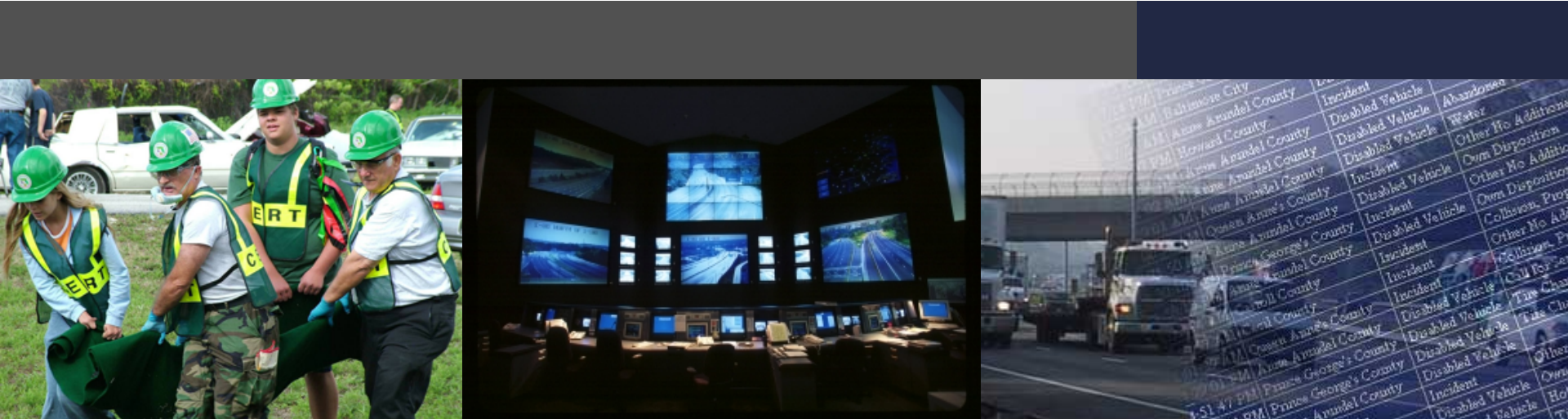


Development of a Traffic Incident Management System for Contending with Non-recurrent Highway Congestion



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Overview

- ❖ Introduction
- ❖ Component 1: Incident response management strategies
- ❖ Component 2: Prediction models for clearance times
- ❖ Component 3: A detour decision support system
- ❖ Contributions, future research, and conclusions

Research Background

- Non-recurrent traffic congestion due to incidents has contributed **up to 60 percent of the total freeway corridor delay** in the United States (Lindley, 1987).
- **About 25 percent of congestion** in the U.S. is incident-related (FHWA, 2005).



The starting time and duration of non-recurrent congestion, **due mainly to incidents**, are random in nature.



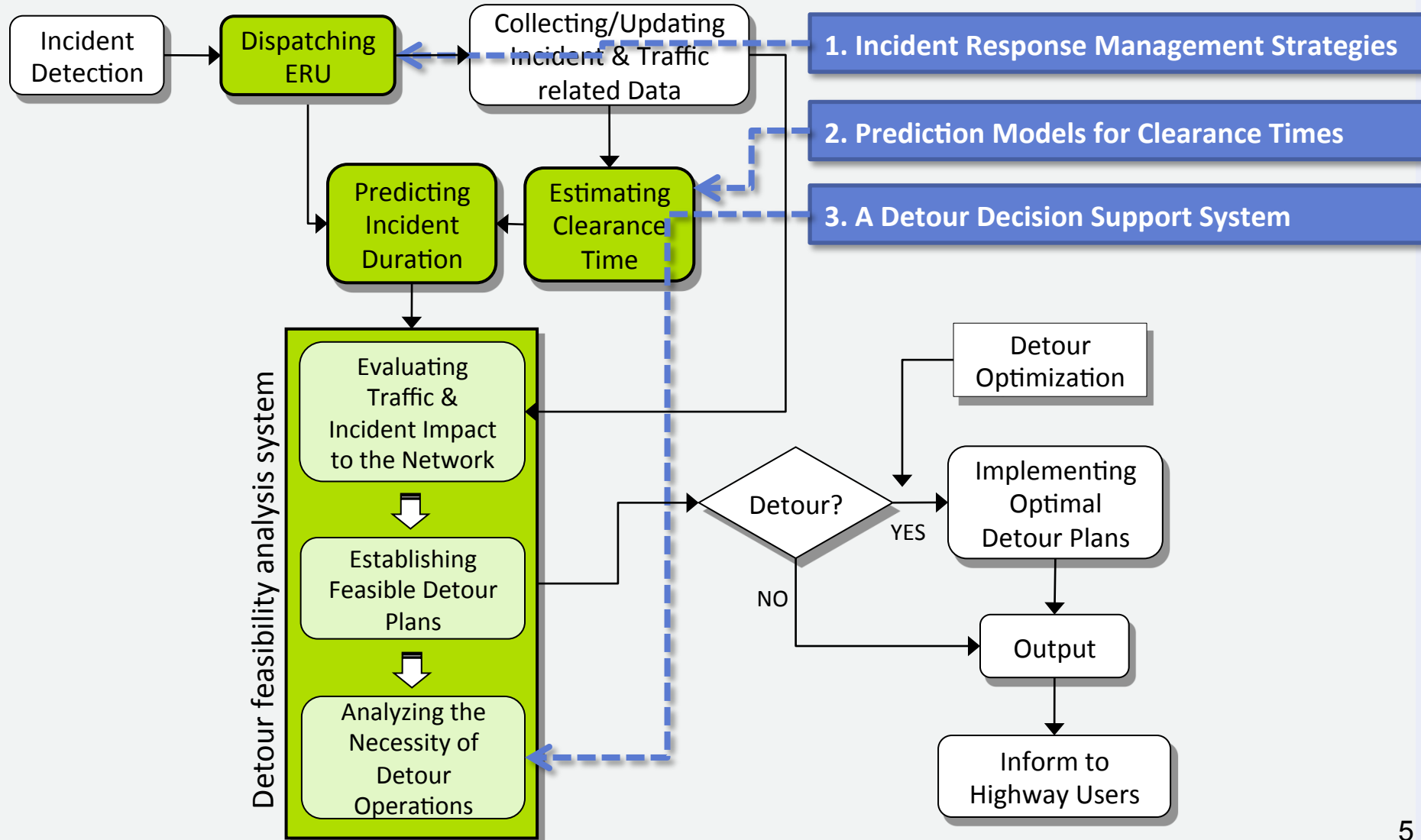
Thus, it is critical to have an efficient and effective incident management system.



Key Tasks in an Incident Management System

- ❖ An optimal deployment strategy for response units
- ❖ Estimation of required clearance times for reported incidents
- ❖ Detour feasibility analysis
- ❖ Detour optimization analysis
- ❖ Provide travel time information to roadway users
 - queue, delay and travel time analysis

Incident Management System



Needs for Each Component

1. Incident Response Management Strategies

- To **maximize contributions** of incident response units **with limited resources** by assigning them to optimal locations.

2. Prediction Models for Clearance Times

- To contend with stochastic nature of clearance times so as to **maximize the system's operational reliability**.

3. A Detour Decision Support System

- To facilitate responsible agencies to **perform efficient traffic management in real time operations**.

Research Objectives

1. Incident Response Management Strategies

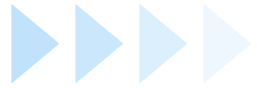
- Develop a deployment strategy for incident response units **to minimize the total incident-induced delay**

2. Prediction Models for Clearance Times

- Develop **a reliable model to estimate the clearance duration** of a detected incident, and to identify critical contributing factors as well as their interrelationships

3. A Detour Decision Support System

- Develop a detour decision support model for control center staff **to determine the necessity of detouring traffic**



1. Incident Response Management Strategies

Literature Review

❖ Facility location problem

- how many response units are needed?
- where should they be allocated in response to the temporal and spatial distribution of incidents?
 - 1) Covering models (Toregas et al., 1971; Schilling et al., 1979; Hogan and ReVelle, 1986; Nair and Miller-Hooks, 2009)
 - 2) P-median models; and (Hakimi, 1964; Carson and Batta, 1990; Haghani et al., 2003; Yang et al., 2005)
 - 3) P-center models (Sylvester, 1857; Garfinkel et al., 1977; ReVelle and Hogan, 1989; Talwar, 2002)

❖ Minimize the number of service stations, the total operational costs, or to maximize the demand (incidents) covered by the pre-determined number of facilities

Data Sources

- ❖ Incident management program operated by Maryland state highway administration (MDSHA)
 - Coordinated Highways Action Response Team (CHART)
 - Has documented incident-related information over the past two decades
 - Date/time, location, nature, involved vehicles, lane closure...



Effectiveness of CHART

CHART responded approximately **81 % (22,796/28,345)** of incidents during last two years



Does the CHART involvement matter?

	w/o CHART	w/ CHART
Mean on CT (mins)	37.91 → 27.51	

CHART **reduced** the avg. clearance time **by 27 %**



Does the **prompt** CHART response matter?

First responder	Others	CHART
Mean on CT (mins)	39.49 → 21.85	

CHART **reduced** the avg. clearance time **by 45 %**

Needs for Research

- ❖ The efficient response of CHART can contribute to the reduction **in not only the response time but also the clearance time** → **reduction in delay**
- ❖ However, not all incidents can be promptly responded by CHART due to their **limited resources**
- ❖ Therefore, it is critical to develop a strategy to **optimally deploy available response units so as to maximize their contributions**

Model Construction

❖ Inputs

- Incident distribution, incident duration, lane blockage information, traffic volume, capacity, and available resources

❖ Objective function

- **Minimize the total delay** induced by incidents

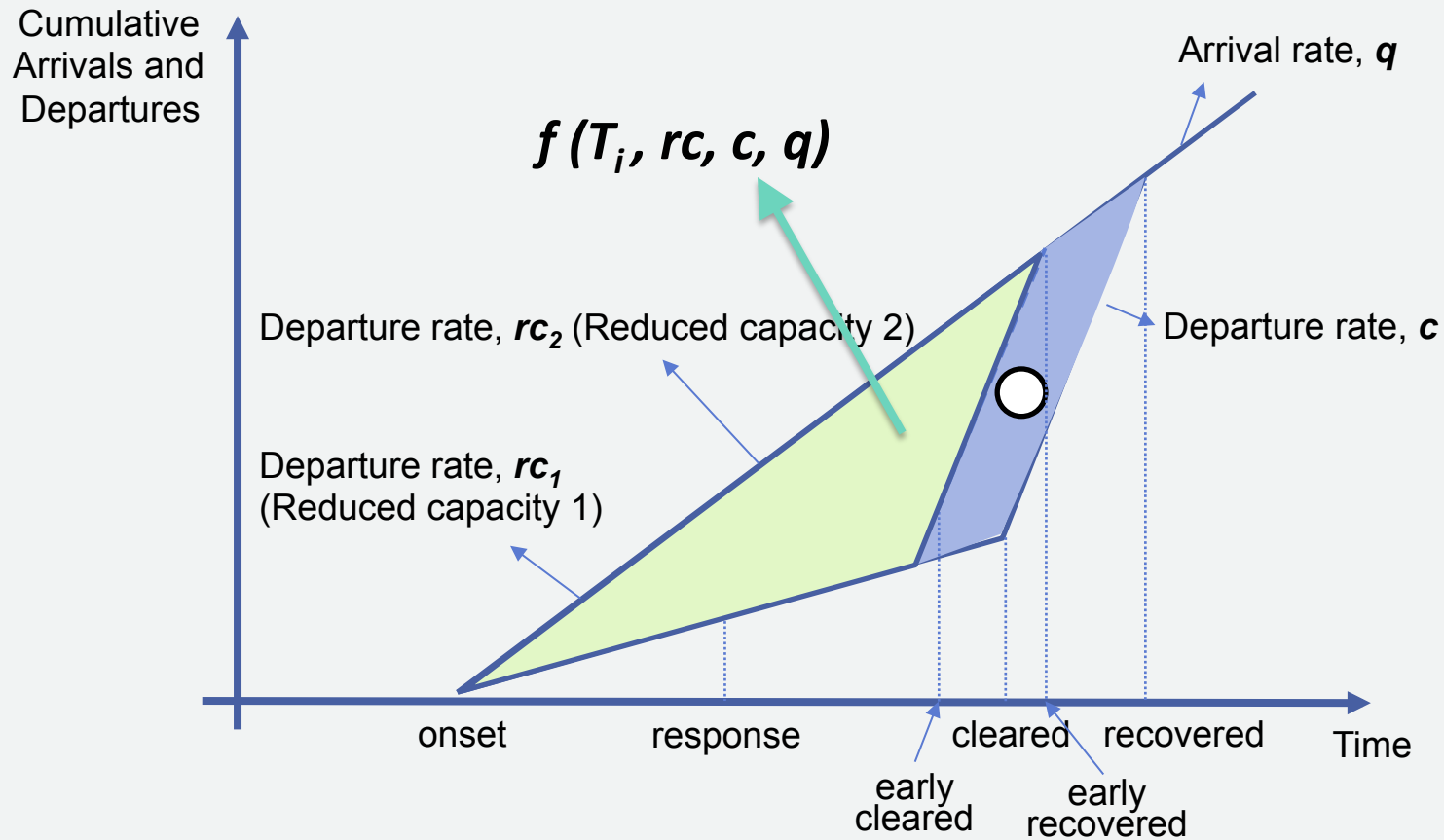
❖ Constraints

- Every freeway segment must be served by one unit
- Response units can only be dispatched from location i if they are stationed there
- The total number of response units is limited by available resources

❖ Outputs

- Assigned station and coverage for each unit

Relations between Incident Duration and Total Delay



- Ⓐ Reduced delay due to additional lanes open
- ✓ Ⓑ Reduced delay due to reduced clearance duration

Model Formulation

Objective Function: Min total delay for responded incidents

$$\min_{x,y} \sum_i \sum_j x_{ij} \cdot f_j \cdot d_{ij}(t_{ij})$$

Total Delay

1. Delay from incidents occurring at node j (Olmstead, 1996)

$$d_{ij}(t_{ij}) = \frac{1}{2} T_{ij}^2 (q_j - rc_j) (c_j - rc_j) / (c_j - q_j)$$

T_{ij} : Response Time + Clearance time

2. Response time and clearance time (Olmstead, 1996)

$$T_{ij}^2 = \{ (RT_1 + CT_1)^2 + Var(CT_1) \} \& (t_{ij} + CT_{2-1})^2 + Var(CT_{2-1}) \& (RT_2 + CT_{2-2})^2 + Var(CT_{2-2})$$

Stochastic nature

CHART is not involved

CHART is involved and first responder

CHART is involved but not the first responder

- $x_{ij} = 1$ if incidents at j are responded by a response unit at i
- $y_i = 1$ if a response unit is stationed at i
- $G(N,A)$: a network of freeways, where N and A are the sets of nodes and links
- i, j : index for nodes $i, j \in N$
- f_j : probability that an incident occurs at node j
- t_{ij} : travel time from i to j
- d_{ij} : delay from incidents occurring at node j according to t_{ij}
- T_{ij} : response time + estimated clearance time according to t_{ij}
- q_j : traffic volume at j
- c_j : capacity at j
- rc_j : reduced capacity at j

Model Formulation (cont'd)

Constraints:

1. Every freeway segment must be served

$$\sum_{i \in N} x_{ij} = 1 \quad \forall i \in N$$

2. Response units can only be dispatched from location i if they are stationed there ($y_i = 1$)

$$x_{ij} \leq y_i \quad \forall j \in N$$

3. The total number of available response units is limited by available resources (R)

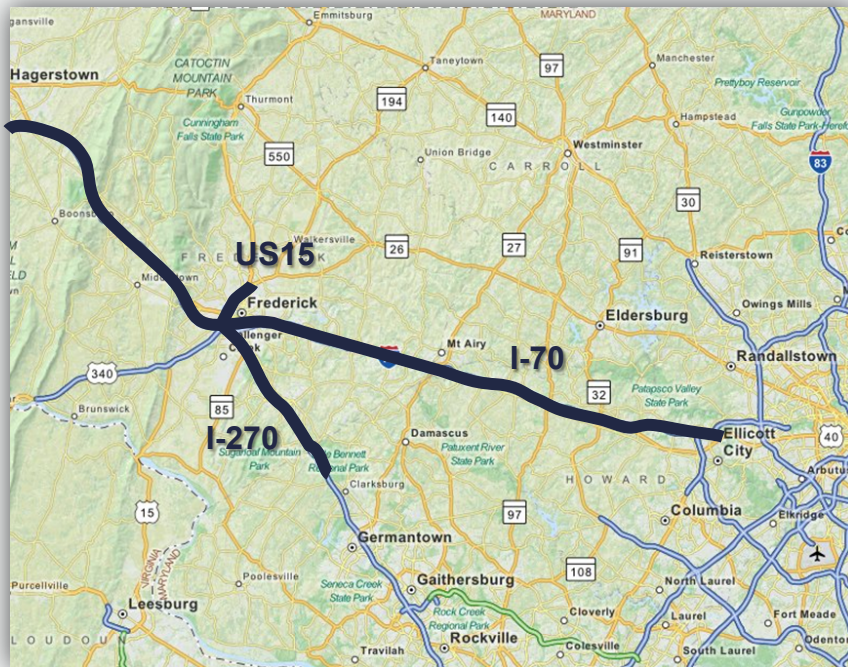
$$\sum_{i \in N} y_i \leq R$$

$$x_{ij} \in [0,1] \quad \forall (i,j) \in N \quad y_i \in [0,1] \quad \forall i \in N$$

- $x_{ij} = 1$ if incidents at j are responded by a response unit at i
- $y_i = 1$ if a response unit is stationed at i
- $G(N,A)$: a network of freeways, where N and A are the sets of nodes and links
- i, j : index for nodes $i, j \in N$
- R : available resources

Empirical Study

❖ Segments of I-70, I-270 and US 15 in MD



Site Characteristics

- 63 miles
- Radial shape of roads
- Frederick, Howard, and Carroll Counties

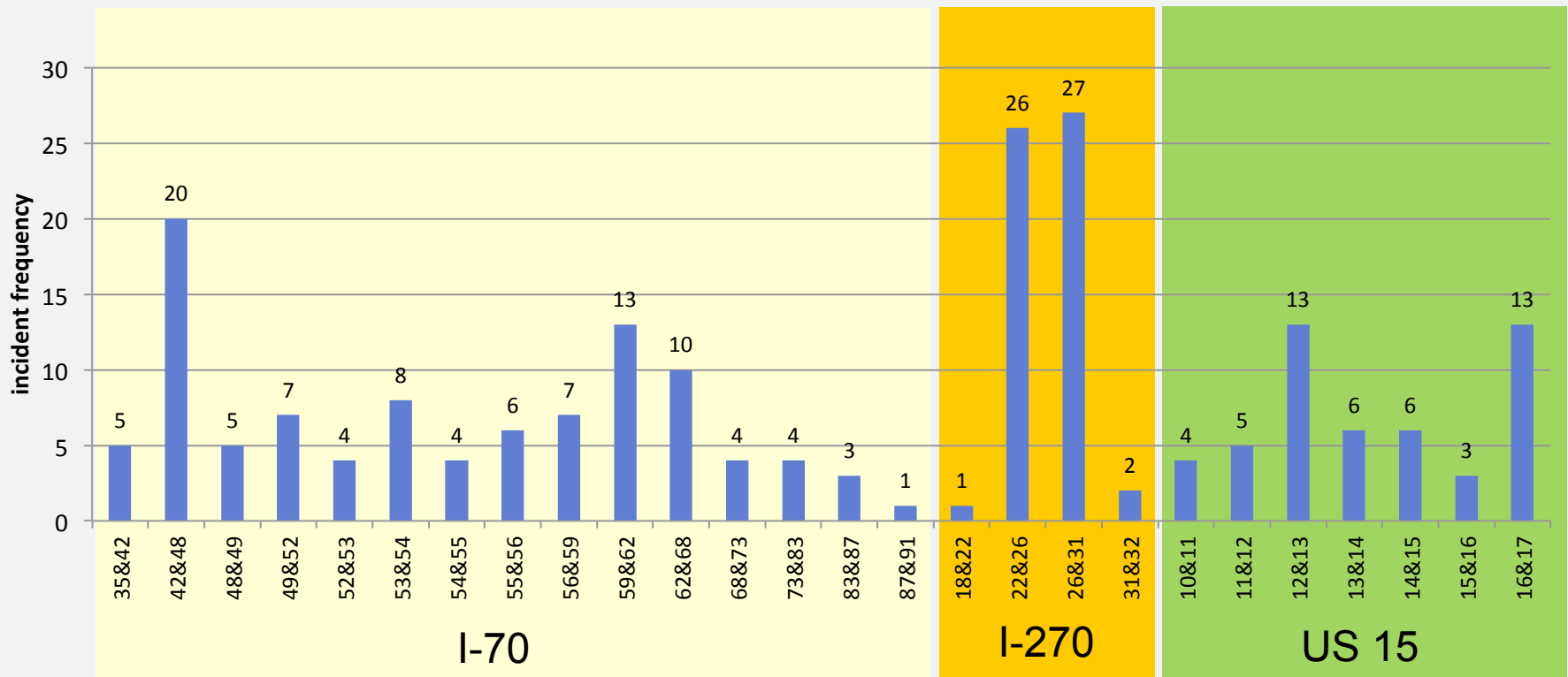
Highway Incident Management

- TOC-7
- 3 units
- Operation Hours:
5AM – 9PM on weekday
- **Study Period: AM peak**
(7AM – 9:30 AM on weekday)



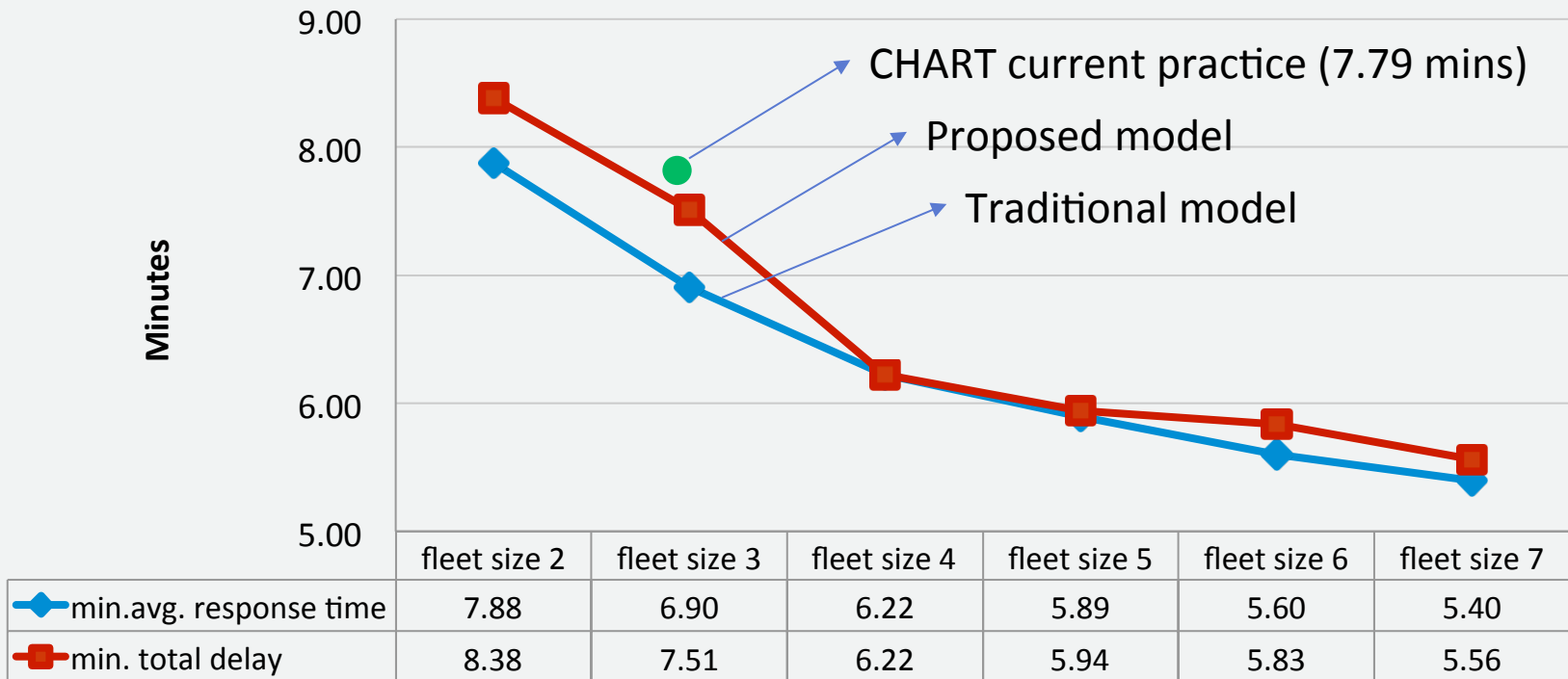
Incident Frequency Distribution

Incident frequencies fluctuate over the network !



Model output Analysis

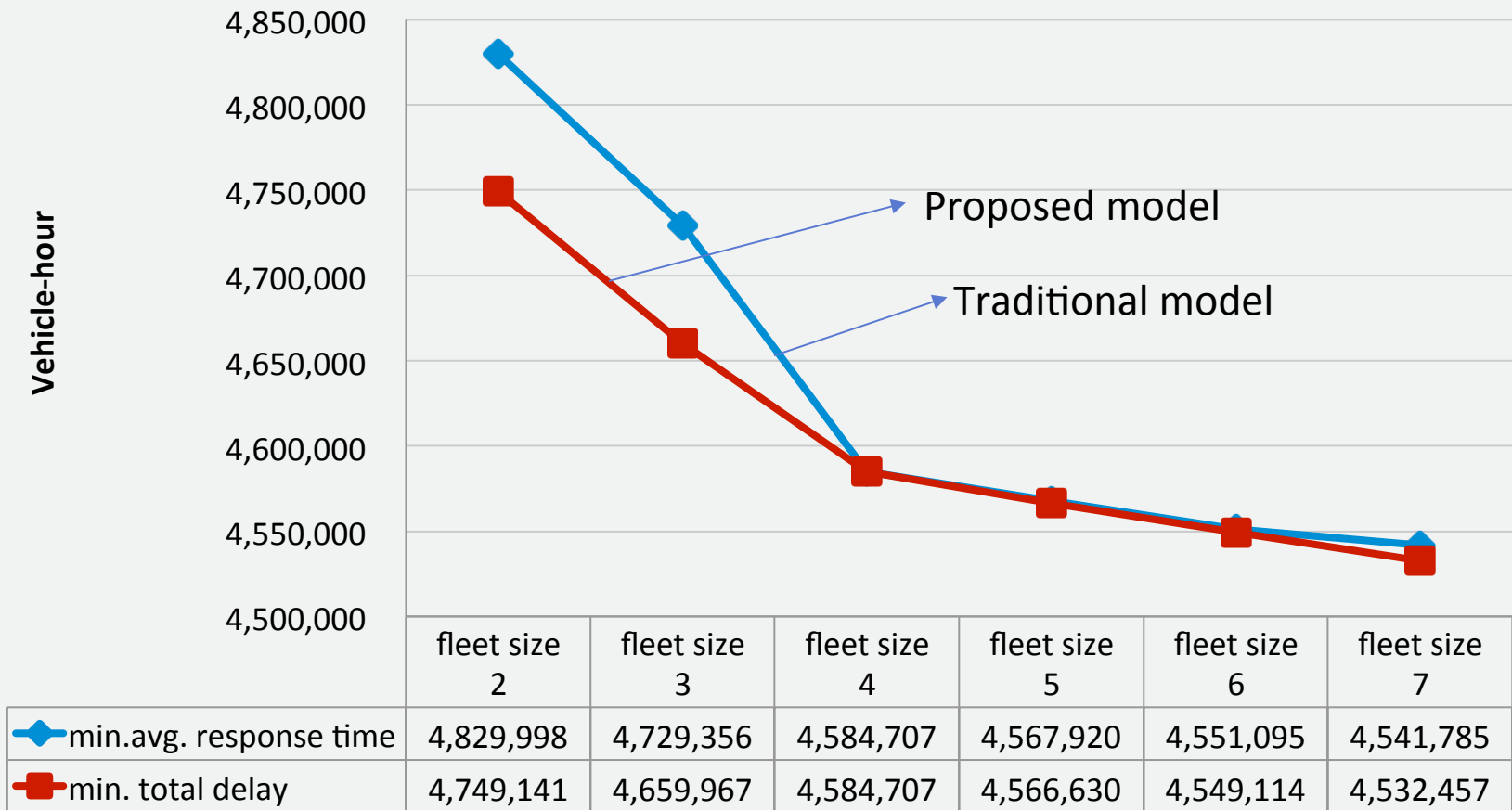
1. Assigned station and service coverage for each unit
2. Average travel time (minutes)



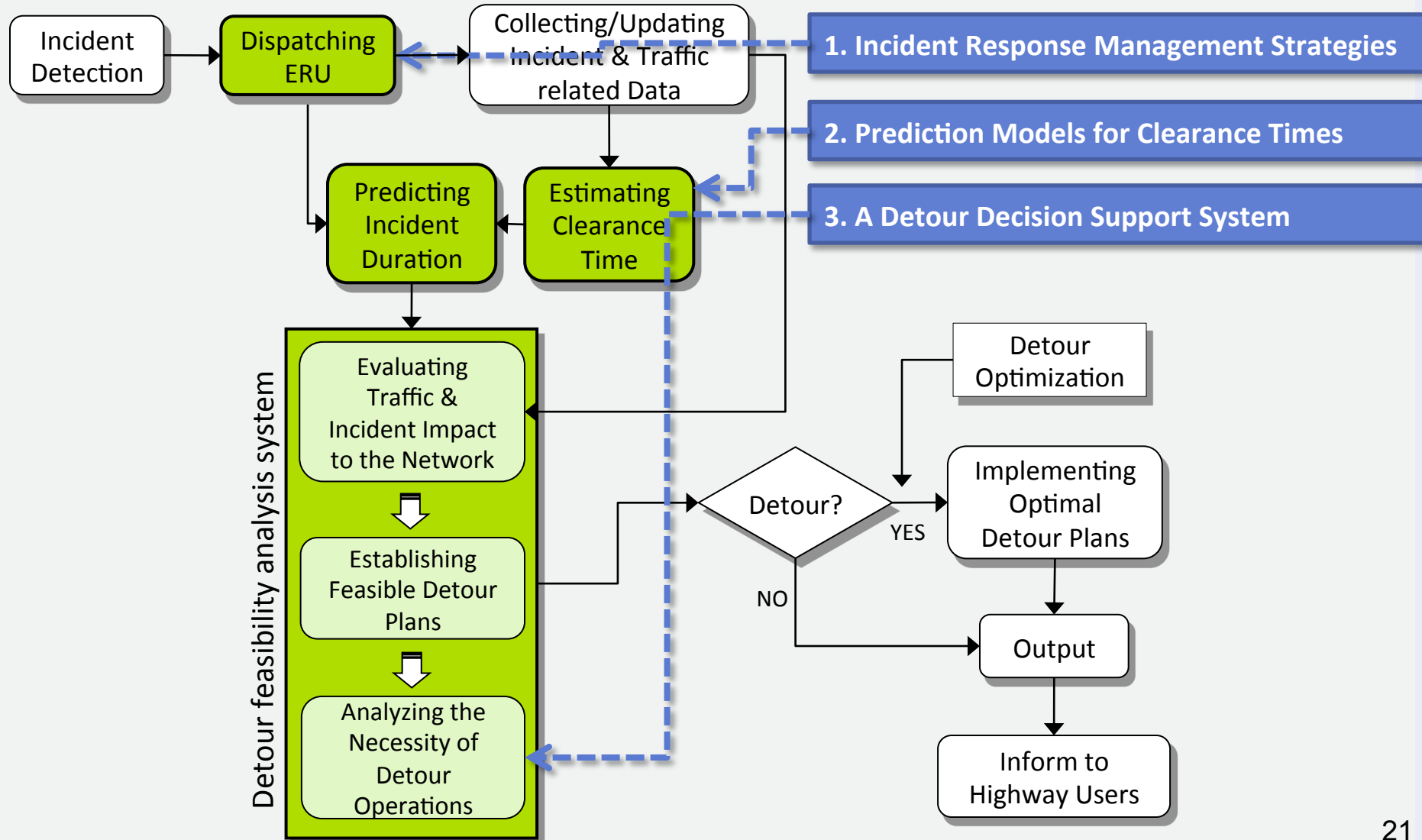
Model output analysis (cont'd)

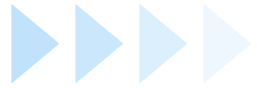
3. Total delay (veh-hr)

● CHART current practice (5,612,805 veh-hr)



Incident Management System





2. Prediction Models for Clearance Times

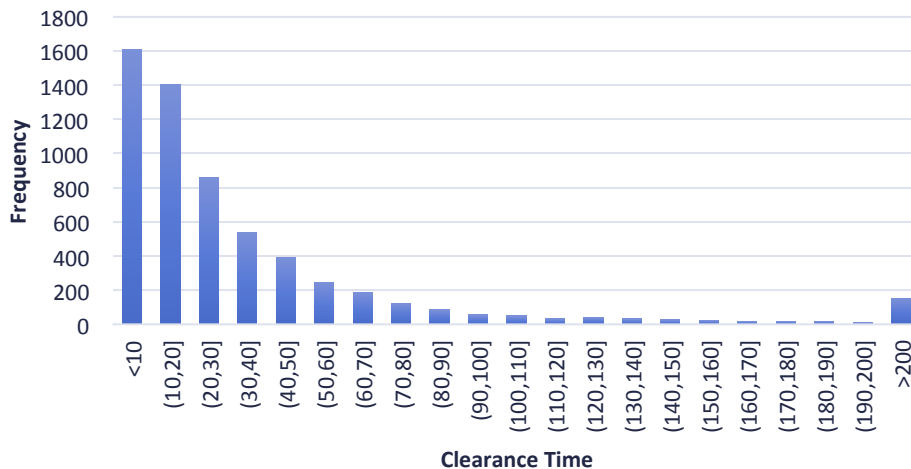
A Model for Estimating Clearance Times

Why do we need such a model?

- ❖ Key input for the incident management system
 - Optimal deployment strategy analysis
 - Detour feasibility analysis
 - Detour optimization analysis
 - Traveler information – queue, delay and travel time analysis
- ❖ However, the required clearance time for a reported incident is **very difficult to reliably predict in advance.**

Challenge to Predict Clearance Times

❖ Skewed shape and distributed in a wide range



CT (mins)	Frequency	Ratio
<=30	3870	65%
30-60	1176	20%
60-90	397	7%
90-120	138	2%
>120	344	6%
total	5925	100%

- Difficult to fit with a continuous or discrete distribution
- Most statistical models cannot perform well
 - They tend to focus on the major classes of the data



However, most studies in the literature applied statistical approaches to develop a model

Literature Review

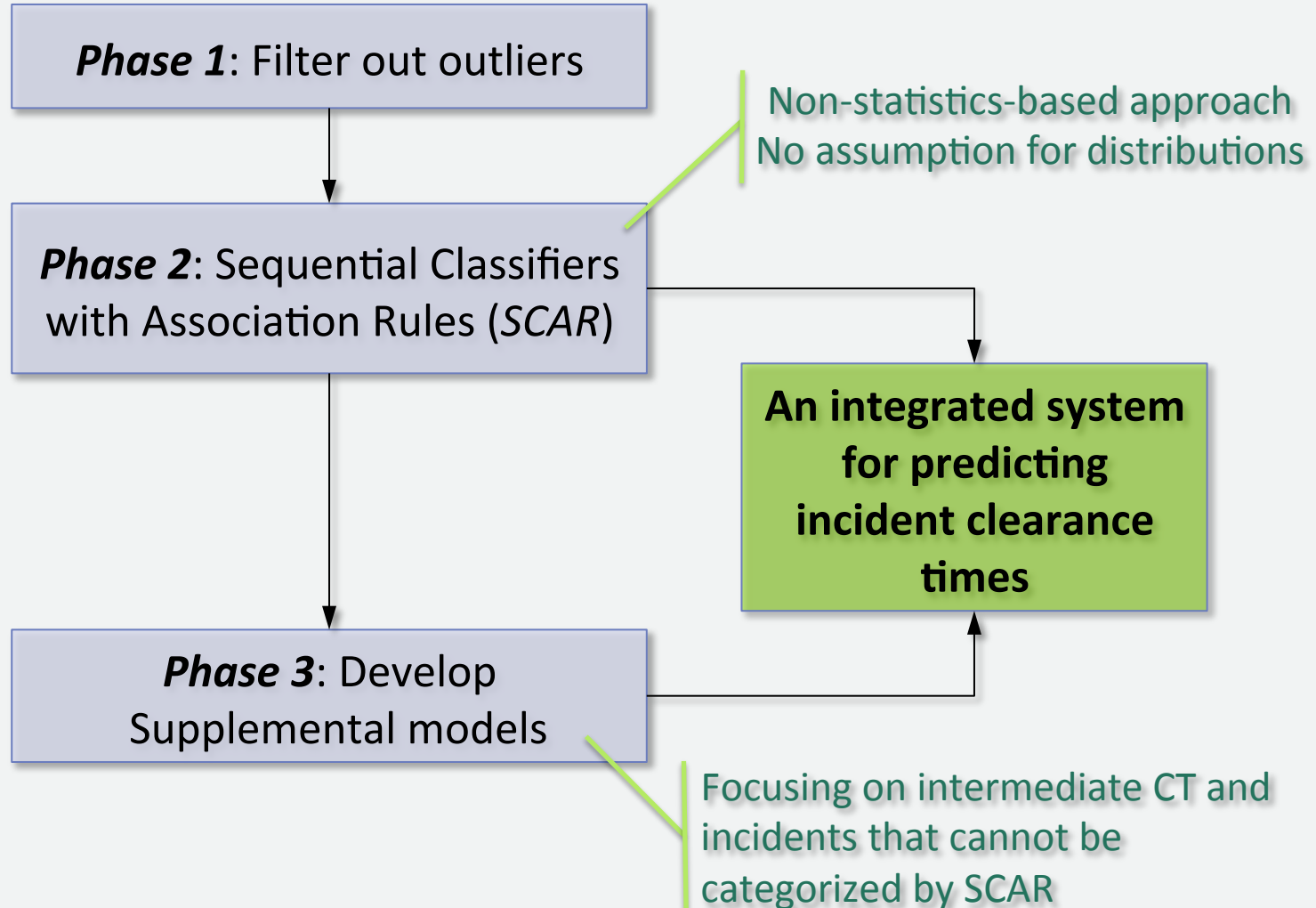
- 1) Probabilistic Distributions (Garib et al., 1997; Giuliano, 1989; Garib et al., 1997; Ozbay and Kachroo, 1999)
 - In the most literature
 - Using **limited scale data**
 - **No validation** for models
- 2) Conditional Models (Boyles et al., 2007; Ring, 2000; Boyles et al., 2007)

- 3) Regression Models (Khattak et al., 1995; Giuliano, 1989; Garib et al., 1997; Ozbay and Kachroo, 1999)

- 4) [The proposed Model is]
 - ✓ **tackling heterogeneity** in most incident data sets
 - ✓ enhancing prediction **accuracies**; and
 - ✓ assessing the prediction model's **robustness** for different data sets
- 5) []
- 6) []

- 7) Unconventional Models (Wang et al., 2005; Wu et al., 2011)

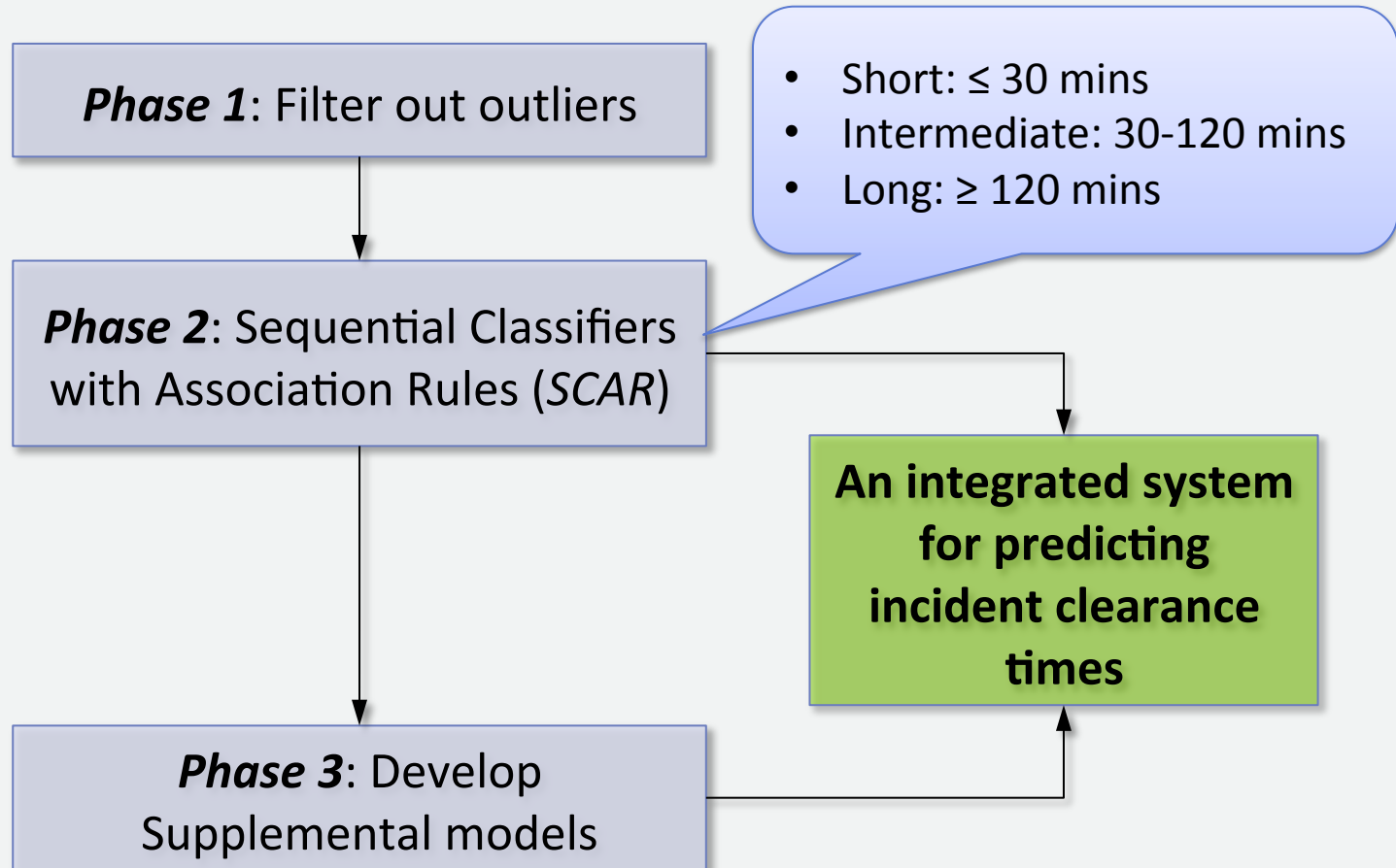
Flowchart to Develop the Proposed Model



Phase 1 – Filter Out Outliers

- ❖ PAM: Partitioning Around Medoids (Kaufman and Rousseeuw, 1990)
 - Medoids: most centrally located elements
 - Goal: detecting a group of clusters including a small number of elements
- ➡ **36/6000 incidents are selected as outliers**

Flowchart to Develop the Proposed Model



Association Rules (Agrawal et al., 1993)

- ❖ Mining explicit relations between clearance time and associated factors in a format of rules.

- For example,

$$\begin{array}{ccc}
 \{\text{personal injury, rainy}\} & \rightarrow & \{\text{intermediate CT}\} \\
 \textit{antecedent} & & \textit{consequent} \\
 \text{LHS} & & \text{RHS} \\
 X & & Y \\
 6 & & 3
 \end{array}$$

- **Support** of an itemset X ($\text{supp}(X)$): the proportion of data entries in the database which include the itemset X
- **Confidence** of a rule: $\text{conf}(X \rightarrow Y) = \text{supp}(X \cup Y) / \text{supp}(X)$

$$\text{supp}(X)=6 \quad \text{supp}(X \cup Y) = 3 \quad \text{conf}(X \rightarrow Y) = 3/6 = 0.5$$

Procedure to Construct the SCAR System

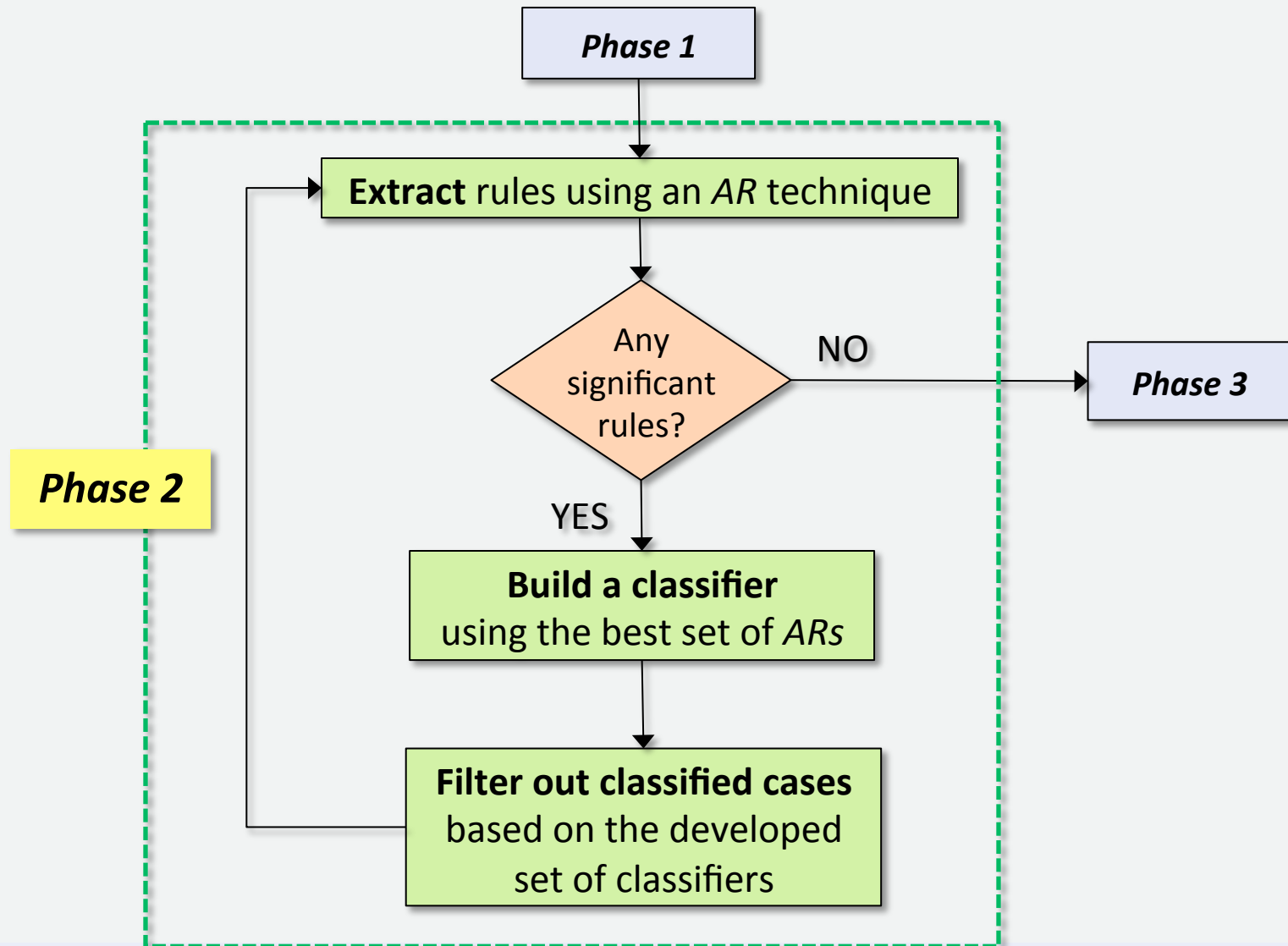


Illustration of the SCAR System



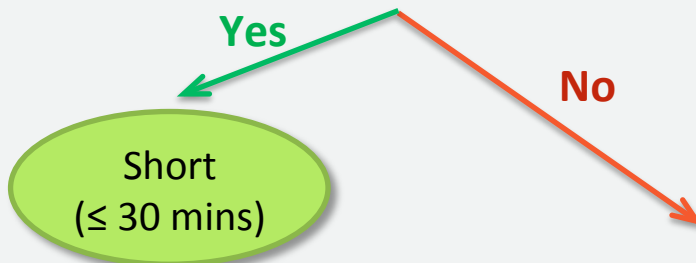
An Incident Is Detected

(TOC3 & Minor lane closure & no info. for pavement condition)
(AOC & 12 lanes on both & on US-50)
(on weekday & disabled veh & CHART detected)

or
or

-----> AR 1
-----> AR 2
-----> AR 3

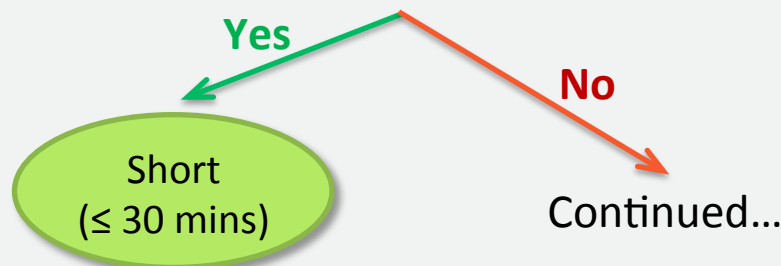
Classifier 1



(2 veh involved & fatality) or
(during daytime & minor road & fatality)

-----> AR 1
-----> AR 2

Classifier 2



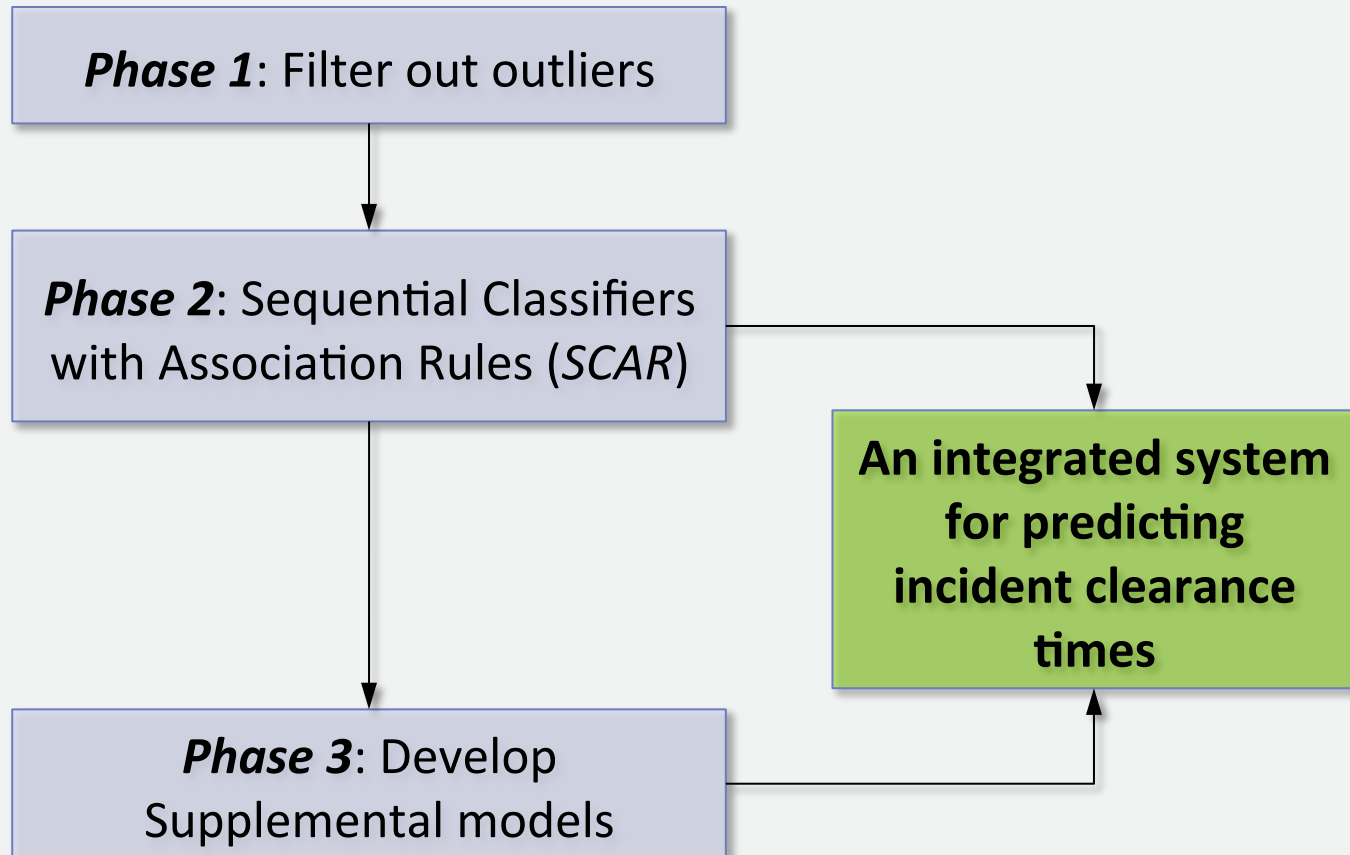
Phase 2 - Model Results

- ❖ 44 Classifiers
- ❖ Each consists of 2 or 3 ARs
- ❖ **About 72% of samples can be explained with SCAR**
- ❖ Accuracies for each category of clearance duration

Clearance Time	Class ratio	# of Classifiers	Accuracy	
			Train	Test
Short	64.98%	27	87.70%	90.37%
Intermediate	28.95%	13	90.50%	92.51%
Long	6.07%	4	75.86%	79.66%

30 -120 mins

Flowchart to Develop the Proposed Model



- ➡ To predict CT for incidents that ***cannot be classified by SCAR***
- ➡ To classify predicted ***intermediate CT*** into shorter intervals

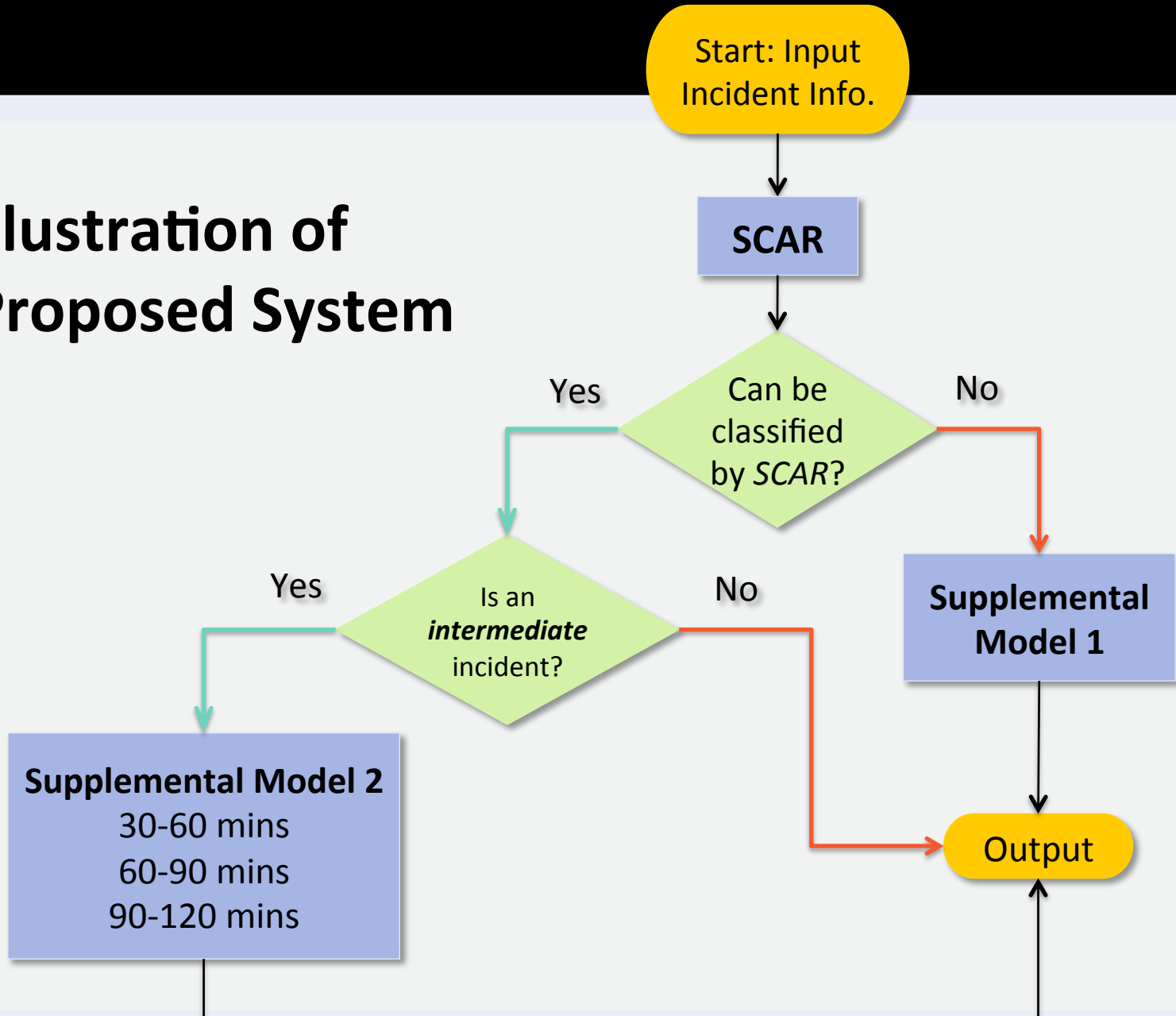
Phase 3 – Developing Supplemental Models

1. A model for data **not classified by SCAR**
 - ≤ 30 minutes
 - 30 – 60 minutes
 - 60 – 90 minutes
 - 90 – 120 minutes
 - > 120 minutes
2. A model to classify the predicted **intermediate clearance times** into **smaller intervals**
 - Intermediate-sub1: 30 – 60 minutes
 - Intermediate-sub2: 60 – 90 minutes
 - Intermediate-sub3: 90 – 120 minutes



Support Vector Machine and Random Forests are applied

Illustration of the Proposed System



MOEs to Evaluate the System's Performance

Contingency Table (c_{ij})

Clearance Duration (minutes)		Observation				
		≤ 30	30 – 60	60 – 90	90 – 120	> 120
Prediction	≤ 30	1068	95	20	3	11
	30 – 60	130	146	50	16	23
	60 – 90	81	96	33	9	5
	90 – 120	13	37	23	9	5
	> 120	8	12	9	8	60

Accuracy

Weights (w_{ij})

Clearance Duration (minutes)		Observation				
		≤ 30	30 – 60	60 – 90	90 – 120	> 120
Estimation/ Prediction	≤ 30	1	0	0	0	0
	30 – 60	0.75	1	0	0	0
	60 – 90	0.5	0.75	1	0	0
	90 – 120	0.25	0.5	0.75	1	0
	> 120	0	0.25	0.5	0.75	1

MOEs to Evaluate the System's Performance

Contingency Table (c_{ij})

Clearance Duration (minutes)		Observation				
		≤ 30	30 – 60	60 – 90	90 – 120	> 120
Prediction	≤ 30	1068	95	20	3	11
	30 – 60	130	146	50	16	23
	60 – 90	81	96	33	9	5
	90 – 120	13	37	23	9	5
	> 120	8	12	9	8	60

Acceptability

Weights (w_{ij})

Clearance Duration (minutes)		Observation				
		≤ 30	30 – 60	60 – 90	90 – 120	> 120
Estimation/ Prediction	≤ 30	1	0	0	0	0
	30 – 60	0.75	1	0	0	0
	60 – 90	0.5	0.75	1	0	0
	90 – 120	0.25	0.5	0.75	1	0
	> 120	0	0.25	0.5	0.75	1

$$= \sum_i \sum_j w_{ij} * c_{ij} / \sum_i \sum_j c_{ij}$$

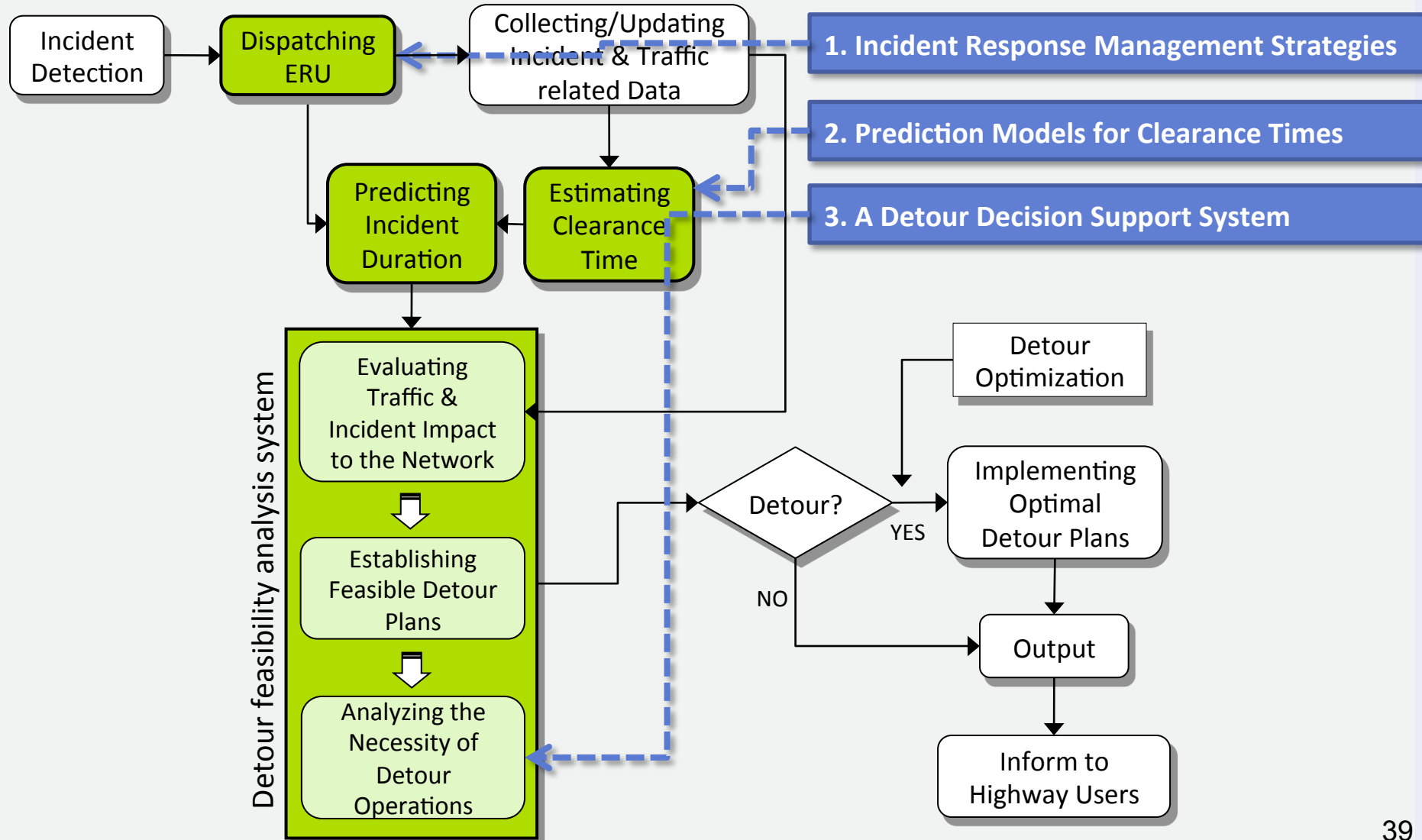
w_{ij} : weights for cells (i, j)
 c_{ij} : number of cases in a cell (i, j)

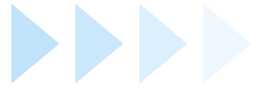
Overall System Performance

Incident Categories	Clearance Duration (minutes)	Class ratio	Accuracy		Acceptability	
			Train	Test	Train	Test
Minor	<= 30	65.0%	80.3%	82.2%	92.0%	93.0%
Intermediate-sub1	30 – 60	20.0%	38.1%	37.8%	58.0%	62.2%
Intermediate-sub2	60 – 90	6.6%	35.9%	24.4%	45.0%	40.7%
Intermediate-sub3	90 – 120	2.4%	46.2%	20.0%	54.8%	33.3%
Major	120 +	6.0%	57.5%	57.7%	57.5%	57.7%
Total		100.0%	66.7%	66.8%	79.1%	80.2%

❖ **Better than five comparable models** developed using support vector regression, random forests, and multiple linear regression, **in terms of accuracy and acceptability**

Incident Management System





3. A Detour Decision Support System

Study Background

Most states consider **only**

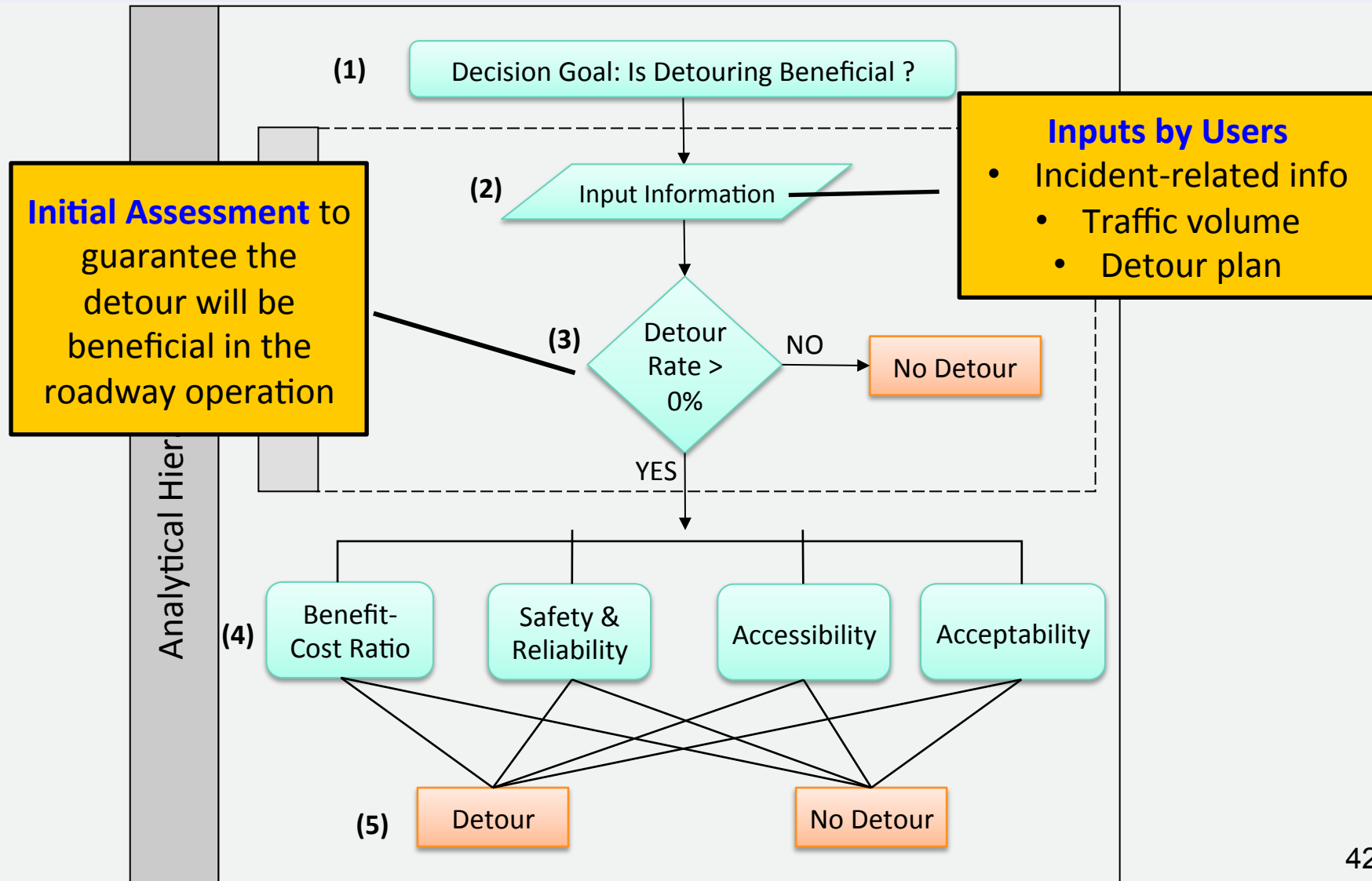
- Incident duration > 30 minutes
- Complete road closure



The proposed model :

- ✓ **Account for more critical factors**
 - Traffic volumes, benefit, cost, safety, travel times, etc.
- ✓ **Allow the decision maker to place different weights to different factors**, based on the either resource constraints or priority.

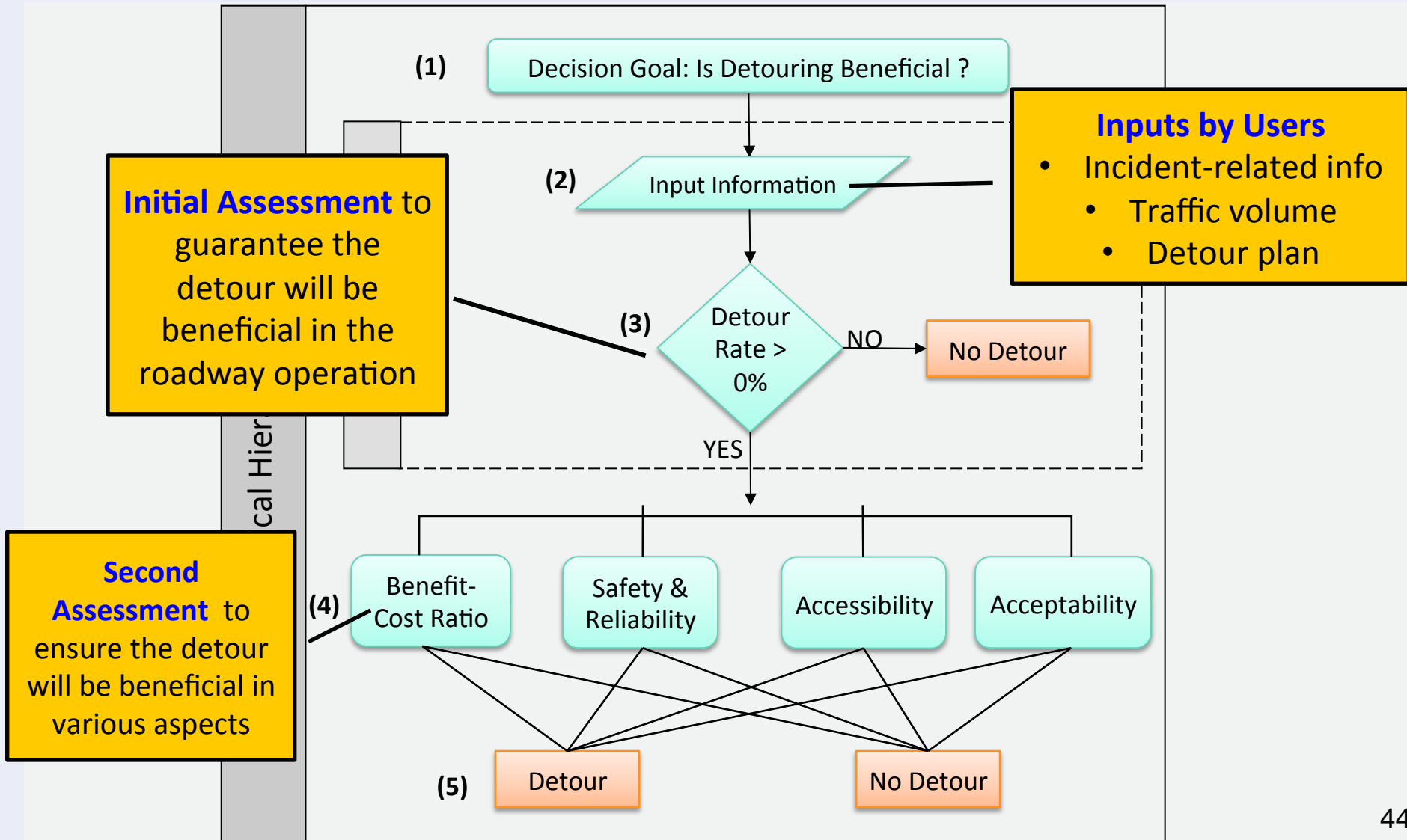
The Proposed System Architecture



Simulation-based Analysis

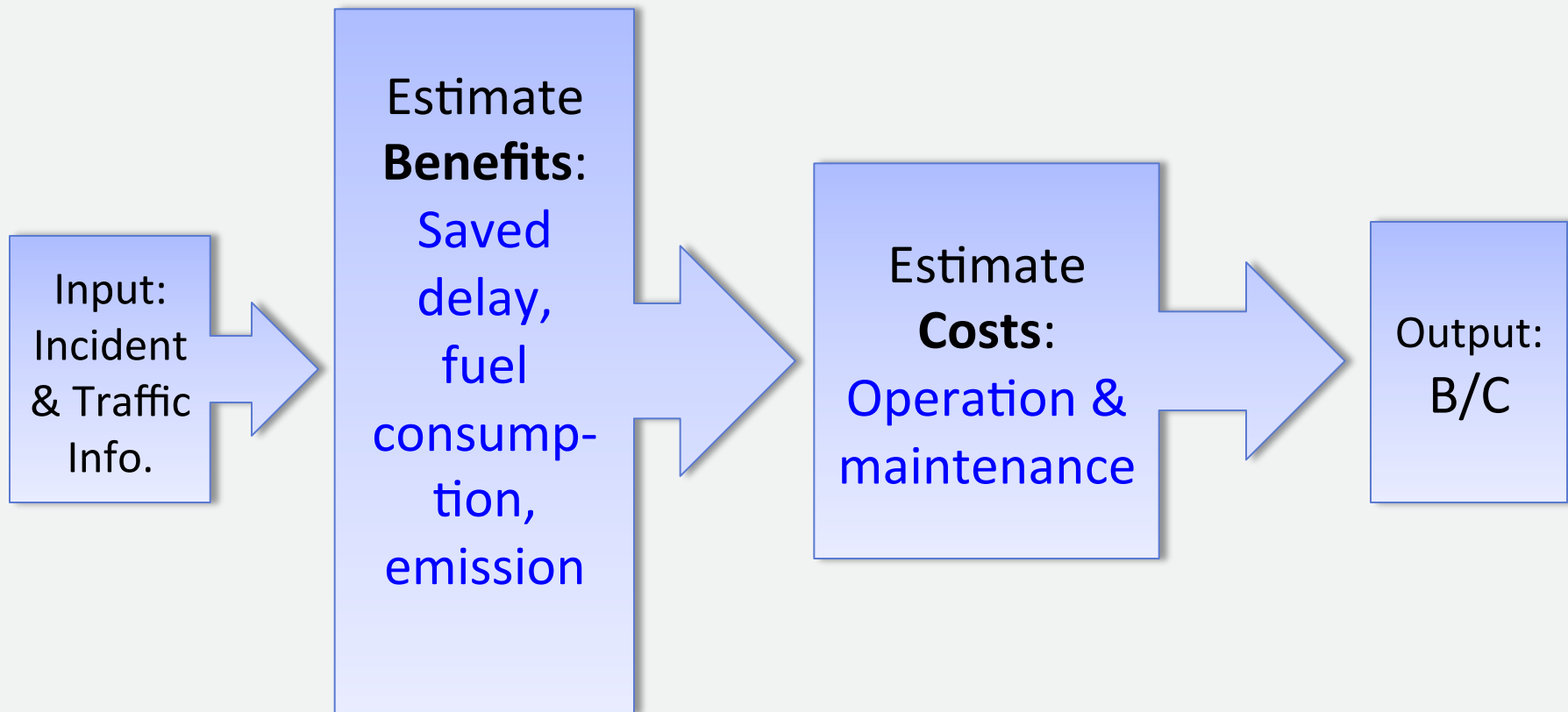
- ❖ **To estimate the optimal diversion rate** from the freeway mainline to mitigate the congestion at the incident segment
 - **Concurrently adjust signal timings** at the arterial intersections to best accommodate the detour traffic
 - **Multi-objective functions**
 - Max total throughput of the freeway corridor
 - Min total time of detour travelers on the detour route
 - **Constraints**
 - Control for signal timing (**min green time**)
 - Control diverging traffic (**max diverging rate**)

The Proposed System Architecture



Decision Criteria on the Second Assessment

❖ Benefit-Cost Ratio



Decision Criteria on the Second Assessment

❖ Safety and Reliability

- Impacted area → reduction in secondary incidents
- Measured by **the max queue length**
 - A multiple linear regression model based on numerous variables regarding incident, location, heavy vehicle volumes, and traffic volumes (Kim et al. 2013)

Decision Criteria on the Second Assessment

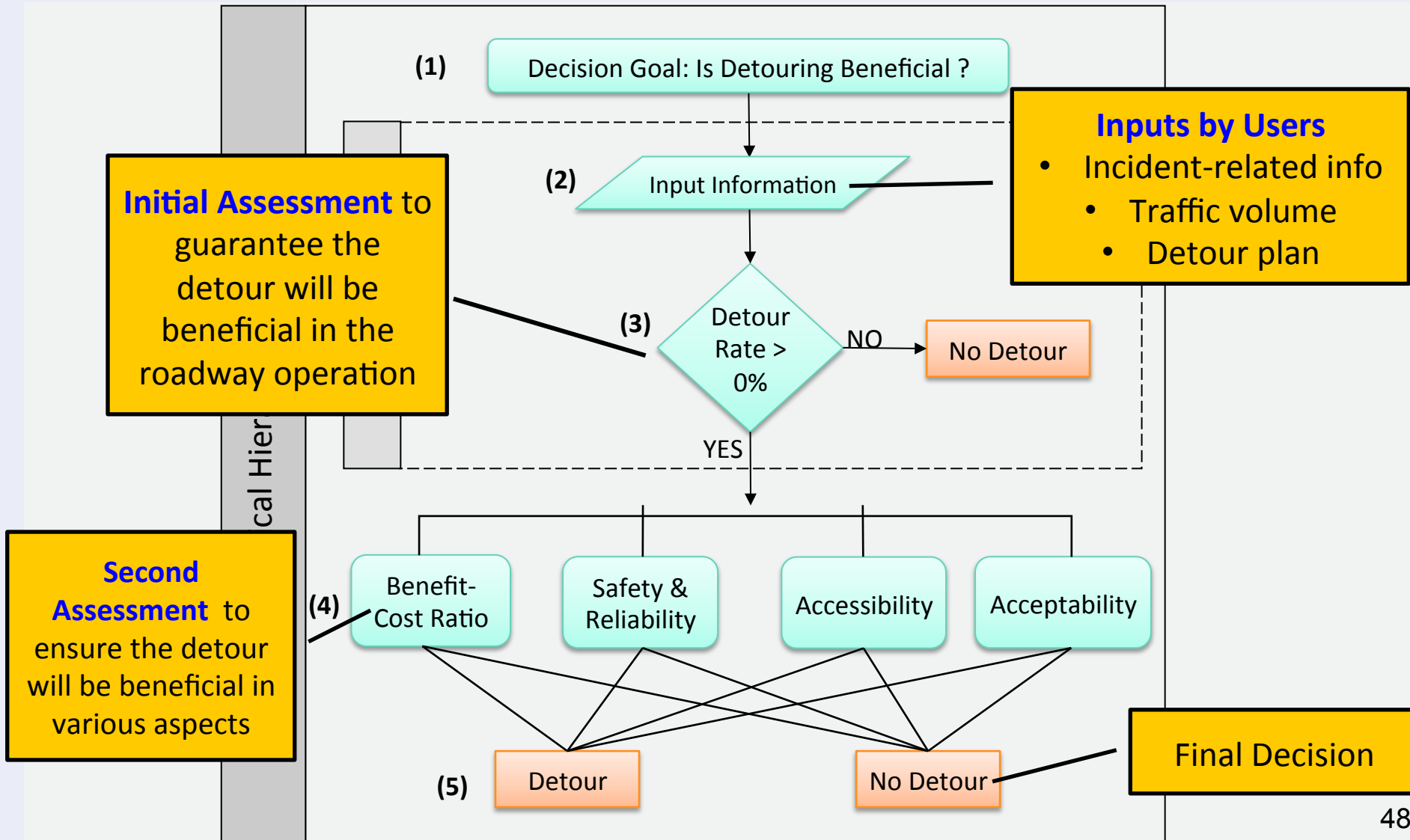
❖ Accessibility

- Traffic signals, stop signs and speed limits on the detour route
- Measured by travel time

❖ Acceptability

- Depend on the characteristics of driving populations and timely supply of the real-time traffic information
- Measured by the anticipated compliance rate (user input)

The Proposed System Architecture



Case Study

Weights for Criteria

- Benefit-cost ratio: 0.31
- Accessibility: 0.18
- Safety and reliability :0.31
- Acceptability: 0.20

Scenario 1

- A Full Road Closure (3/3)
- 60 minute-incident duration
- System Recommendation: **Detour operations are beneficial** (recommended) with 60% confidence.
- # of signals on detour route: 2
- Speed limit on detour route: 50 mph

Scenario 2

- A Full Road Closure (3/3)
- 90 minute-incident duration
- System Recommendation: **Detour operations are NOT beneficial** (recommended) with 62% confidence.
- # of signals on detour route: 5
- Speed limit on detour route: 40 mph

Comparisons of Decisions by Agency

	Scenario No.	1	2
Decision Criteria (used by agencies in the literature)	Lane Blockage (# of closed lane(s)/total # of lanes)	3/3	
	Incident Duration (minutes)	60	
Decisions by Agency	NC DOT-main office	Y	
	NC DOT-Charlotte	Y	
	NJ DOT	Y	
	Oregon DOT	Y	
	NY DOT	Y	
	FL DOT	N	
	ARTIMIS (Ohio/Kentucky)	Y	
	Idaho (Ada County)	Y	
	Wisconsin DOT	Not clear	N
Decision by Proposed System		Y	N

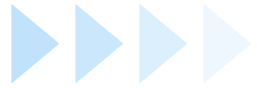
Scenario No.	1	2
optimal detour flow	0.85	0.54
total travel time (hr) w/ detour	3,232	10,163
total travel time (hr) w/o detour	3,617	10,182
saved travel time (hr)	386	19
B/C w/ detour	14.74	0.60
B/C w/o detour	0.07	1.68
max queue w/ detour (mile)	1.37	2.24
max queue w/o detour (mile)	1.66	2.59
travel time (min) via freeway	2.52	2.52
travel time (min) via detour	6.55	7.52

The System Flexibility with Relative Importance

❖ Base scenario

- 15 minutes incident duration with full lane blockage (3/3)

B/C: 0.31 S&R: 0.31 Acces: 0.18 Accep: 0.20	Case A	<ul style="list-style-type: none">• Higher weights on B/C and safety and reliability• Detour operations are recommended with 58% confidence.
B/C: 0.18 S&R: 0.20 Acces: 0.31 Accep: 0.31	Case B	<ul style="list-style-type: none">• Higher weights on accessibility and acceptability• Detour operations are Not recommended with 53% confidence.
B/C: 0.25 S&R: 0.25 Acces: 0.25 Accep: 0.25	Case C	<ul style="list-style-type: none">• Equal weights on all factors• Detour operations are recommended with 53% confidence.



Contributions, Future Research, and Conclusions

Contributions

- ❖ Empirically **investigated the effectiveness** of a well-operated incident response program
 - An efficient response operation can also reduce the incident clearance duration and produce significant benefits.
- ❖ Developed **an efficient model** for optimally allocating the available response units **from a new perspective** of minimizing the total incident-induced delay
 - The developed model's performance and robustness have been confirmed from the extensive numerical results and the comparative study with the existing models and the current practice in Maryland

Contributions (cont'd)

- ❖ Developed an integrated system to provide **a reliable prediction of the clearance duration** for a detected incident.
 - Incident clearance duration is one of the essential parameters for estimating the resulting traffic impacts and assessing the operational efficiency
- ❖ Provided **some insightful information** on the interrelationships between key factors contributed to incident duration and their collective impacts on clearance times
 - Would be useful for traffic agencies to plan and improve their incident management programs.

Contributions (cont'd)

- ❖ Provided operational **guidelines and tools** for responsible agencies to conduct their assessment of traffic diversion plans as well as design of control strategies during the incident management period
- ❖ Developed **an integrated system** that can assess the necessity of traffic detour/diversion **based on the comprehensive review** of associated factors

Future Research

- ❖ Enhancing **reliability** of the incident response management strategy
 - Considering the **likelihood** of having **multiple incidents** over a short time period
 - Taking into account of **the stochastic nature of incident patterns**
 - Investigating the pros and cons **between the dispatching and patrolling** strategies for different times of a day under various traffic conditions and incident patterns
 - Studying **the optimal fleet size** based on the benefit-cost analysis for a given incident distribution, resource constraints, and operational costs

Future Research (cont'd)

- ❖ Enhancing **computational efficiency** for real-time operations of the detour decision support system
 - To supplement or replace simulation- or optimization-based models
 - To generate key traffic control parameters such as **optimal diversion rate** and **reduced total travel time** by detour operations.

Conclusions

- ❖ My field experimental analysis has confirmed the need to contend daily non-recurrent congestion with an efficient and effective incident management program.
 - An efficient incident management needs to optimal use available resources, and best coordinate all responsible agencies.

Conclusions (cont'd)

- ❖ This study **enhanced the efficiency and effectiveness** of the current traffic incident management system in Maryland by developing more reliable models embedded in the system.
 - An incident management system with the proposed key models, incident detection, and detour optimization tools can substantially **reduce the delay, fuel consumption, and emission** caused by incidents.
 - Such a system, if properly integrated with travel time information system, can substantially **improve the quality and efficiency of commuters** over congested highways.



Thank You
Q & A