPI1	CPI2	CPD1		CF	Overall
sponse units] ≥30 mins	1	100%(1/1)	Yes			CITI					overail
≥30 mins	1	NaN%(0/0)	No		Training set	87.80%	85.37%	86.17%	87.50%	100%	86.94%
↓] AND [More than 4 response units] ≥30 mins	4	0%(0/1)	No	[Tow service arrived] Yes ≥30 mins	(2016-2018)	(36/41)	(35/41)	(81/94)	(35/40)	(6/6)	(193/222
involved vehicles] AND ≥30 mins rcycle involved]	. 6	NaN%(0/0)	No	[More than 4 response units] Yes ≥30 mins	Test set	100.00%	68.75%	/8.26%	90.00%	100%	81.6/%
AOC center] ≥30 mins	. 6	NaN%(0/0)	No	No [Dry pavement]	(2019)	(10/10)	(11/16)	(18/23)	(9/10)	(1/1)	(49/60)
	11	NaN%(0/0)	No	No local-unique Prediction rule mining							
↓ lane blocked] AND [Pickup [More than 6 response units]	101	0%(0/1)	No	[Snow-ice pavement] Yes ≥30 mins			Con	clue	inne		
an 3 response units] OR [Police ved] OR [Truck involved]) AND re than 3 involved vehicles	103	NaN%(0/0)	No	No Else then (# of remaining data = 2)				CIUS			
me] OR [More than 2 involved ≥30 mins	104	NaN%(0/0)	No								
D [venicle over turned]	1 104	1411/2(0/0)									
[Winter] OR [Night] ≥30 mins	104	29%(2/7)	No								
[Winter] OR [Night] ↓	104 105 204	29%(2/7) 89%(8/9)	No Yes		To circul ovtonsity	nvent the	demanding	developing	g efforts ar	nd the ne	eed for an
[Winter] OR [Night] ↓ [Dry pavement] [Tow service arrived] [More than 4 response unit arrived] [Dry pavement]	104 105 204	29%(2/7) 89%(8/9)	No No Yes Then >30 >30 < 30	Cases with 0 < CT < 30 minutes	 To circum extensive study h construction 	mvent the e dataset f as propos tion of a r ed rules wit	demanding for calibrat ed an in new systen ch an auton	developing ion of an novative K n to take a nated proce	g efforts ar IDPM's pre TA model advantage	nd the ne ediction that a of existin	eed for an rules, this llows the ng IDPMs'
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 $Min \sum (|\delta_i^* - r_i^1| + |\delta_i^* - r_i^2|) \times w_i$

Prioritizing Candidate Rules

- The rule for estimating incident durations of <30, <60, or <120, add 200

Transferability Effectiveness Test

Two MOEs, 1) confidence level and 2) support *level*, are used to determine the transferable rules among the rules from the Rule Box



Model Evaluation