

Project No. 2406C

<<Collaboration with Government Organizations>> Research on Efficient Accident Prevention Measures Using Artificial Intelligence Project Leader: Akinori Morimoto, Waseda University



1. Research Overview

人工知能を用いた効率的な事故防止対策に関する研究

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1.1 プロジェクトメンバー



PL	Akinori Morimoto	Professor, Faculty of Science and Engineering, Waseda University		
Member	Kazunari Kato	Professor, Faculty of Commerce, Keio University		
	Rumiko Iwasada	Motor Journalist		
	Akihiro Nakamura	Professor, Faculty of Economics, Chuo University		
	Hidekatsu Hamaoka	Professor, Graduate School of Science and Technology, Akita University		
	Yuka Nakagawa	Professor, Faculty of Law, Chukyo University Lawyer, Nakagawa Law & Management Office		
	Nobuaki Takubo	Deputy Director, Research Department, Institute for Traffic Accident Research and Data Analysis (ITARDA)		
Special Observer	Daisuke Kamiya	Associate Professor, Department of Engineering, Faculty of Engineering, University of the Ryukyus		
	Kyoko Manaka	Associate Professor, Faculty of Economics, Ryutsu Keizai University		
	Yoshimi Furukawa	Advisor, International Association of Traffic and Safety Sciences (IATSS) Professor, Professional University of Automotive Engineering		
	Jun Teraoku	Chief Engineer, Road Traffic Department, Chubu Branch, Construction Engineering Research Institute Co., Ltd.		
	Toyoki Kurihara	1st Year Master's Student, Graduate School of Waseda University		

<Observers>

• Takashi Koizumi (Traffic Guidance Division, Traffic Bureau, National Police Agency), Katsuki Masuda (Traffic General Affairs Division, Traffic Department, Tokyo Metropolitan Police Department)

- •Nozomu Watanabe, Daiken Suzuki (Road Bureau, Ministry of Land, Infrastructure, Transport and Tourism)
- •Atsunori Sugiura, Takuya Aso (Sales Department, Informatix Inc.)
- •Emi Satake, Sion Honda (Police Security Solution Division 2, NEC Corporation)
- Hironori Kimura, Yuki Hara (Construction Engineering Research Institute Co., Ltd.)

As of March 2025

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- The 11th Basic Traffic Safety Plan identifies "promoting traffic guidance and enforcement to help prevent traffic accidents" as a priority measure.
- The Digital Agency was established in 2021, necessitating standardization and other responses in the field of traffic safety as well.
- IATSS has been publishing the "Traffic Enforcement Handbook" since 2014, providing continuous information to those involved in traffic enforcement.



development of general-purpose applications, and consider regionally based countermeasures by accumulating countermeasure effects.

Develop a basic model that utilizes rapidly advancing

artificial intelligence to advance accident prevention

measures.















2. Traffic Guidance and Enforcement Activity Support System: General Public Release

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2.1 Overview of the Traffic Guidance and Enforcement Activity Support



- A system combining a "traffic accident risk prediction model" and an "enforcement effectiveness evaluation model."
- Supports the determination of enforcement locations by visualizing the model calculation results in a series.





2.2 General Public Release of the Traffic Guidance and Enforcement Activity Support System

ATS

- The "Traffic Guidance and Enforcement Activity Support System" developed through research up to the previous year has been made publicly available on the International Association of Traffic and Safety Sciences website.
- One organization has applied to have the system data sent to them, and that was completed.

<IATSS HP Top Page>







このシステムは、近年急速に活用が進む人口知能(AI)を活用し、効率的な事故抑止対策箇所を提案するものです。システムの 送付を希望される方は下記にアクセスしてください。 送付のお申込みはこちら



A paper on the theoretical aspects of the "Traffic Guidance and Enforcement Activity Support System" was published in IATSS Research (2024).

IATSS Research 48 (2024) 129-135



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Keywords: Traffic safety Traffic enforcement PDCA Traffic enforcement is one of the 3E's of traffic safety measures. Traffic enforcement has been reported to reduce traffic accidents worldwide; however, significant regional differences exist in its effectiveness. This study summarizes the history of the effects of traffic enforcement on reducing traffic accidents in Japan. In particular, scientific approaches and more efficient enforcement since 2000 have been describ

Based on the results of past research on efficient traffic enforcement using AI, a research project on efficient accident prevention measures using AI was started in 2022, with a view toward future practical applications. The goal is to develop a hybrid model that combines a forecasting model based on knowledge gained from conventional statistical methods with an AI-based sequential update model using big data (see Fig. 11).





3. Prefectural Police Hearings

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3.1 Hearing Implementation Status





3.2 Hearing Results (Main Opinions and ICT)

accidents, the lowest in history, and the number of traffic

They have an accident risk AI system

at the headquarters and will start full-scale operation soon





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They analyze accident congestion using probe data

Kanagawa Police

When enforcement numbers were reduced from 2016 to 2018, the number of fatalities increased

They utilize an AI system for traffic monitoring patrol route selection and selection of accident-prone areas

(C)小学生の学習教材【ちびむすドリル】 http://kotoba.littlestar.jp/syogaku.

3.3 システムの導入可能性





Traffic Guidance and Enforcement Activity Support System



No.	Target	Main Opinion		
1	Police A	 Predicting and visualizing the effects of street activities could lead to more focused activities 		
		 The jurisdiction area is large, and patrol cars also patrol, so it's difficult to conduct street activities at system-proposed locations 		
2	Police B	•While there is a system within the prefectural police, its usability depends on individual abilities. A system that supports this would be useful		
		 It's unclear whether daily system input of street activity records is feasible 		
3	Police C	 Proposing enforcement activity locations and times alone is innovative and would be helpful if it reduces the effort required for enforcement activity planning 		
		 It might be difficult for local police station personnel to handle daily data input 		
4	Police D	 Understanding accident risk is effective and can be utilized for future enforcement 		
		 It would be better if it could propose risk details (rear-end collisions, intersection accidents, etc.) and activity methods 		
5	Police E	•There is a sense of routine in selecting activity locations, which could be resolved		
		 Adding location characteristics (sunrise, road surface freezing, etc.) would allow consideration of regional characteristics 		



4. Consideration of System Generalization Based on Hearings

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PRC*=Predicted Risk Value

4.2 Accident Occurrence Risk Prediction Model

ATSS



4.3 Examination Method

ATS

Conducted an examination of two publicly available datasets

Automotive Probe Data: (Traffic Vision Probe Data)

• Sales of statistically processed Honda probe data.

•Target data: Honda vehicle travel data excluding commercial and freight vehicles.

Available data

①Trave time-related

• Link travel time data: Calculates trave time for each road segment from trave data/ Targets roads with a width of 5.5m or more.

②Sudden deceleration-related

•Sudden deceleration occurrence point data (deceleration, occurrence date and time, latitude and longitude)



Mobile Probe Data (Human Flow Data) (Profile Passport)

• Targets users who have granted permission to use location information from partnered smartphone applications (approx. 30 million people).

Acquires user location information at 5-15 minute intervals.

Available data

1 Point data \rightarrow movement mean and speed data can also be attached

·data in latitude and longitude unit for each user

- ②Non-aggregated OD data
 - ·Data organized by origin and destination for each user





4.3 Examination Method



•Compare "traffic volume (data volume)," "travel speed," and "number of sudden deceleration occurrences" for each analysis mesh in each data set to verify the possibility of substitution.

• The measurement units for each data set are as follows:

Product Name	ETC2.0 Probe Data	Traffic Vision Probe Data	Profile Passport
Data type	Point	Road link	ポイント
traffic volume measurement unit	Every 200m when changing direction by 45°or more	For each DRM road link 5-minute intervals	Every 125 meters (anonymized within a 125m mesh)
Travel speed measurement unit	Every 200 when changing direction by 45°or more	Road link length ÷ average travel time	Every 125 meters (anonymized within a 125m mesh)
Sudden deceleration count measurement unit	Every 200 meters When changing direction by 45° or more When abrupt maneuvers occur	When maneuvers of -0.25G or more occur (point data)	No data on abrupt maneuvers
Remarks	Scaling process based on the installation rate of ETC2.0 onboard units		Used only for data with "car" as the means of transportation.

4.3 Examination Method



- For each data set, we will compare 'traffic volume (data volume)', 'travel speed', and 'number of sudden decelerations' for each analysis mesh to verify the possibility of substitution.
- The calculation method for each data set is as follows

Product Name	ETC2.0 Probe Data	Traffic Vision Probe Data	Profile Passport
Traffic Volume Aggregation Method	Interpolate data with spline curves and aggregate data volume for each analysis mesh	Traffic volume of road links overlapping each analysis mesh. If multiple road links apply, sum the traffic volume of all links	Aggregate data volume for each analysis mesh
Travel Speed Aggregation Method	Interpolate data with spline curves and aggregate the maximum data value for each analysis mesh	Maximum travel speed of road links overlapping each analysis mesh	Aggregate the maximum data value for each analysis mesh
Number of Rapid Decelerations Aggregation Method	Aggregate the number of maneuvers of -0.25G or more for each analysis mesh (excluding -1.00G or more)	Aggregate the number of maneuvers of -0.25G or more for each analysis mesh (excluding -1.00G or more)	
Remarks	Scaling process based on the installation rate of ETC2.0 onboard units	_	Used only for data with "car" as the means of transportation



•Area where trial operation was conducted (Shinjuku Ward, Tokyo, 50m mesh x 2064 units).



4.5 Examination Results (Traffic Volume)

(1) Traffic Vision Probe Data

Data plotted on roads in both cases
Traffic Vision has a slightly higher proportion of arterial roads
Relatively strong correlation (Correlation coefficient r = 0.647)







4.5 Examination Results (Traffic Volume)

(2) Profile Passport

Profile Passport has plots on arterial roads, but a certain amount of plots outside roads
Profile Passport has a high proportion of Shinjuku Station and its surrounding areas
Weak correlation

(correlation coefficient: 0.116)





4.5 Examination Results (Travel Speed)

(1) Traffic Vision Probe Data

Data plotted on roads in both cases.

•Traffic Vision has a slightly higher proportion of arterial roads.

• ETC2.0 probe has more detailed changes.

Strong correlation

(correlation coefficient: 0.811).

▼ETC2.0 probe data (1 day)







4.5 Examination Results (Travel Speed)

(2) Profile Passport

- Profile Passport has plots on arterial roads, but a certain amount of plots outside roads.
- Profile Passport has high travel speeds on railway lines. Possible misidentification of "train" as "car."
- Weak correlation

(correlation coefficient: 0.444).

▼ETC2.0 Probe Data(1 day)







4.5 Examination Results (Sudden Deceleration Occurrence Count)



(1) Traffic Vision Probe Data

- •Data plotted on roads in both cases.
- •Traffic Vision has a slightly higher proportion of arterial roads.
- •ETC2.0 includes data from narrow streets and building parking lots.
- Strong correlation
- (correlation coefficient: 0.720).

▼ETC2.0 Probe Data (30 days)







5. Revision of the Traffic Enforcement Handbook

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5.2 Revised Sections and Content



Chapter	Title	Revision Content
Chapter 1	Characteristics of Traffic Enforcement	No change
Chapter 2	Relationship Between Traffic Enforcement and Traffic Accidents	No change
Chapter 3	Regional Differences in Traffic Enforcement	No change
Chapter 4	Characteristics of Traffic Violations	No change
Chapter 5	Impact of Traffic Enforcement on Driver Awareness	Updated content of "Safety Awareness by Showing Traffic Street Activities"
Chapter 6	Towards Traffic Enforcement that Suppresses Accidents	Newly added the following two items: • Classification of traffic accidents based on differences in enforcement effectiveness • Differences in enforcement effectiveness by accident type
Chapter 7	Efficient Accident Prevention Measures Using ICT	 Based on the results of this research, the following seven items are included: What is artificial intelligence? Transition of traffic safety research using artificial intelligence Example 1 of traffic accident prediction model using artificial intelligence Example 2 of traffic accident prediction model using artificial intelligence Model constructed by the International Association of Traffic and Safety Sciences Overview of the traffic enforcement activity support system Introduction of how to use the traffic enforcement activity support system

5.3 Revision Details





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5.3 Revision Details



(2) Updated Content of Chapter 5



While it can be observed that showing traffic street activities in this way has the effect of suppressing the severity of traffic accidents, the decline in fatal accidents has shown a marked tendency to bottom out in recent years. Even looking at the figures from the Kanagawa Prefectural Police (below), the number of traffic accident fatalities, which had been on a downward trend, reached 162 in 2018, and apart from the recent COVID-19 pandemic, the number of fatal accidents shows a tendency to bottom out. This suggests that the effect of 'showing activities' is difficult to continue to raise drivers' traffic safety awareness without a certain level of arrests. The Kanagawa Prefectural Police has also adopted a policy of aiming to reduce traffic accidents while continuing to exert the effect of showing traffic street activities and giving consideration to the accident prevention effect of arrests in recent years.

From the above, it can be seen that not only arrests and direct guidance of drivers who commit traffic violations but also traffic street activities focused on warning can raise drivers' awareness of safe driving and be an effective traffic safety measure. A measure that combines direct guidance and traffic street activities in a balanced manner would be effective.

5.3 Revision Details

(3) Added Content of Chapter 6 (Example)







(1) Effective Traffic Safety Measures





交通事故抑止対策の高度化、合理化につながる

交通女全対策については、事故、違反、規制、街頭記動、民間が保有する情報等のデータを 組み合わせた GSを活用し、事故発生生所と取為り箇所の位置関係、事故の内容等を警察徴 員が分析し、交通配置や交通取論り、あるいは道路標識などの交通規制見直し等を行っている のが現状である。

そこで、AIを注用した分析により、様々なタイミングで交通事故の発生を抑止するための 必要な対策を予測し、提示するシステムがあれば、限られた人員を必要な時間帯に必要な場所 に効果的に配置することができ、交通事故の発生を抑止できるほか、これまで個々の警察職員 が時間をかけて分析していた業務の省力化にもつながるものと思われる。

22章、四国事業・皆有事気の予測にはたんに1.5月10年6万に到すべち落ち研究会、発揮、西国事業・皆有事気の予測におけんに1.5 用の作り方に関する緑言書を26, 2018

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(2) Overview of the IATSS Model



[Step 1] Input area and date/time for which you want to conduct enforcement activities •Input the jurisdictional area (local police station unit) and date/time for which you want to conduct enforcement activities.



[Step 2] Display current traffic accident risk on a map.

•Display the current traffic accident risk in the jurisdictional area and date/time input in Step 1 on the screen in a map and check the locations with high risk.





[Step 3] Input activity conditions for the enforcement activity day.

•Input the activity conditions for the day of enforcement activity (number of activity hours, number of street activity locations, precipitation, snowfall, sunshine hours, maximum wind speed) with reference to the weather forecast.

* Necessary information can be obtained from online weather forecast information (Japan Weather Association website, etc.)



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[Step 4] The system proposes optimal enforcement activity locations.

•Based on the activity conditions in Step 3, have the system calculate the traffic accident risk when enforcement activities are conducted. Furthermore, based on the calculation results, have the system propose the optimal enforcement activity locations.





[Step 5] Calculate and map the enforcement effect when conducted at system-proposed locations.

•Based on the current traffic accident risk Step 2, calculate the reduction effect of traffic accident risk when enforcement activities are conducted at the enforcement activity locations proposed by the system and display it on a map.



[Step 6] Set the planned locations for enforcement activities.

•With reference to the enforcement activity locations proposed by the system, set the planned locations for conducting enforcement activities.





[Step 7] Calculate and map the enforcement effect when conducted at planned locations. •Based on the current traffic accident risk Step 2, calculate the reduction effect of traffic accident risk when enforcement activities are conducted at the planned enforcement locations and display it on a map. Here, the risk reduction effects of multiple candidate cases can be compared. Based on these results, finally determine the enforcement activity locations.



[Step 8] After conducting enforcement activities, register the actual locations and times of implementation.

•After actually conducting enforcement activities, register the locations and times of implementation in the system.





6. Summary of Research

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6. Summary of Research



Based on the results of trial operation of the constructed system and hearing surveys, the conditions were verified.

• It was found that there are differences in the data held and available data among the prefectural police departments.

• In the future, it is expected that the system will be deployed nationwide through improvements to the evaluation model, improvements in system versatility, and accumulation of data.



