

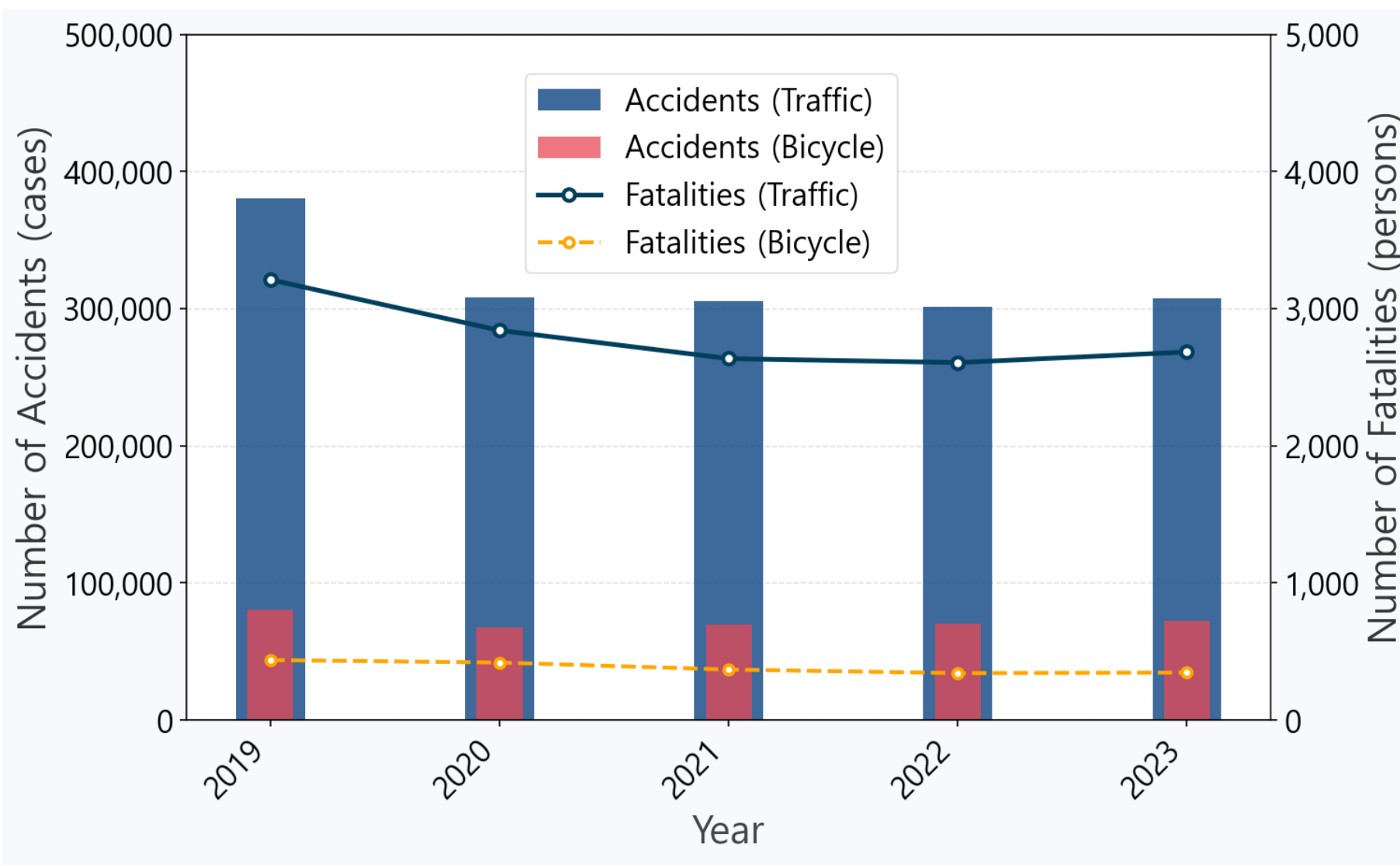
Traffic and bicycle accidents in Kyoto: Impact of POIs, cornercuts and road design



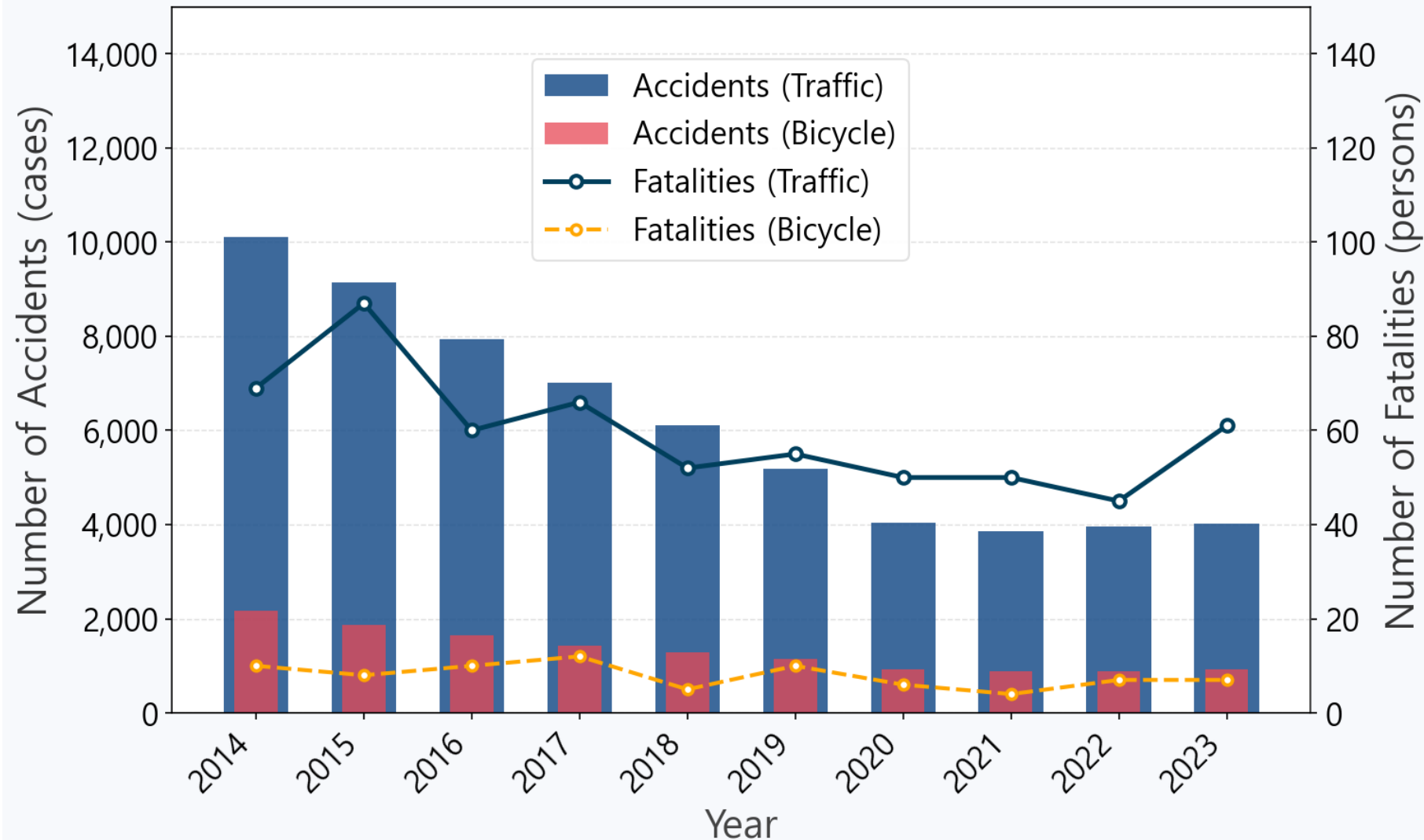
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Traffic and bicycle accidents in Japan

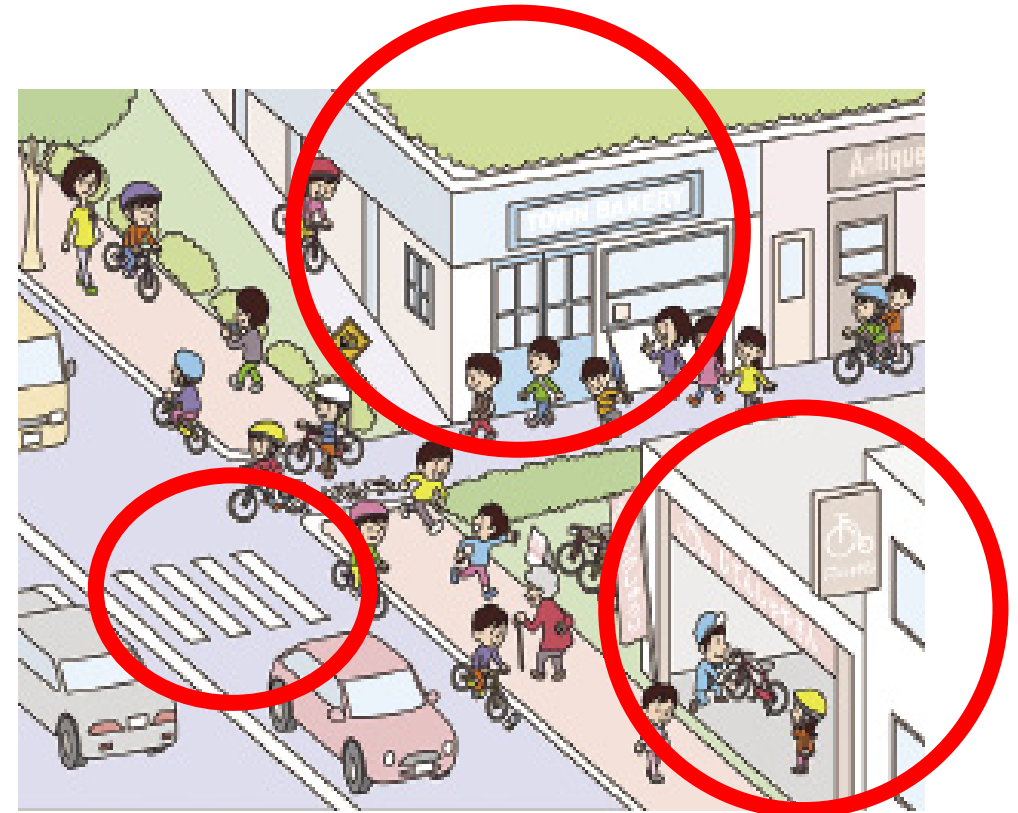


Traffic and bicycle accidents in Kyoto



Study 1: POIs and accidents

- Shops such as convenience stores are often located directly at junctions.
- Do such POIs contribute to accident risk controlling for other factors, including traffic volume and land-use
 - Causality is difficult to establish!
- We focus specifically on smaller junctions



Study area and data used

Target Intersections

- 26,553 intersections located in the densely populated area of Kyoto City
- Among them, 1,124 intersections with available traffic volume data ($\geq 10,000$ vehicles/day)

Target Traffic Accidents

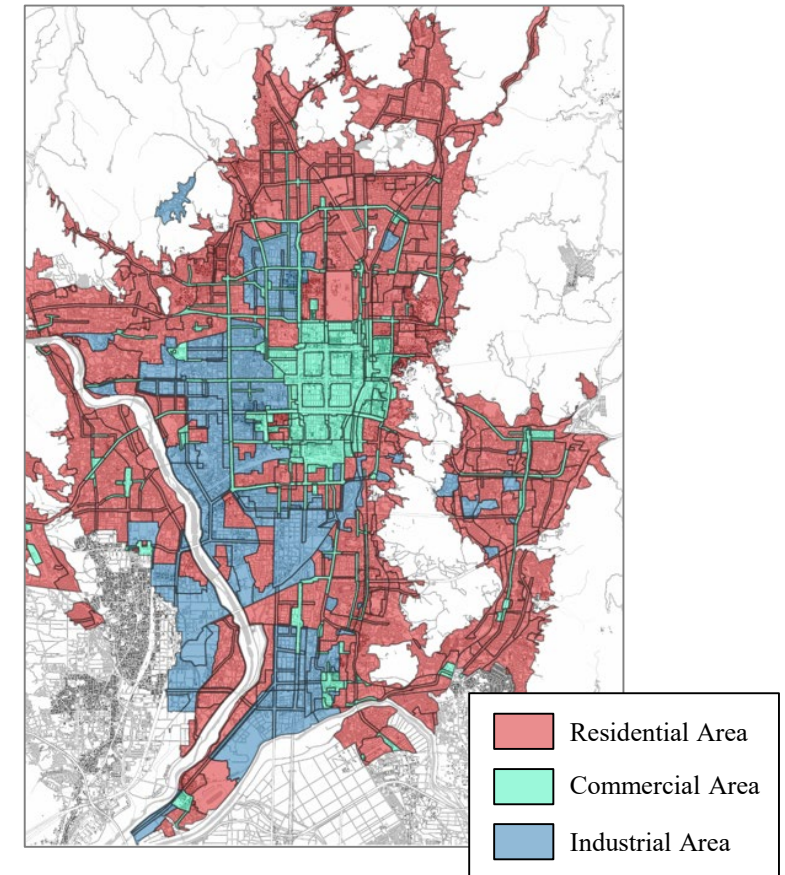
- 8,124 injury accidents in Kyoto City from 2017 to 2019

Spatial Data

- Building footprint data from OpenStreetMap
- Land-use data from the National Land Numerical Information (NLNI) dataset

Traffic Volume Data (used for a subset of intersections)

- 5-minute interval traffic counts at 331 locations in Kyoto City (2017–2019), published by JARTIC



Data preparation

Road network data obtained from OSM to construct intersection-level data



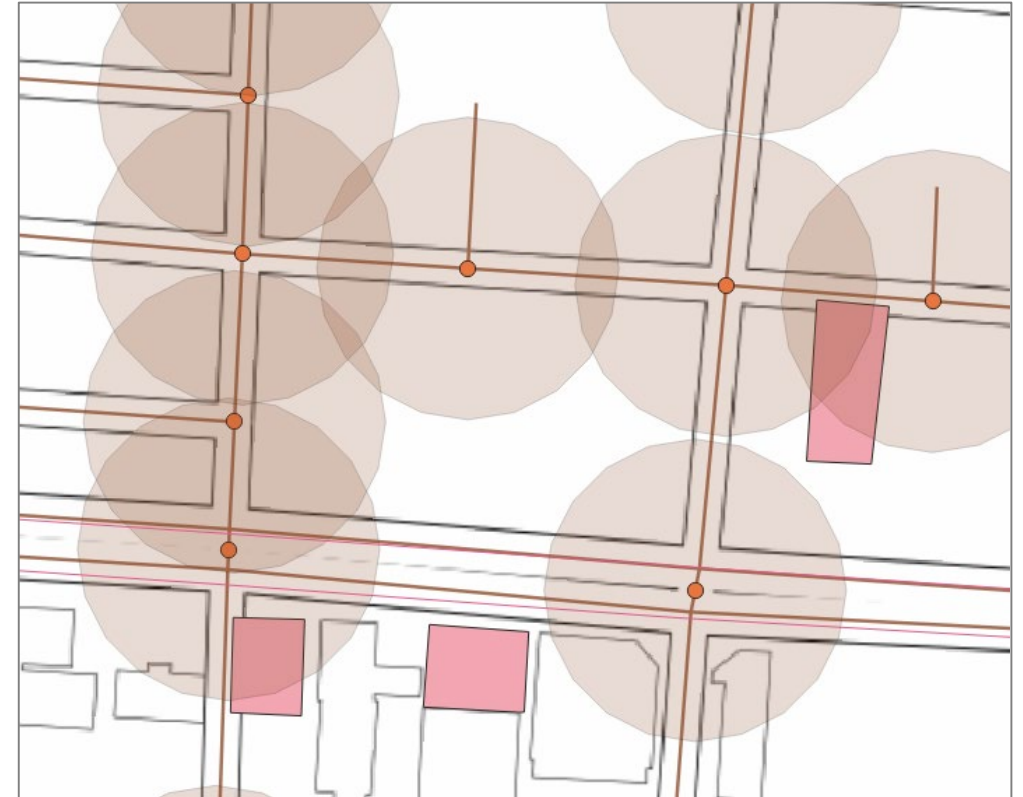
The presence of traffic signals and crosswalks within a 10m radius was identified



The presence of surrounding facilities within a 30m radius was identified



Land-use type and population density data were merged

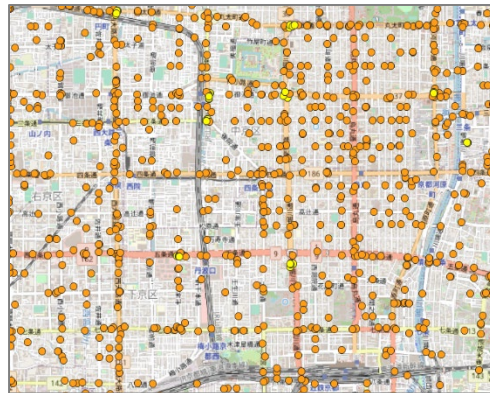


A 30m buffer around intersections and building footprint data (e.g., parking lots)

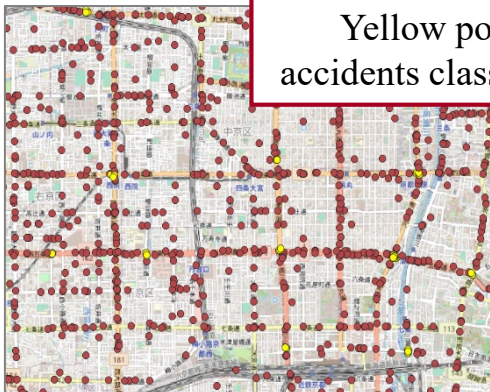
Accident hotspots



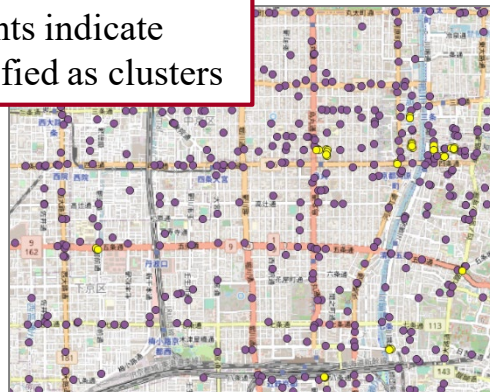
Car (30 m, 10 points)



Bicycle (30 m, 5 points)



Motorcycle (30 m, 5 points)



Pedestrian (30 m, 3 points)

Yellow points indicate accidents classified as clusters

DBSCAN is applied to all traffic accident data. The results confirm that traffic accidents are concentrated around intersections.

- Cars and motorcycles show similar spatial patterns, with clusters distributed but concentrated at major intersections.
- Bicycle accidents are likely concentrated in areas with high population density.
- Pedestrian accidents are clustered around central commercial districts, particularly near Shijo Street.

Junction examples



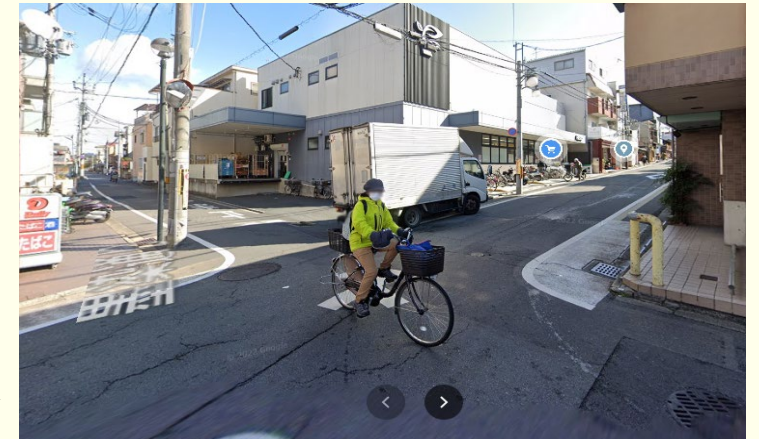
Small Intersection, with Restaurants, High Accident Frequency

- Located in a busy commercial district with high pedestrian activity
- Presence of an approach road with limited visibility
- Adjacent to Kawaramachi Street (narrow sidewalks), likely associated with high bicycle traffic



Small Intersection, with Supermarket, High Accident Frequency

- Narrow roadway with little or no sidewalk space
- A parking lot located near the intersection, likely generating frequent vehicle, motorcycle, and bicycle movements



* Source: Google Maps Street View

Junction examples (2)



Medium Intersection, with Convenience Store, High Accident Frequency

5 accidents over 3 years

- Car × Motorcycle: 2
- Car × Pedestrian: 2
- Motorcycle × Bicycle: 1

High traffic volume, limited visibility



Large Intersection, with Park, No Accidents

- Wide sidewalk corners and good visibility.
- Areas with parks tend to have more available space, which may allow for wider sidewalks and improved sight distance.

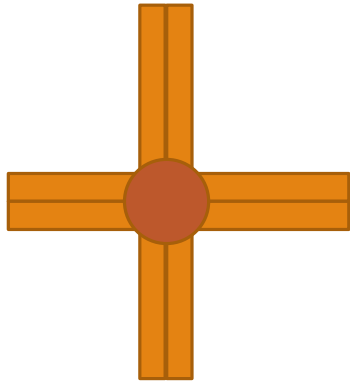


* Source: Google Maps Street View

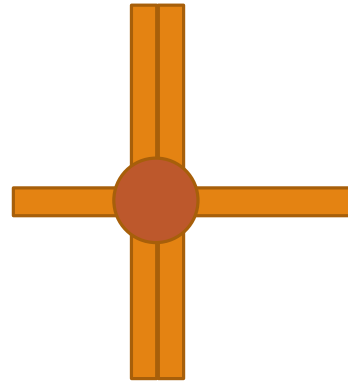
Intersection types

Intersections are classified into three types:

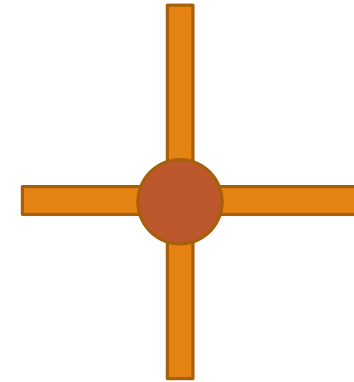
- Large Intersection: two or more lanes on both approaches
- Medium Intersection: two or more lanes and a single-lane road
- Small Intersection: single-lane roads



Large



Medium



Small

Regression analysis: Junction types

Variable	Large Intersections	Medium Intersections	Small Intersections
Constant	-0.3689 †	-1.1694 **	-3.0962 **
Signal	1.969 *	1.2159 **	2.2278 **
Crosswalk	0.3723 †	0.4931 **	0.5717 **
Population Density	-1.66×10^{-05}	-3.32×10^{-06}	6.21×10^{-06}
Commercial Area	0.9002 **	1.0991 **	1.1854 **
Industrial Area	0.5887 *	0.6451 **	0.5417 *
Residential Area	-0.7445 *	0.1139	0.185
Park	-0.7409 *	-0.4133 **	-0.0295
Restaurant	0.4433	0.2358 *	0.534 *
Supermarket	0.224	0.5695 **	1.0836 **
Convenience Store	-0.1936	0.8911 **	1.0459 **
Bus Stop	0.0462	0.3592 **	0.0106
Station	1.1334	0.342	-0.6026
Parking Lot	-0.1504	0.2171 **	0.2802 *
Number of Target Intersections	762	5916	19875
Number of Accident Intersections	524	2383	1282
Adjusted R-squared	0.2125	0.1119	0.0419

Signals and crosswalks are significantly positive.

→ this may reflect the influence of higher pedestrian and vehicle volumes at these locations.

The likelihood of traffic accidents across land-use types appears to follow the order: commercial areas, industrial areas, and then residential areas.

→ commercial areas tend to have more pedestrians, whereas industrial areas tend to have more vehicles.

† $p < 0.10$, * $p < 0.05$, ** $p < 0.01$

Pedestrian–Vehicle / Vehicle–Vehicle Accidents

Variable	Pedestrian–Vehicle	Vehicle–Vehicle
Constant	-4.8868 **	-3.2323 **
Signal	2.1531 **	1.9992 **
Crosswalk	0.8779 **	0.3329
Population Density	1.03E-06	4.76E-06
Commercial Area	1.4423 **	1.1387 **
Industrial Area	0.5119 **	0.5426 **
Park	0.268	-0.1186
Restaurant	1.0397 **	0.3472 *
Supermarket	1.7297 **	0.8111 **
Convenience Store	0.9734 **	0.9845 **
Bus Stop	-0.5569	0.1736
Parking Lot	0.2289	0.2588 **
Number of Target Intersections	19875	19875
Number of Accident Intersections	250	1067
Adjusted R-squared	0.07144	0.0329

† p < 0.10, * p < 0.05, ** p < 0.01

Restaurants, supermarkets, and convenience stores show positive and significant effects for both accident types; however, the magnitudes are substantially larger for pedestrian–vehicle accidents.

These facilities are more strongly associated with conflicts involving pedestrians, likely due to increased pedestrian activity and mixed traffic movements near the intersection.

Parking lots also show a significantly positive effect.

This likely relates to frequent vehicle access movements near intersections.

Study 2: Further analysis with PLATEAU data

Project PLATEAU | The Initiative of Digital Twin in Japan



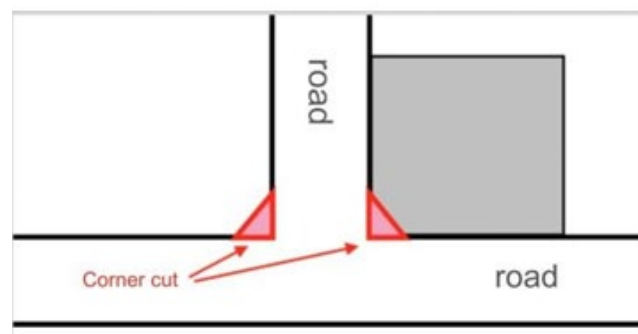
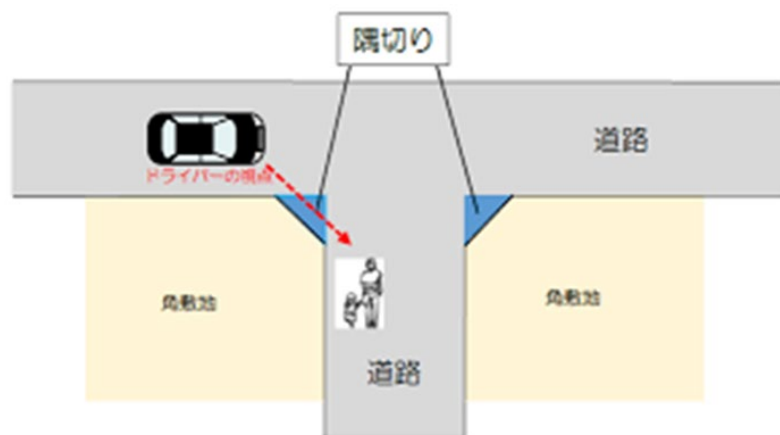
“Launched in 2020, PLATEAU continues to grow as a platform for the participation of many local governments, private companies, and diverse researchers, engineers, and creators.

PLATEAU is a platform for urban management, consolidation of urban functions, and sustainable urban development.

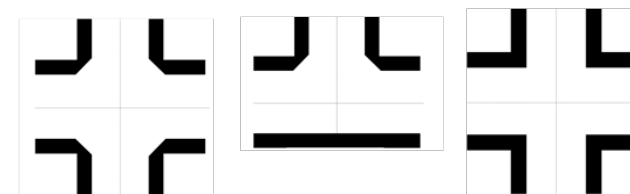
The digital transformation of city planning based on PLATEAU will solve social and regional issues.”

Field of view quantified by “corner cuts”

The number of “corner cuts” was calculated according to the number of vertices at the intersection.



Example of intersection geometry and number of corner-cuts



4 cuts

2cuts

0cuts

Corner-cut : Cut corners facing the street to increase visibility at intersections.

Number of vertices	Number of corner cuts
8 or more	4
7	3
6	2
5	1
4 or less	0

Junction examples



Kawabata Street

Number of corner cuts: 3
(17 accidents in 5 years)



Ayanokoji

Number of corner cuts: 0
(2 accidents in 5 years)

* Source: Google Maps Street View

Regression analysis

	Coefficient	t-value	
constant term	-0.112	-19.876	**
Intersection area	0.005	156.416	**
Numer of corner cuts	-0.085	-23.445	**
Convenience store dummy	0.208	5.934	**
Supermarket dummy	0.079	1.321	
Park dummy	-0.100	-6.114	**
Restaurant dummy	0.183	9.745	**
Station dummy	0.301	4.030	**
school dummy	0.159	1.221	
Kindergarten dummy	-0.095	-1.999	*
Bus stop dummy	0.025	1.317	
Parking lot dummy	0.069	6.700	**
Traffic light dummy	0.572	20.721	**
Crosswalk dummy	0.227	10.173	**
Adjusted coefficient of determination		0.533	

** 1% significant * 5% significant † 10% significant

- Corner cuts significant with expected sign - good proxy for visibility?
- Other parameters stay larger unchanged
- Note: Intersection area as surrogate variable for flow

Ongoing research: Focus on bicycle accidents

Public concern and policy response



Media reports highlighting unsafe and improper bicycle riding in urban areas

令和6年11月1日 道路交通法の改正
自転車の危険な運転に新しく罰則が整備されました

運転中ながらスマホ



スマートフォンなどを手で保持して、自転車に乗りながら通話する行為、画面を注視する行為が新たに禁止され、罰則の対象となりました。
※停止中の操作は対象外

違反者は、**6月以下の懲役又は10万円以下の罰金**
交通の危険を生じさせた場合、**1年以下の懲役又は30万円以下の罰金**

酒気帯び運転および酒類の提供



自転車の酒気帯び運転のほか、酒類の提供や同乗・自転車の提供に対して新たに罰則が整備されました。

違反者は、**3年以下の懲役又は50万円以下の罰金**
自転車の提供者は、**3年以下の懲役又は50万円以下の罰金**
酒類の提供者・同乗者は、**2年以下の懲役又は30万円以下の罰金**

「運転中ながらスマホ」、「酒気帯び運転」は自転車運転者講習制度の対象となります。

自転車運転者講習制度

自転車の運転に関し、交通の危険を生じさせるおそれのある一定の違反(危険行為)を反復して行った者は講習制度の対象となります。 ※受講命令違反 5万円以下の罰金

危険行為 信号無視、指定場所一時不停止、道断踏切立入り、安全運転義務違反、通行区分違反 など

重大事故を防ぐため、交通ルールを遵守しましょう。

Recent regulatory amendments and enhanced penalty systems for bicycle traffic

令和8年4月1日より 交通反則通告制度
自転車の違反に**青切符**が導入

青切符(交通反則通告制度)とは...

運転者が比較的に軽微な交通違反をした場合、一定期間内に**反則金を納めると、刑事手続を受けずに事件が処理される制度。**

例

- 携帯電話使用等(保持) 12,000円
- 信号無視 6,000円
- 一時不停止 5,000円
- 酒類の提供等(保持) 12,000円
- 車道の右側通行 6,000円
- 公安委員会遵守事項違反 5,000円

【対象】**16歳以上** 118の反則行為

違反手続の流れ

```

    graph TD
      A[軽微な違反  
(交通反則通告制度対象)] --> B[青切符を交付]
      A --> C[悪質な違反  
(交通反則通告制度の対象外)] --> D[赤切符を交付]
      B --> E{8日以内に  
反則金を納付}
      E -- する --> F[11日以内に  
反則金を納付]
      E -- しない --> G[交通反則通告  
センターに出席し、  
反則金納付の通告  
を受ける]
      F -- する --> H[手続終了]
      G -- する --> H
      G -- しない --> I[刑事手続へ  
有罪なら前科  
(拘禁刑・罰金等)]
      
```

一定の危険な行為を**3年以内に2回以上**繰り返すと自転車運転者講習の受講が命ぜられます(受講しない場合は、5万円以下の罰金)。

「飲酒運転」や「妨害運転」等、特に悪質な違反行為は、交通反則通告制度の対象外のため従来通り**赤切符**を受け、刑事手続となります。

京都府警察

“Bike lanes” in Kyoto



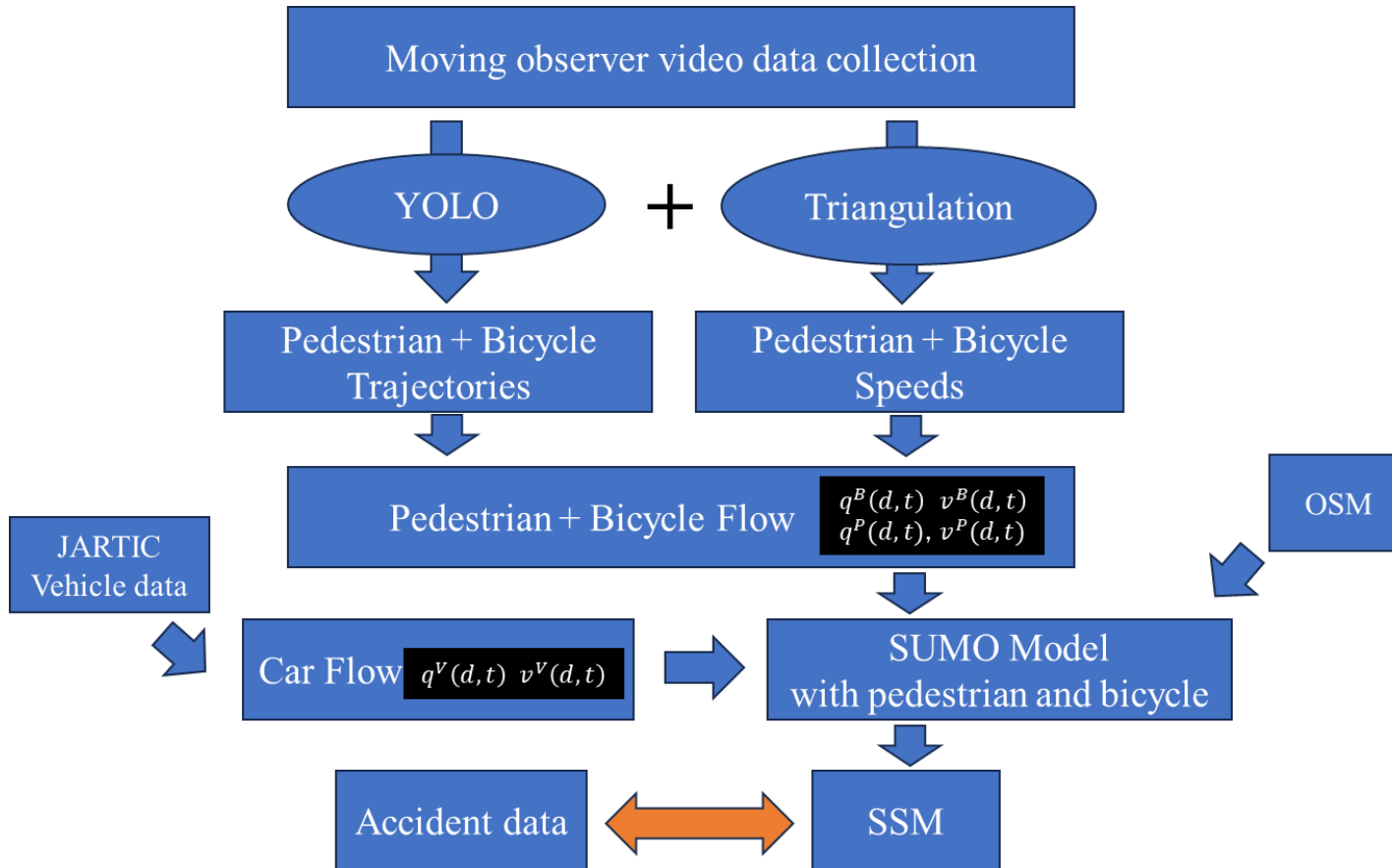
Public concern and policy response (2)



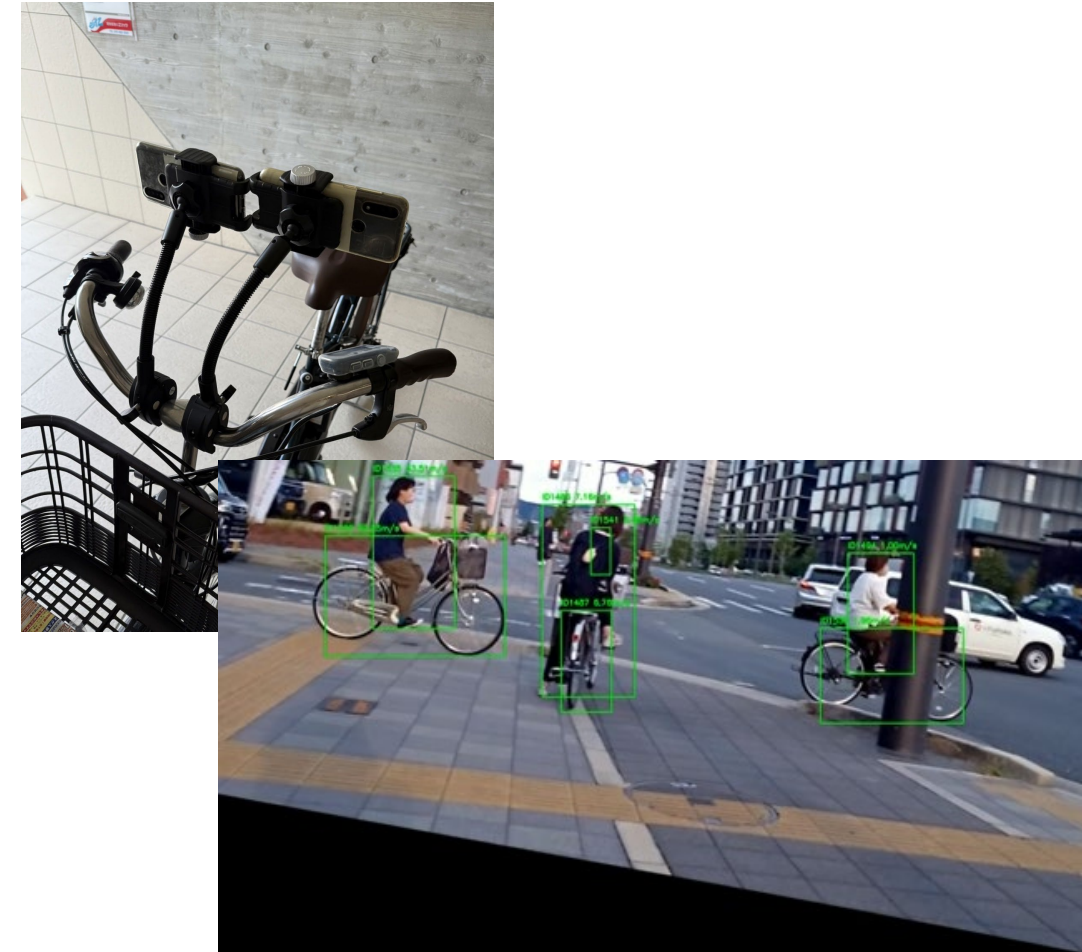
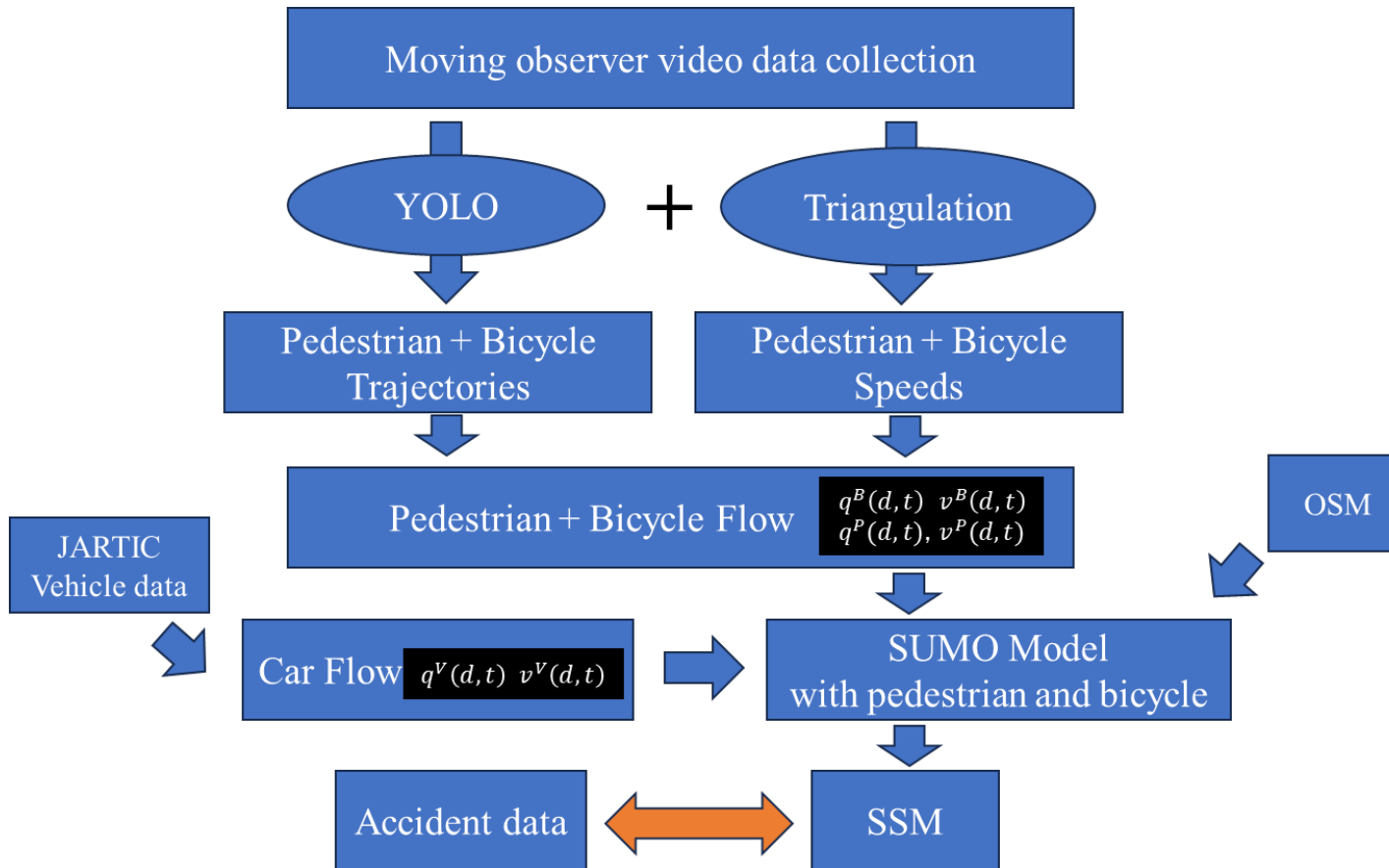
※ 京都先端科学大学経済経営学部の学生のアイデアを元で作成しました



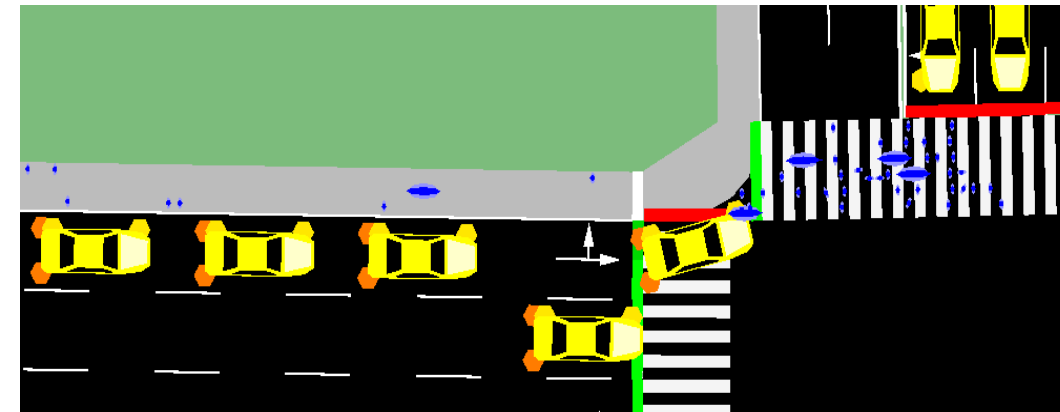
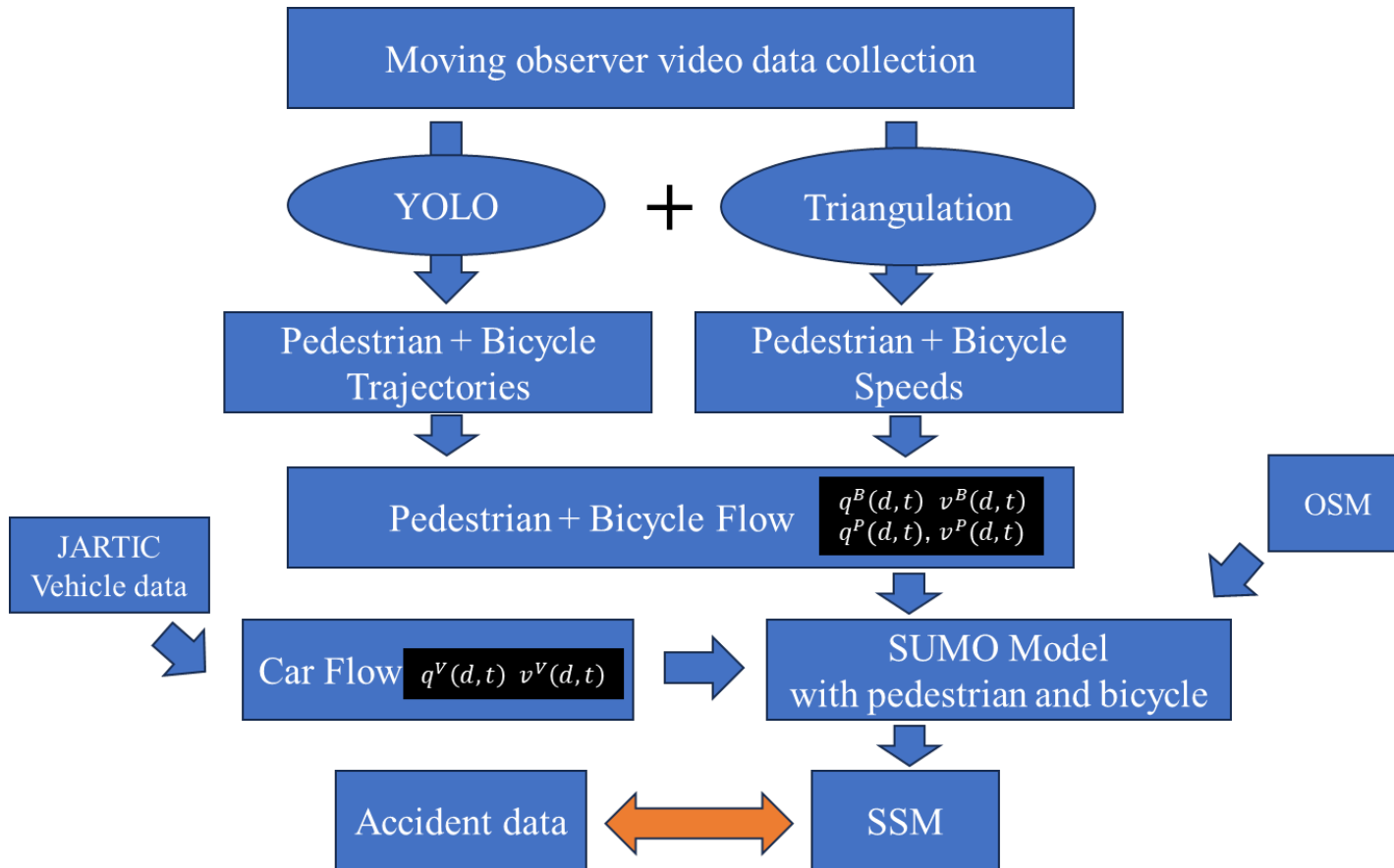
Ongoing research: Focus on bicycle accidents



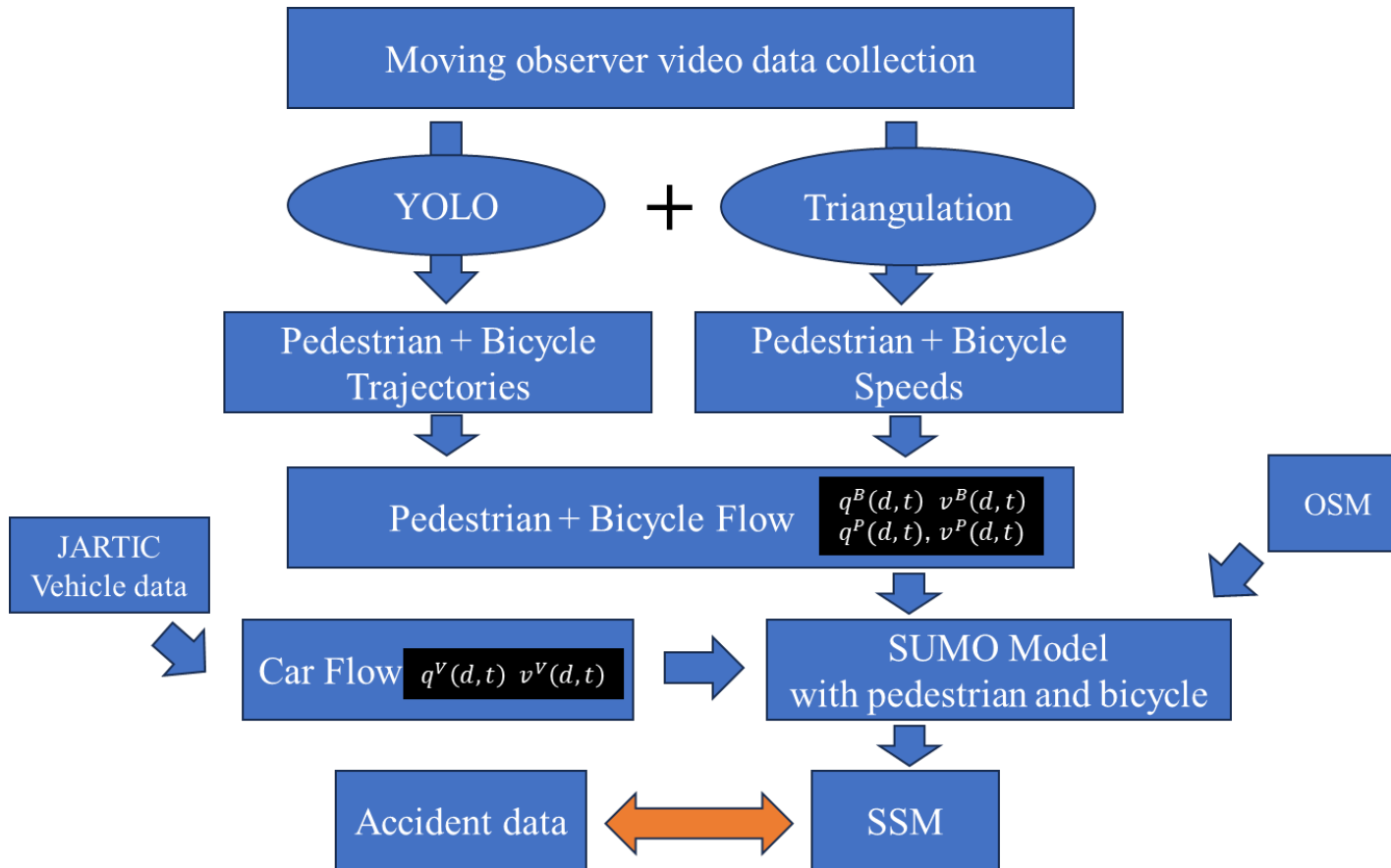
Ongoing research: Focus on bicycle accidents



Ongoing research: Focus on bicycle accidents

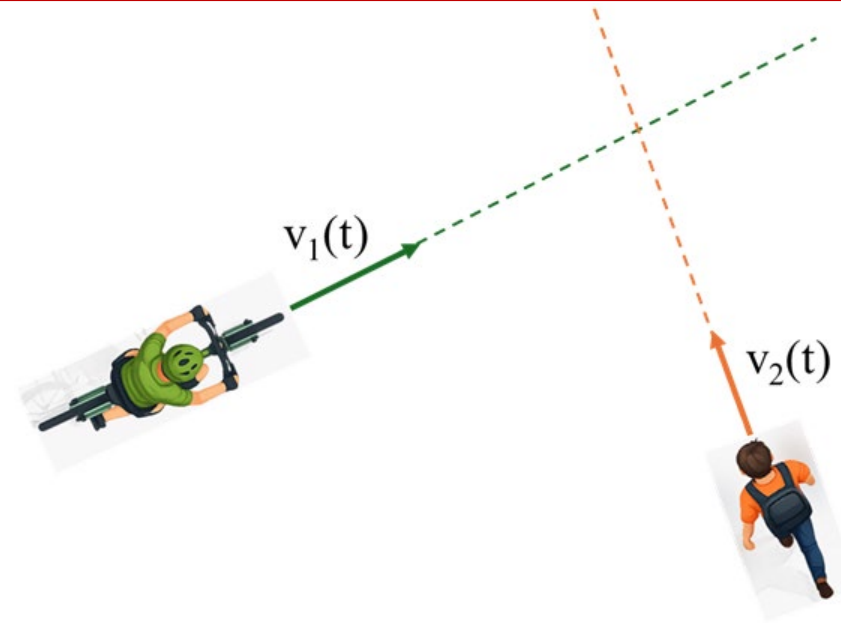
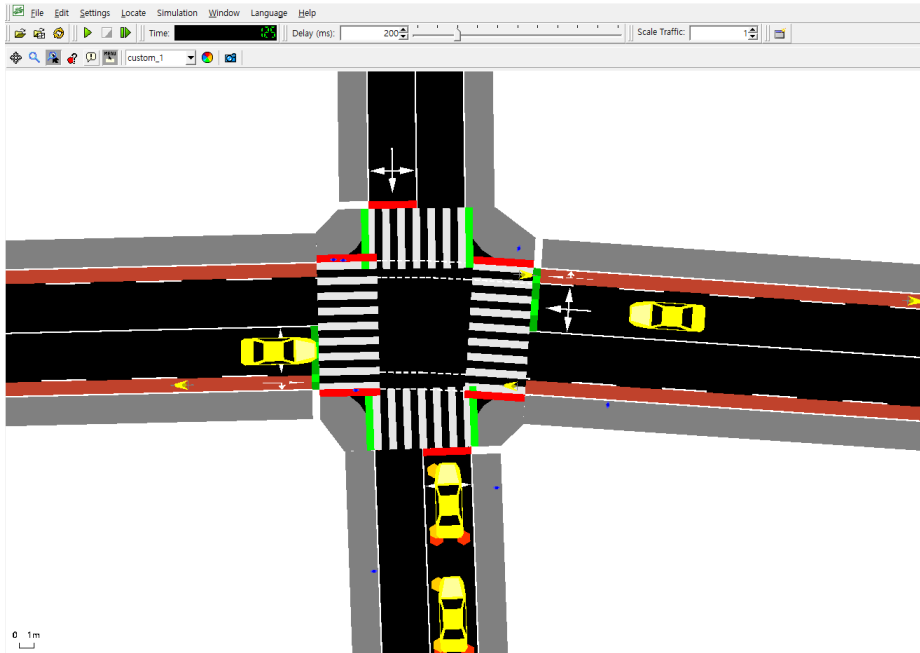


Ongoing research: Focus on bicycle accidents



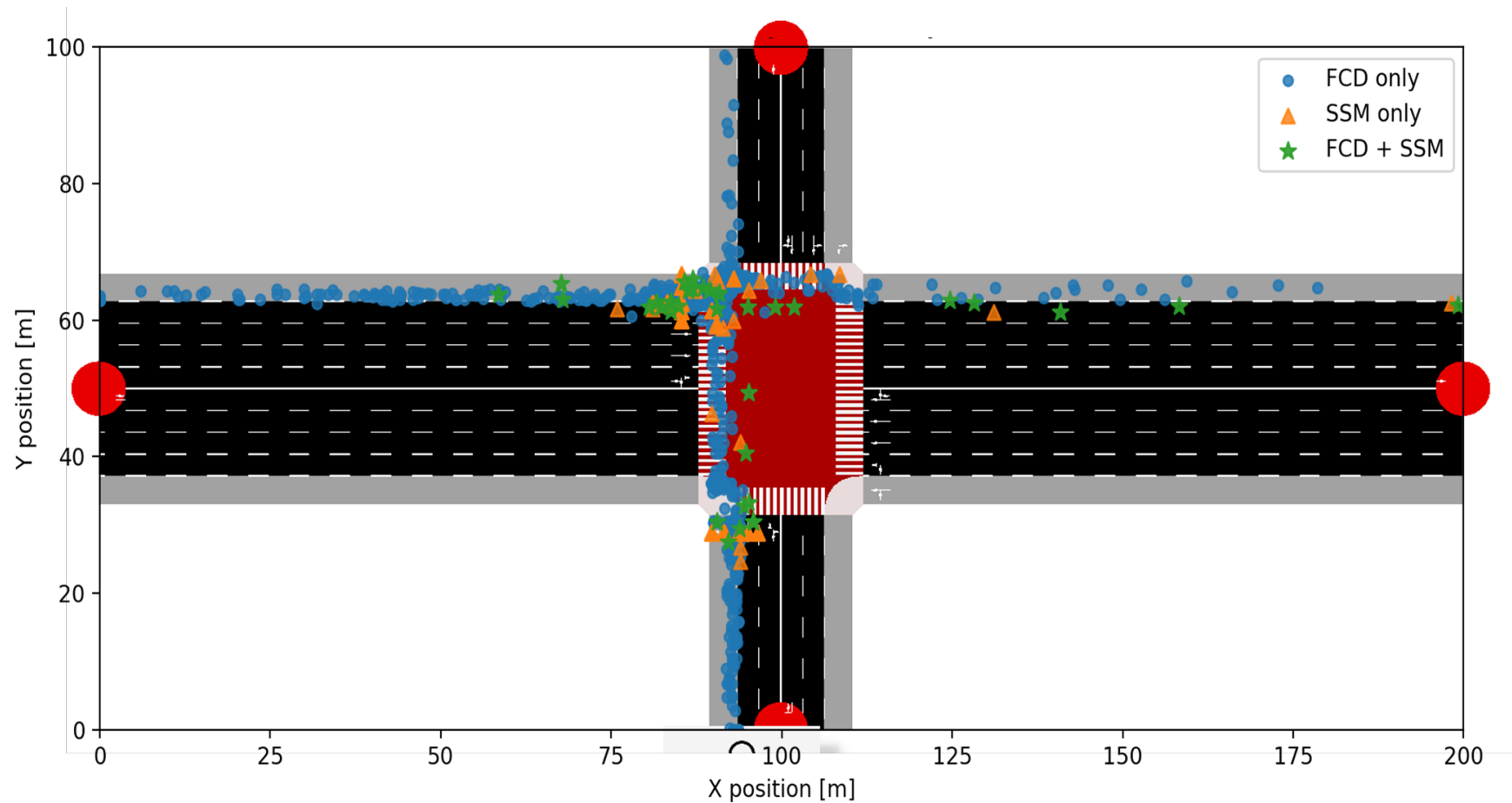
交差点 名称	御前通	七本松通	千本通	壬生川通	大宮通	猪熊通	関連 性
実際の 事故							
6 区画の 場合の 衝突地点							○
12 区画 の場合の 衝突地点							○
TTC							△
PET							△

Surrogate Safety Indicator for Mixed-Agent simulation models

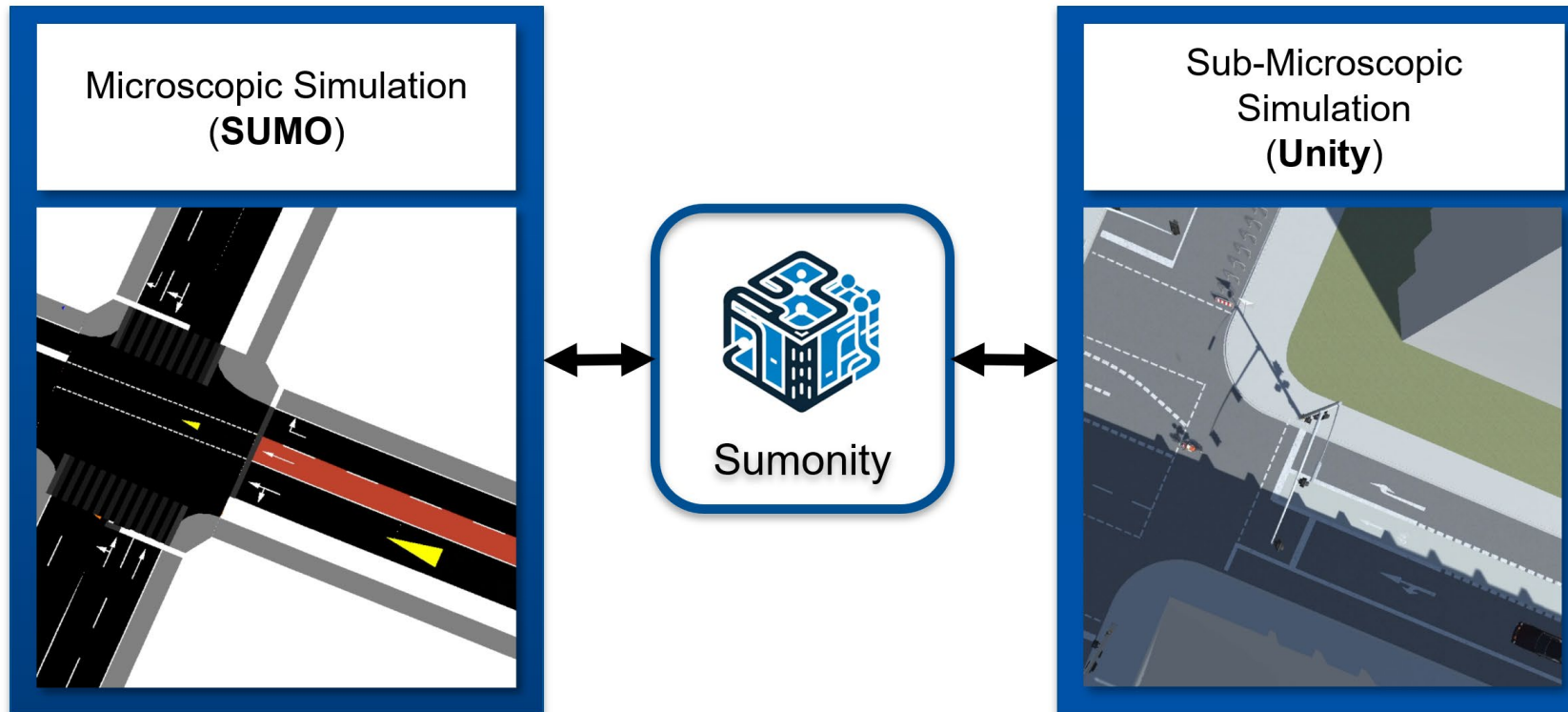


- Micro-simulating trajectories enable the inference of latent collision risk even in the absence of observed collisions.
- SSI for traffic more established than for bicycles

Conflict locations comparison: FCD, SSM



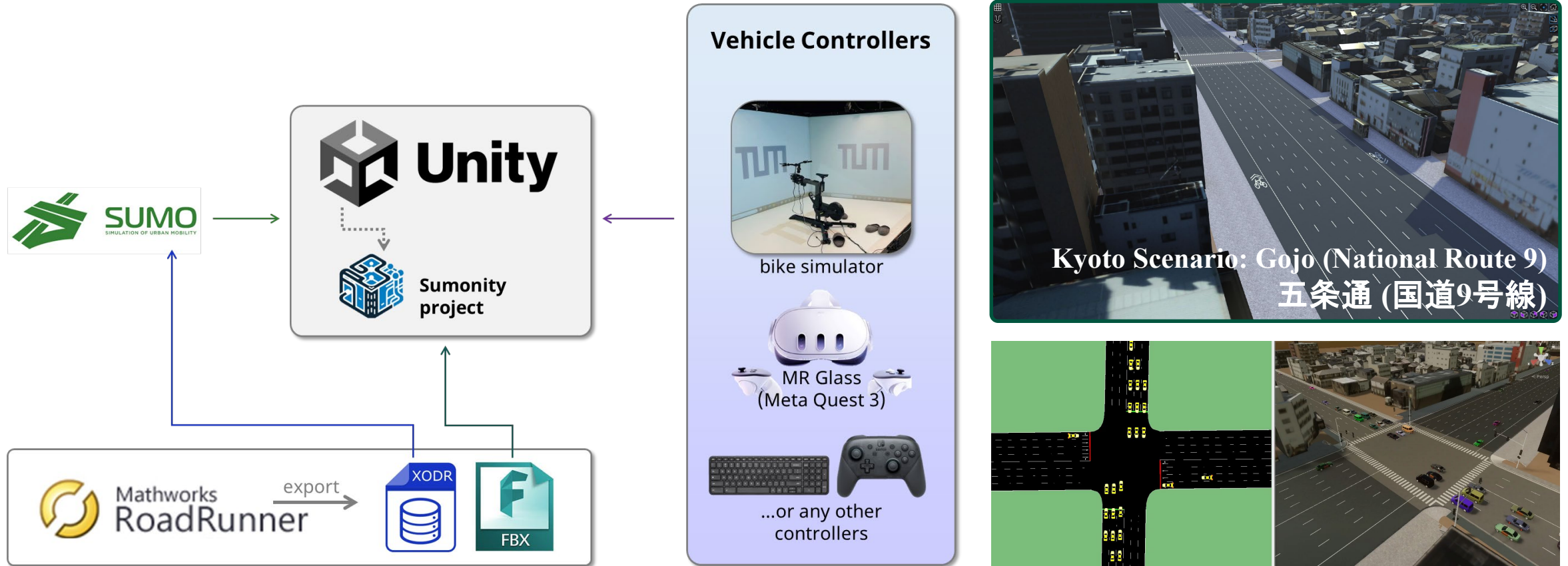
Sumonity



handles routing, car-following,
and lane-changing

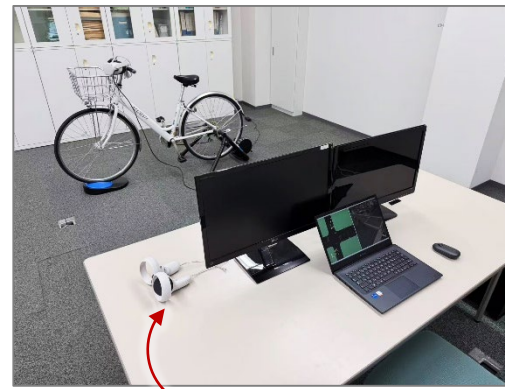
handles vehicle–e-scooter interaction,
acc/dec, turning, body motion, and collisions

Sumonity Framework



Adding Background Vehicles/Pedestrians/Bikes/Micro-Mobilities

Bike simulator



Simulation System



Path Sensor



Speed Sensor and the Resistance

Evaluation of scenarios via feedback from participants as well as SUMO output measures:

- observed flow differences
- safety surrogate measures such as "time to collision"

“Conclusions” and discussion

- Role of POIs, in particularly restaurants, convenience stores and supermarkets at junction vicinity discussed, controlling as much as possible for “liveliness”: Planning implications?
- Importance of field of view can be quantified through number of corner cuts
- Analysis of 2-wheeler accidents deserve specific attention and require additional tools: SSM definition and SUMONITY as tool for collaborative research