

# IDD-X: A Multi-View Dataset for Ego-relative Important Object Localization and Explanation in Dense and Unstructured Traffic

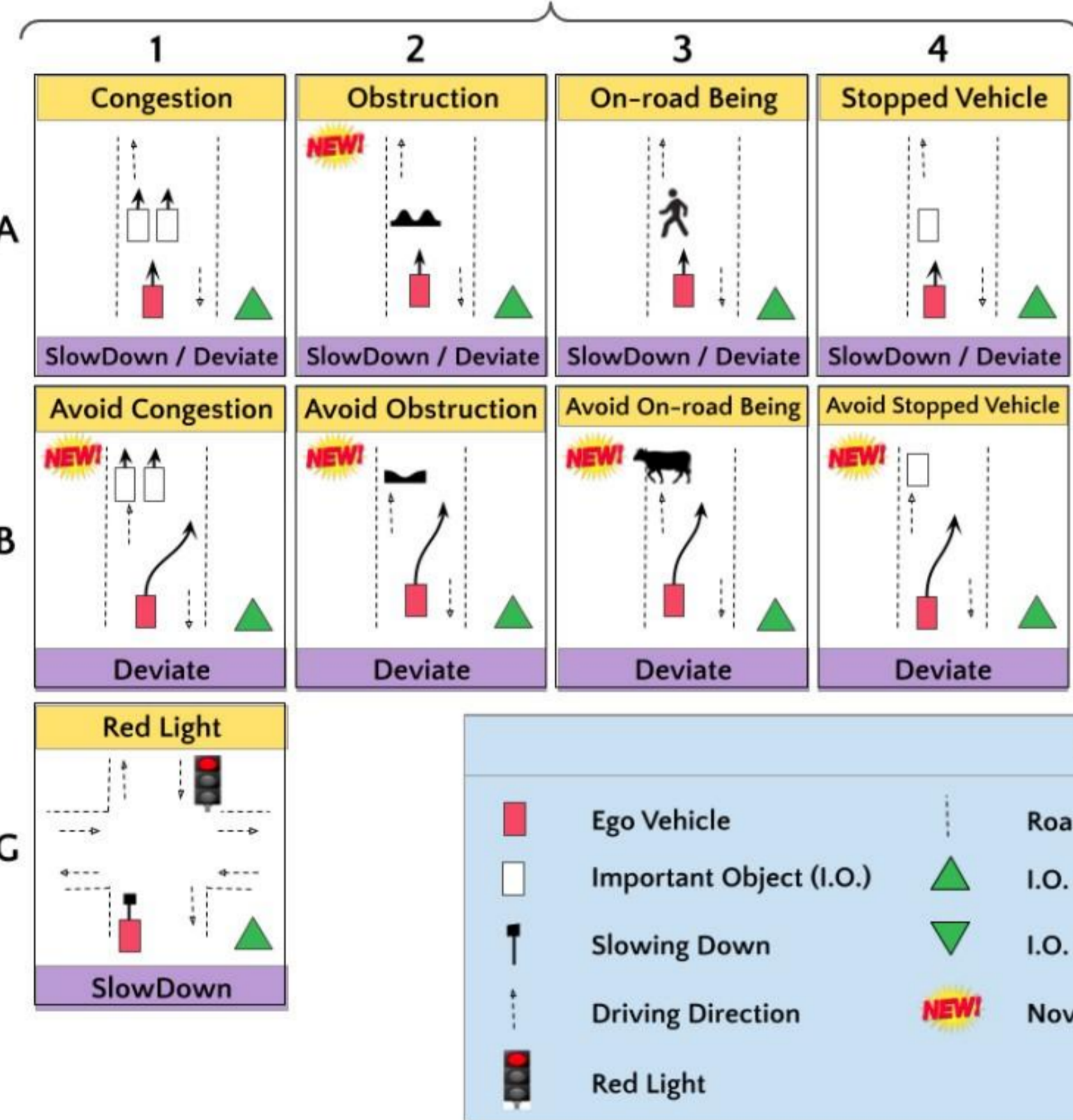
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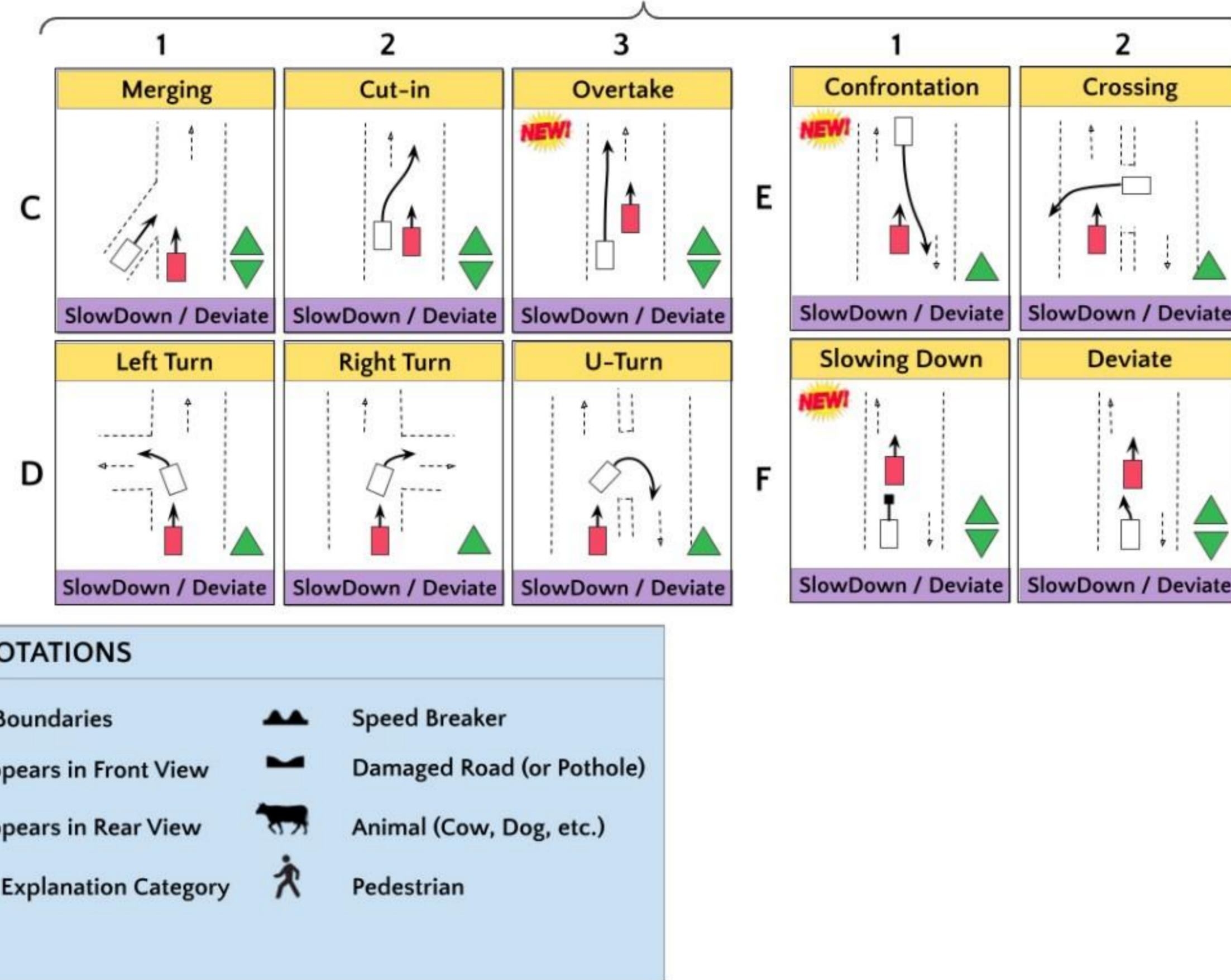
## DATASET MOTIVATION & OVERVIEW

- Intelligent vehicle systems require a deep understanding of the interplay between road conditions, surrounding entities, and the ego vehicle's driving behavior for safe and efficient navigation.
- Existing datasets, predominantly geared towards structured and sparse traffic scenarios, fall short of capturing the complexity of driving in unstructured dense, and heterogeneous traffic environments in developing countries.
- To fill this gap, we present **IDD-X**:
  - A large-scale (> 85 hours) dual-view (front and rear) driving video dataset
  - Captured in **Dense, Heterogeneous, and Unstructured** Traffic
  - With **697K** Important Object Bounding Boxes (**9K** Tracks) of **10** different object categories.
  - Containing **1-12** Important Objects (Per Video) with **19** Ego-relative Explanation Categories (depicted below)

### (I) Passive influence on ego's driving decision

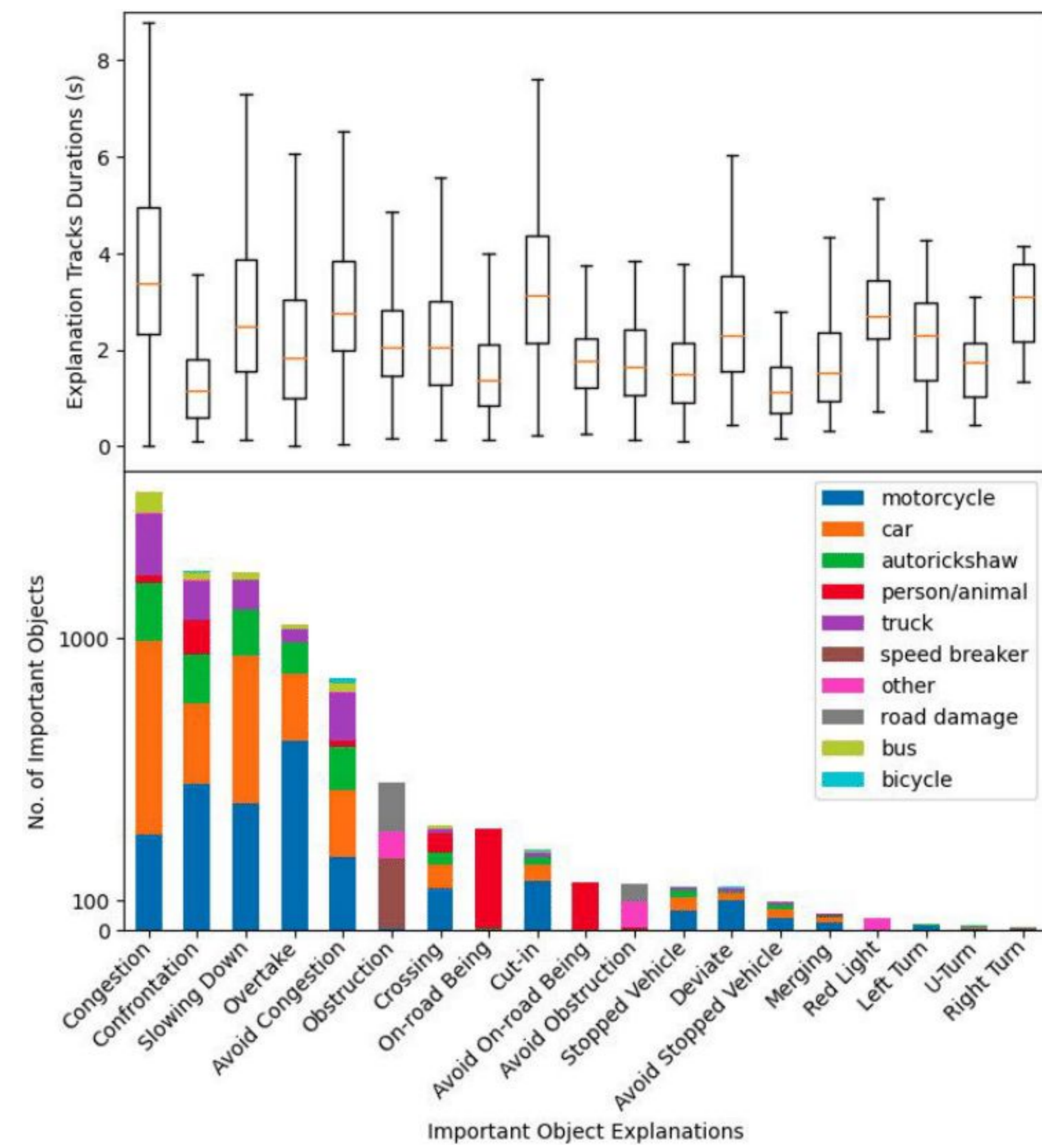


### (II) Ego-relative maneuvering styles of Important Objects



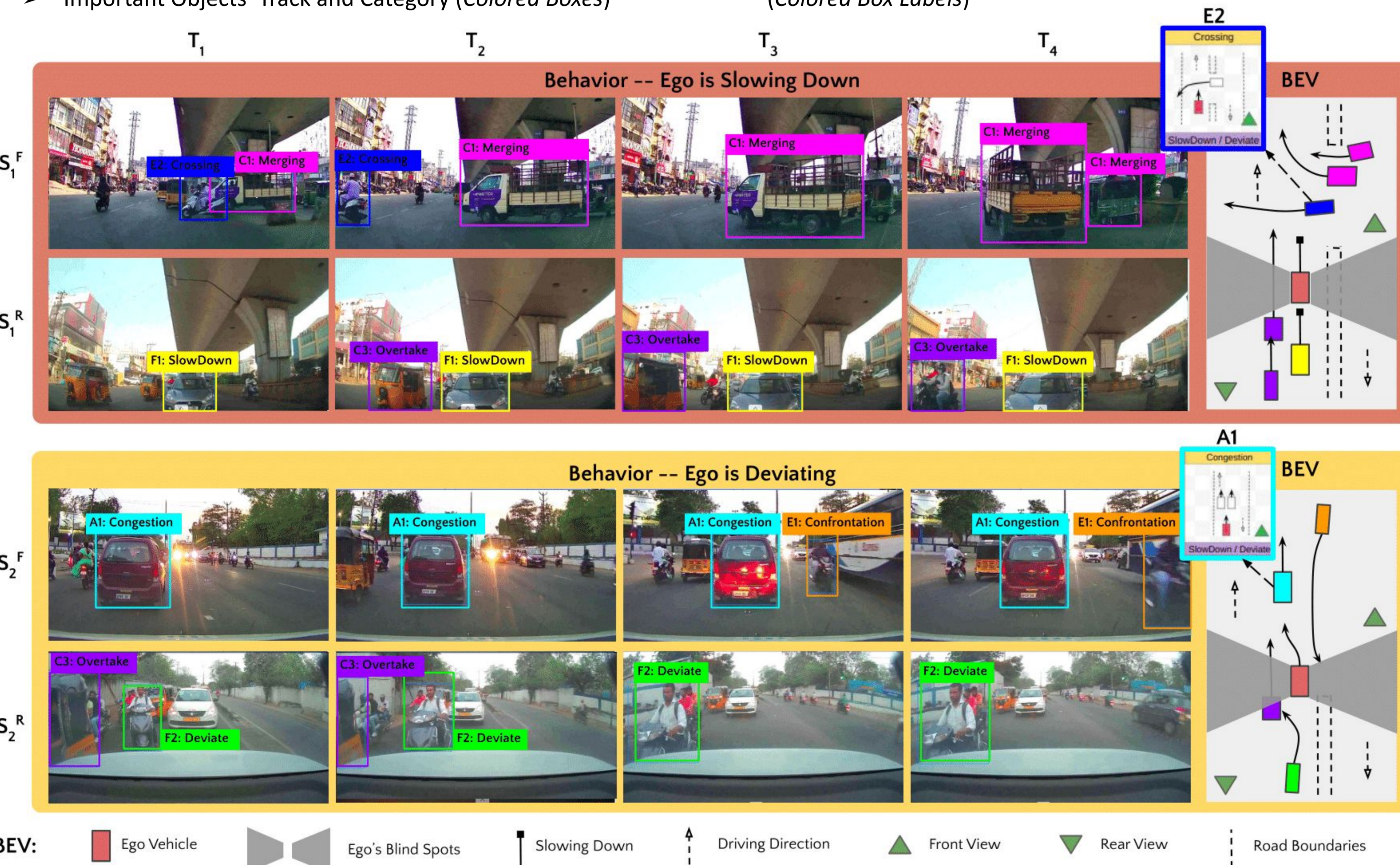
## DATA STATISTICS

- Important Object Explanations' Duration and Heavy-Tail Distribution



## IDD-X DATASET ANNOTATIONS

- Ego Vehicle's Driving Behavior (e.g. *Slowing Down*)
- Important Objects' Track and Category (*Colored Boxes*)
- Important Objects' Ego-relative Explanation Category (*Colored Box Labels*)

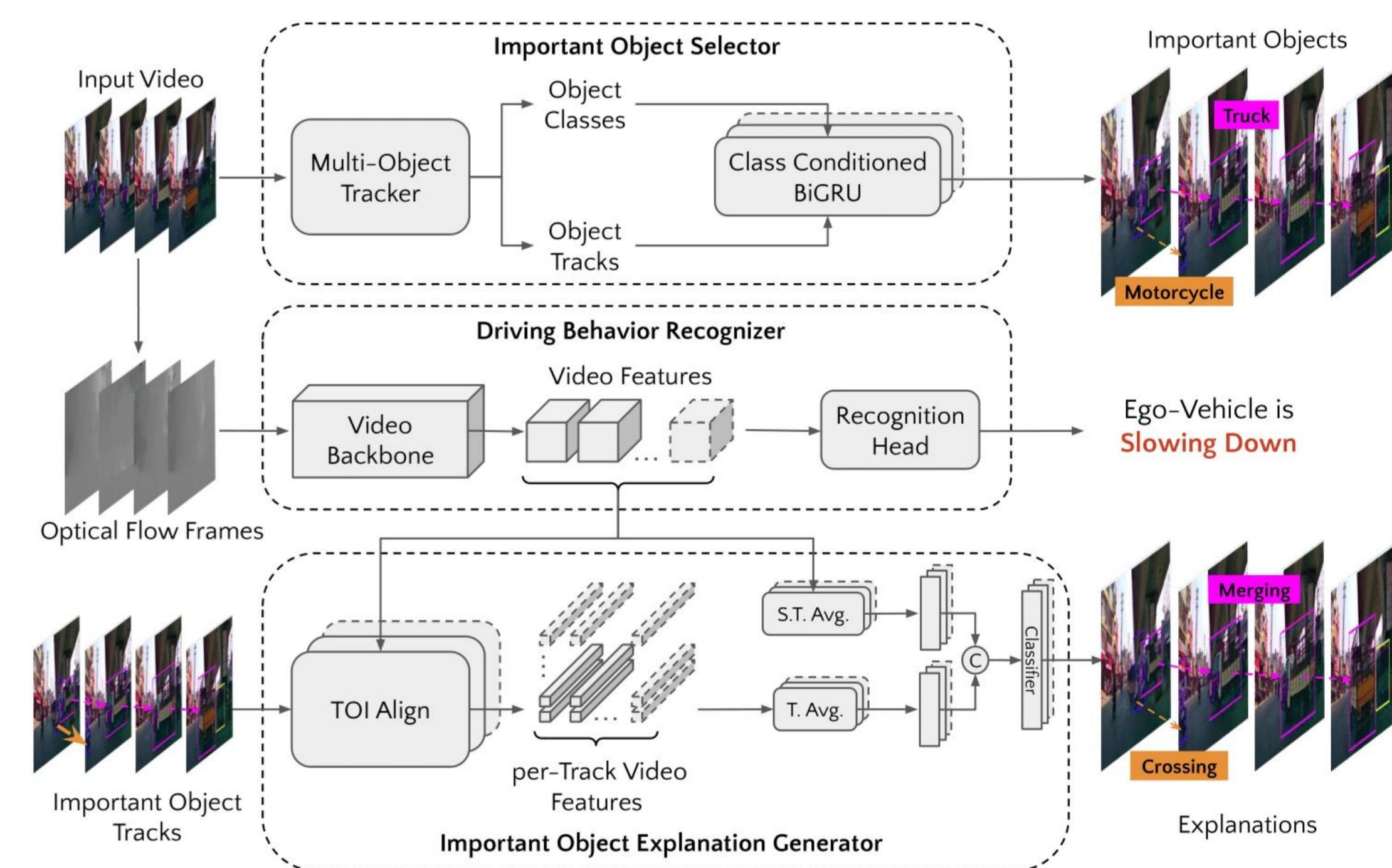


BEV: Ego Vehicle, Ego's Blind Spots, Slowing Down, Driving Direction, Front View, Rear View, Road Boundaries

Figure shows Front (F) and Rear (R) Views of Sample Driving Scenarios ( $S_1, S_2$ ) at Timeframes ( $T_1-T_4$ ) with Important Object Location and Explanation annotations in colored boxes and the Ego-vehicle's Driving Behavior label at the top.

## APPROACH

- Introduced New Methods for:
  - Important Object Track Identification
  - Important Object Explanation Prediction



## RESULTS

### 1. Important Object Track Identification

Model	Conditioning	Precision	Recall	F1-Score
BiGRU	-	35.0	86.1	49.8
Class Conditioned BiGRU	At $t = 0$	34.8	86.5	49.6
Class Conditioned BiGRU	At $t = T$	34.9	85.6	49.6
Class Conditioned BiGRU	At all $t$	35.8	86.8	50.7

- Object's Class information conditioned throughout its trajectory improves performance.

### 2. Important Object Explanation Prediction

Input Features	Congestion	Confrontation	Avoid Congestion	Overtaking	Crossing	Interfering Being	Cut-in	Avoid On-road Being	Avg. F1-Score	Wl. Avg. F1-Score
TOI-Aligned	72.2	48.6	53.3	53.9	37.5	17.4	7.3	0.0	36.3	54.8
TOI-Aligned + Context	71.0	53.6	47.2	48.6	44.7	22.4	34.7	25.0	43.4	55.8

- Video Features (Context) along with per-Track Features gives best average and overall performances.

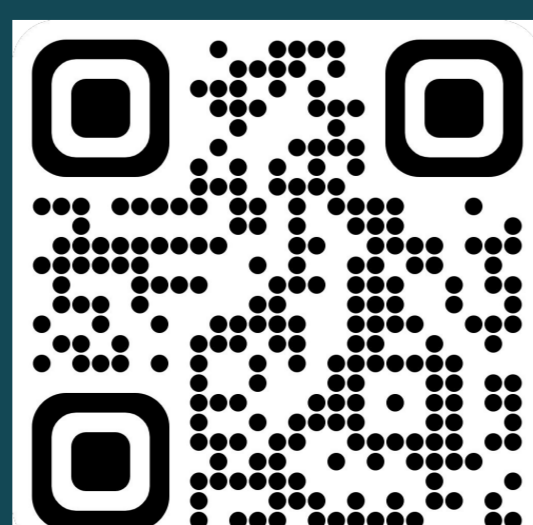
## EXISTING DATASETS COMPARISONS

Dataset	Multi-View	I.O. Location	I.O. Track	Irregular Road Surface Description	Dense Traffic	Unstructured Traffic	#I.O. Bounding Boxes	#I.O.s Per Video	#I.O. Explanation Categories	#I.O. Categories	#I.O. Explanation Type
IDD-X	✓	✓	✓ (9K)	✓	✓	✓	697K	1-12	19	10	Categorical
DRAMA [3]	-	✓	-	-	-	-	17K	1	-	3	Textual
H3D [1]	-	✓	-	-	-	-	8K	1-3	-	4	-
BDD-OIA [4]	-	-	-	-	✓	-	-	-	8 <sup>1</sup>	6	Categorical
OIE [2]	-	✓	-	-	-	-	4K	1-2	-	2	-
HDD [6]	-	-	-	-	-	-	-	1	5	4	Categorical
BDD-X [5]	-	-	-	-	-	-	-	-	-	-	Textual
METEOR [7]	-	-	-	-	✓	✓	-	-	-	-	-

\* I.O. = Important Object

## ACKNOWLEDGMENT

The project is funded by the iHubData and Mobility at IIIT Hyderabad. We thank the data collection and annotation team for their effort.



Scan for Code, Dataset download and related resources on the Project Page.

idd-x.github.io

