

PEDESTRIAN TRAFFIC

Data Drivers & NSDC

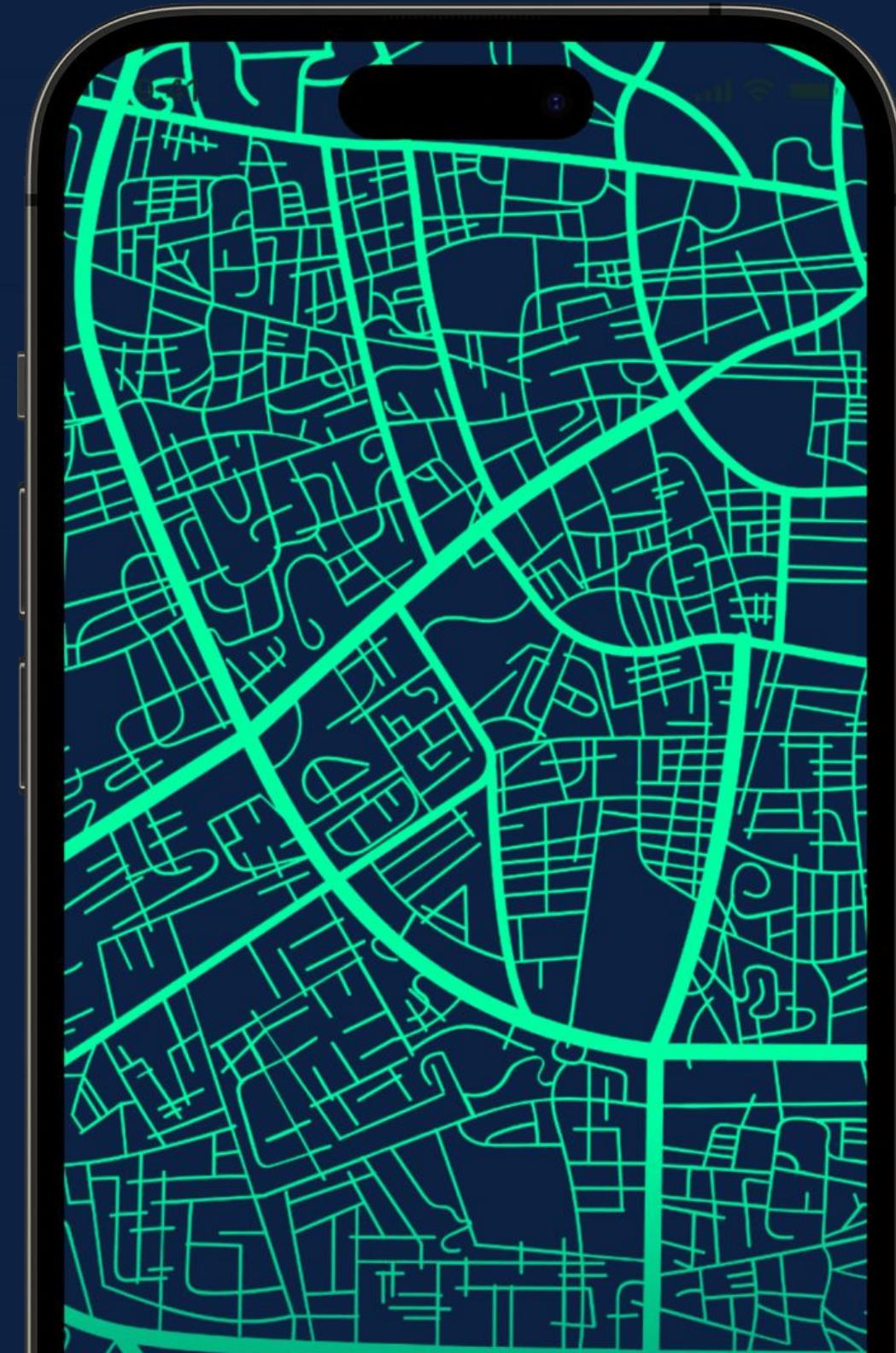


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Our Team

Data Drivers



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ABOUT THE DATA

- Multi-agent Pedestrian Tracking (UCI Machine Learning Repository)
 - Multivariate, Sequential, Time-Series
 - 4760 observations x 14 attributes
 - Recorded from a vehicle in southern Germany
- **Our Research Question:** How does the presence & movement of other agents affect the head angle of pedestrians in a given scene?
- Label: head_yaw: yaw head angle in degrees (float)
- Suggested Applications: Autonomous driving, traffic management, urban planning, multi-agent motion prediction



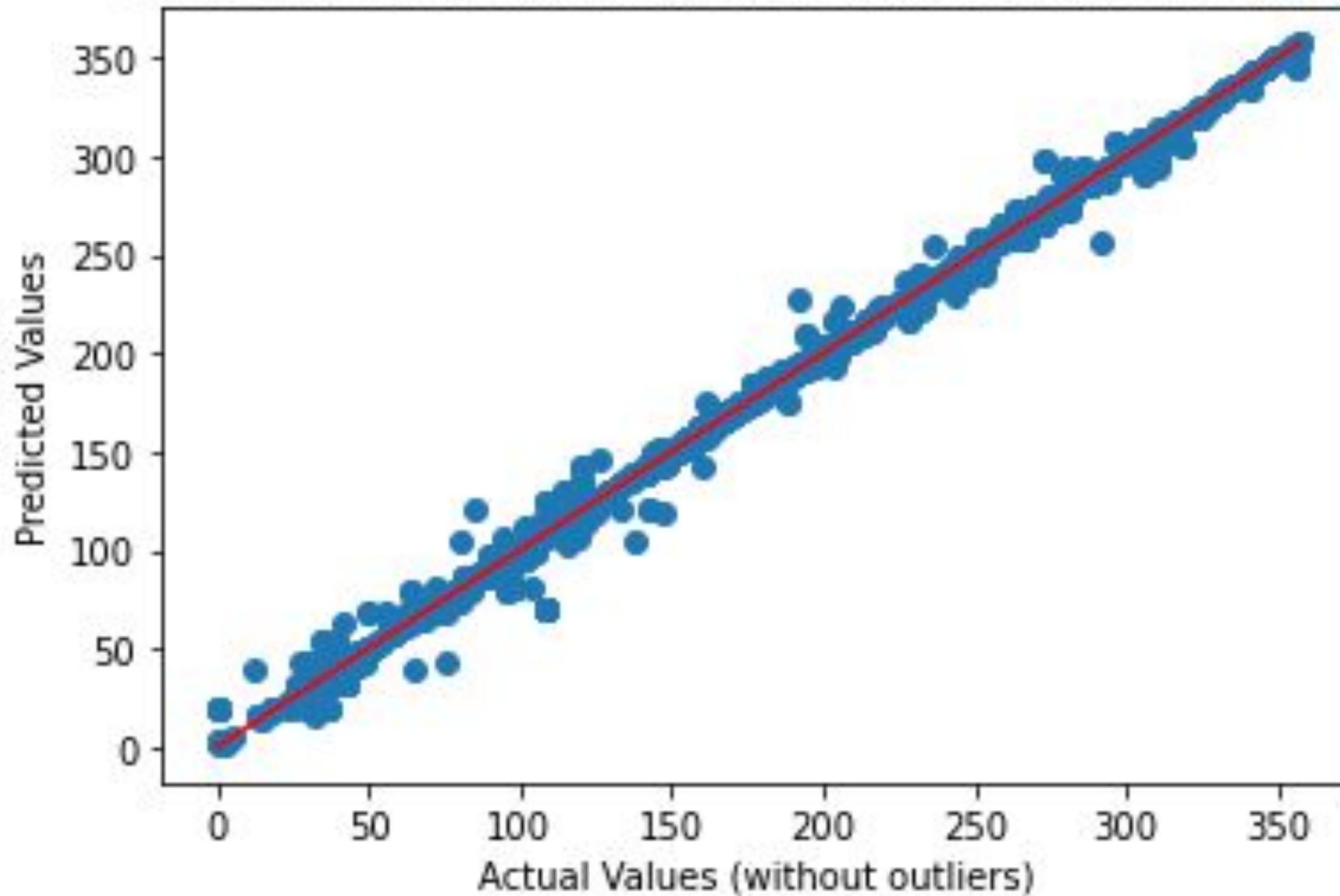
CLEANING THE DATA

- Removal of missing data values for body & head angles
- Exploding coordinate lists for roll, pitch, & yaw for both body & head
- Visualizations of color-coded map with agent information
 - 2D & 3D Scatterplots
 - Heatmaps
 - Box plots
- Feature engineering using polynomial and multivariate regression
 - Linear & logistic regression



OUR PROGRESS

Scatter Plot of Actual vs. Predicted Values (without outliers)



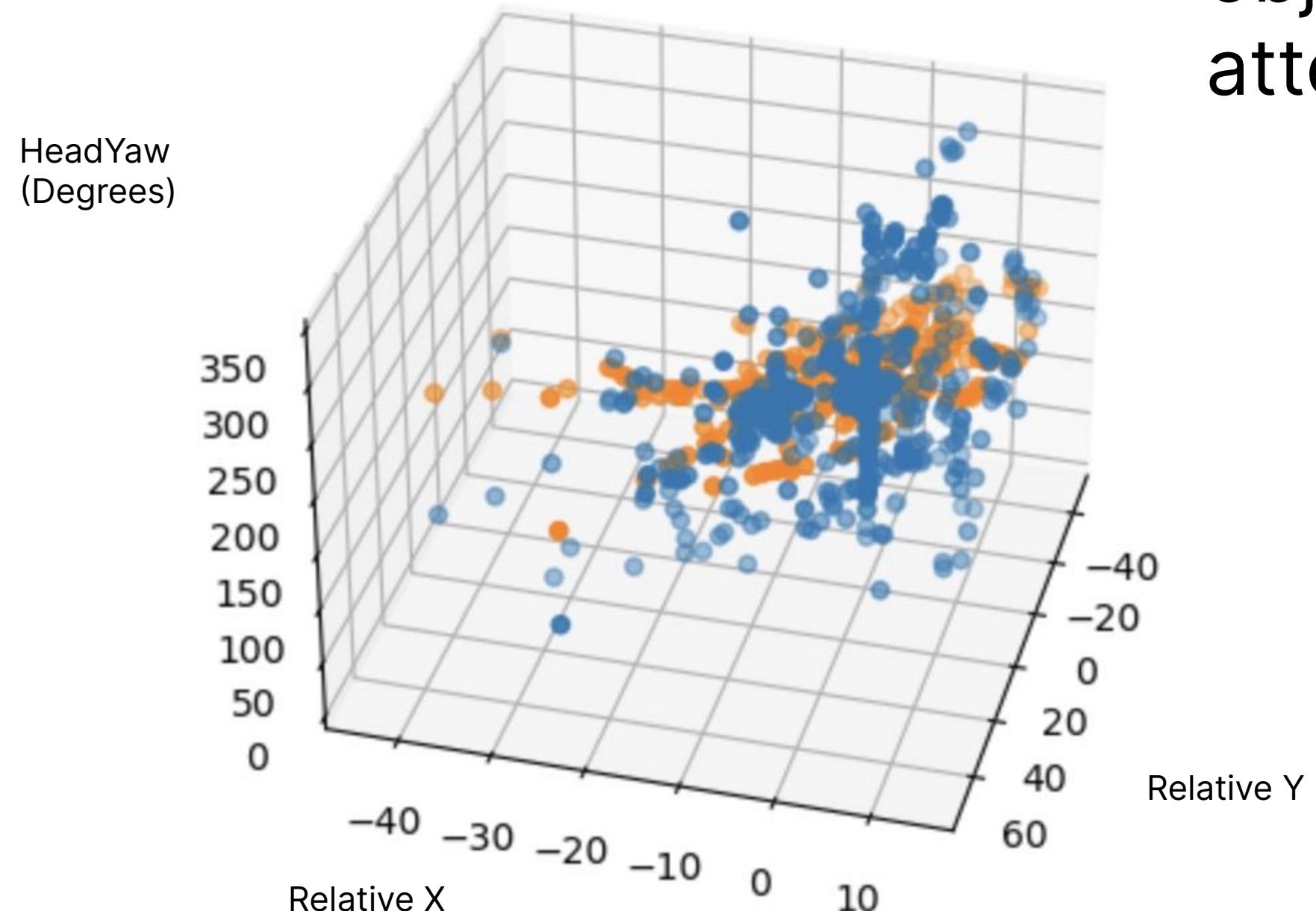
Mean Squared Error (original model): 164.56

Mean Squared Error (no outliers): 18.67

R Squared (original): 97%

R Squared (no outliers): 99.7%

P values for models created using different classes to determine which object were most likely to grab the attention of the pedestrian



$$-0.19504917(\text{relative_x}) - 0.13526806(\text{relative_y}) = \text{headyaw}$$

EGO CAR	<code>rel_x</code>	<code>4.360604e-15</code>
	<code>rel_y</code>	<code>1.176505e-32</code>
	<code>other_class</code>	<code>NaN</code>
	<code>headyaw</code>	<code>0.000000e+00</code>
	Name: <code>P> t </code> , dtype: float64	
CAR	<code>rel_x</code>	<code>0.822146</code>
	<code>rel_y</code>	<code>0.000004</code>
	<code>other_class</code>	<code>0.920038</code>
	<code>headyaw</code>	<code>0.000000</code>
	Name: <code>P> t </code> , dtype: float64	
PEDESTRIAN	<code>rel_x</code>	<code>1.000000</code>
	<code>rel_y</code>	<code>0.000466</code>
	<code>other_class</code>	<code>0.004170</code>
	<code>headyaw</code>	<code>0.000000</code>
	Name: <code>P> t </code> , dtype: float64	

CONCLUSIONS

- **Conclusion:**
 - Head angle linked to x-y positions
 - High correlation, low R-squared for head angle vs. closest object's position
 - Egocars draw most attention, likely due to safety concerns
- **Next Steps:**
 - Enhance data collection with advanced sensors and cameras
 - Raise public awareness about egocars' impact on pedestrians
 - Refine models for improved generalizability

THANK YOU

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