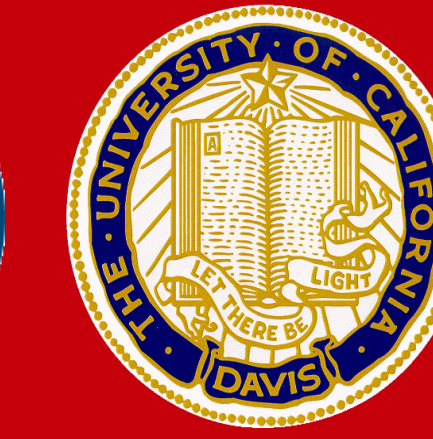


KrishnaCam: Using a Longitudinal, Single-Person, Egocentric Dataset for Scene Understanding Tasks


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Objective

Organize a large egocentric video collection (of real-world data from a single individual) into a richly annotated database that facilitates rapid analysis and enables exploration of new visual data understanding applications.

Dataset



Walking in urban/campus/residential areas, waiting at intersections and for bus

Time-span: 9 months

Duration: 70 hours

Total clips: 460

Device: Google Glass

Data: 720 p, 30 fps

Accelerometer, Gyroscope, Orientation, GPS

Shopping, eating

Evening and night recording

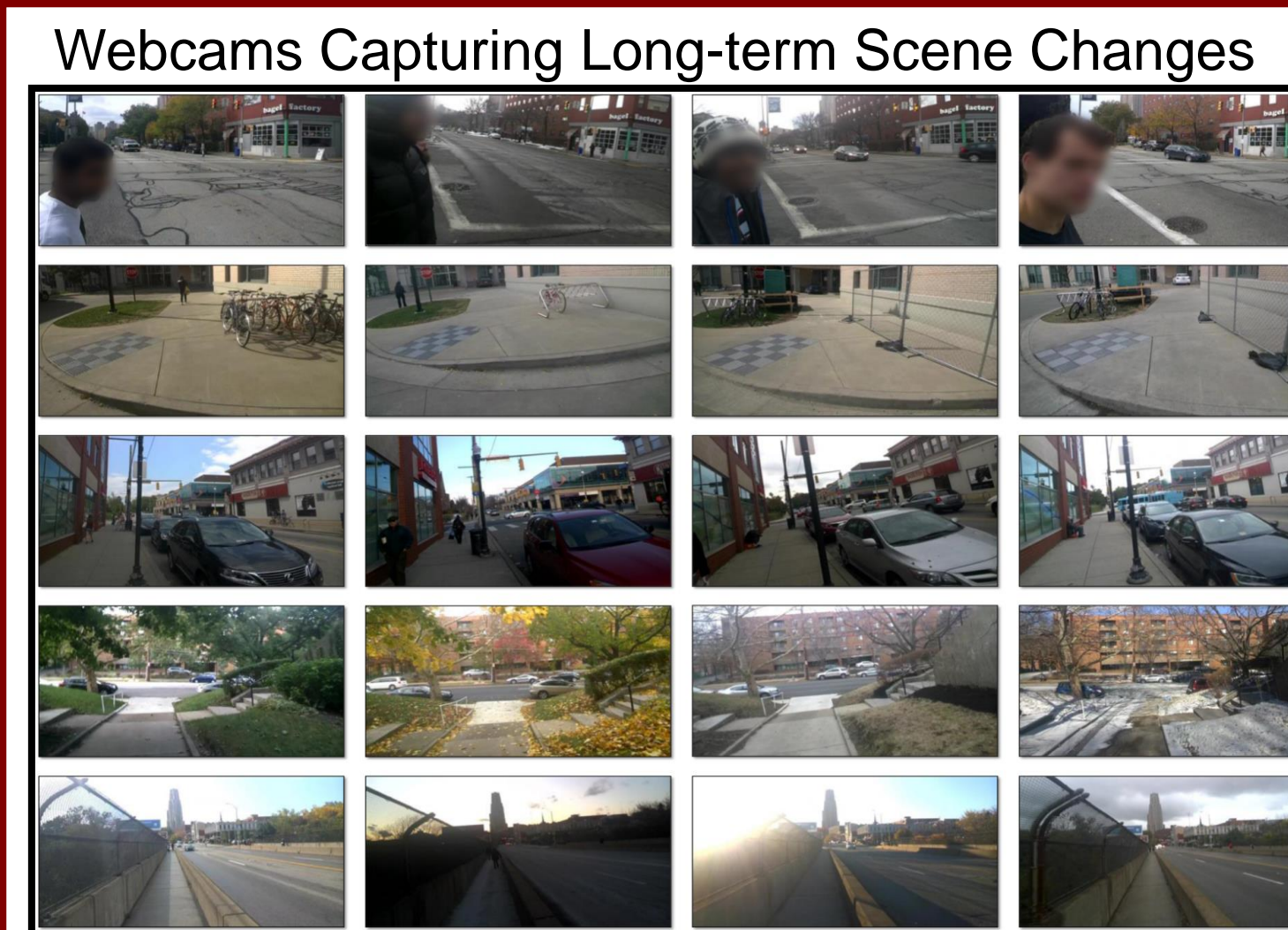
Activities in parks, at events

Seasonal change

Socializing with friends


Virtual Webcams

Webcams Capturing Long-term Scene Changes



- Although the egocentric camera is not stationary, long-term recording captures changes in a scene over time.
- Each row in figure captures (from top to bottom) changes in companion (person), movement of bicycle stand, parked cars, season and lighting

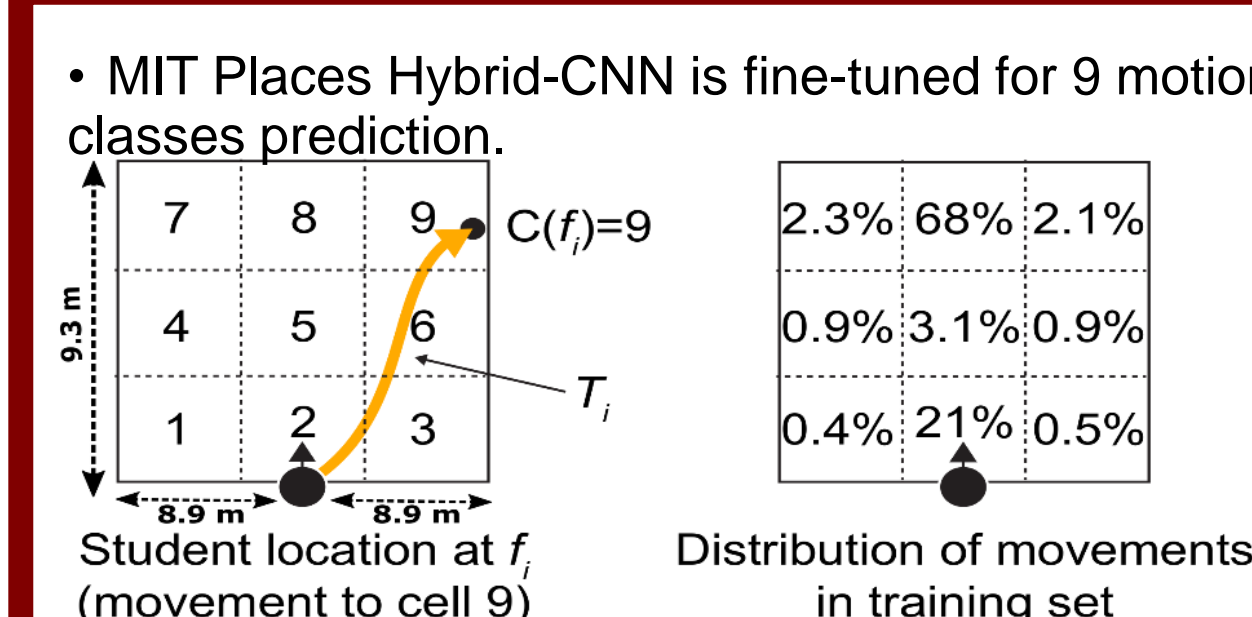
Trajectory Estimation



- Motion trajectory is estimated by accelerometer and orientation sensor in the smart phone.
- Trajectory represents motion in next 7 seconds.
- Yellow color indicates motion, red means stop.
- GPS was not accurate for short term trajectories.

Motion Class Prediction

MIT Places Hybrid-CNN is fine-tuned for 9 motion classes prediction.



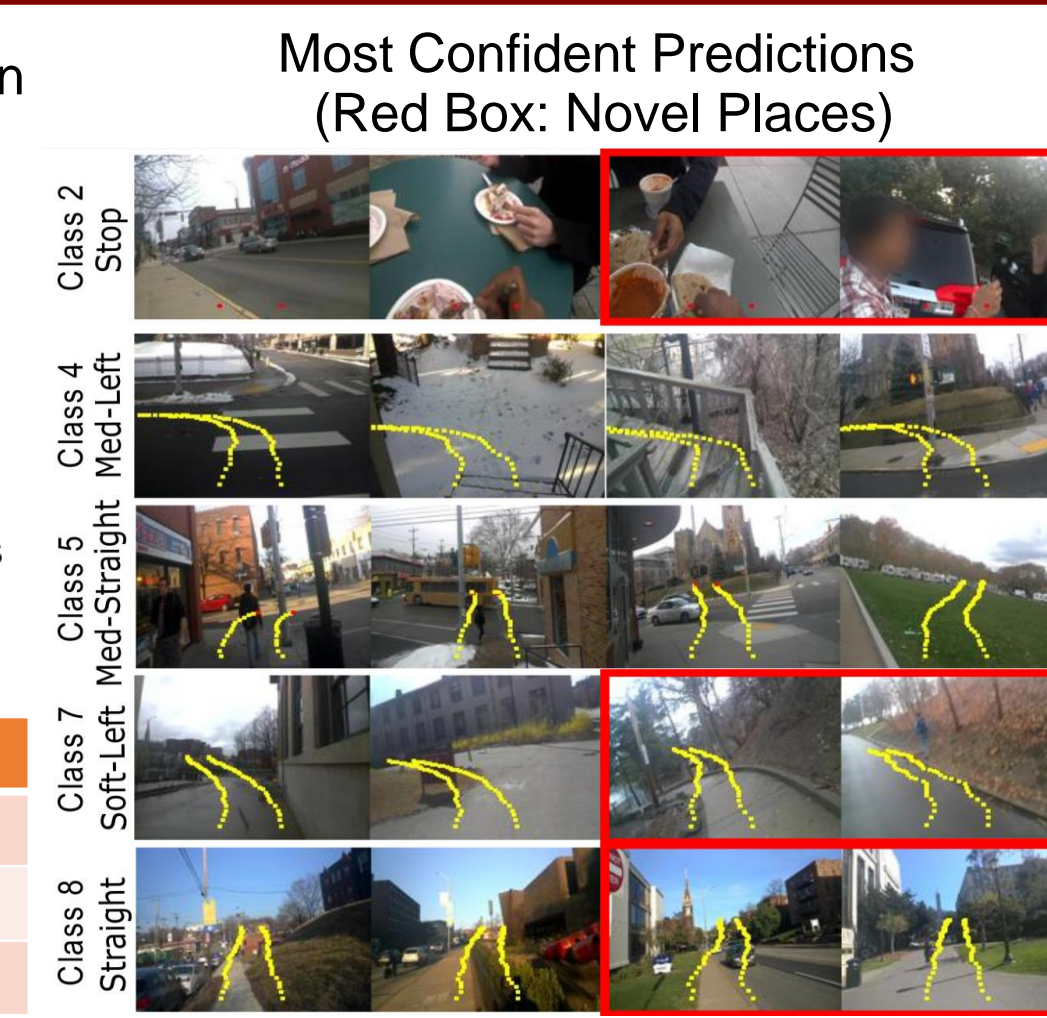
Student location at t_i (movement to cell 9)

Distribution of movements in training set

	Unvisited(%)	Visited(%)	Overall(%)
Fine-tuned	58.4	81.2	73.4
NN	54.9	81.4	72.2
Chance	43.2	51.3	48.5

Motion Class Prediction Accuracy

Most Confident Predictions (Red Box: Novel Places)

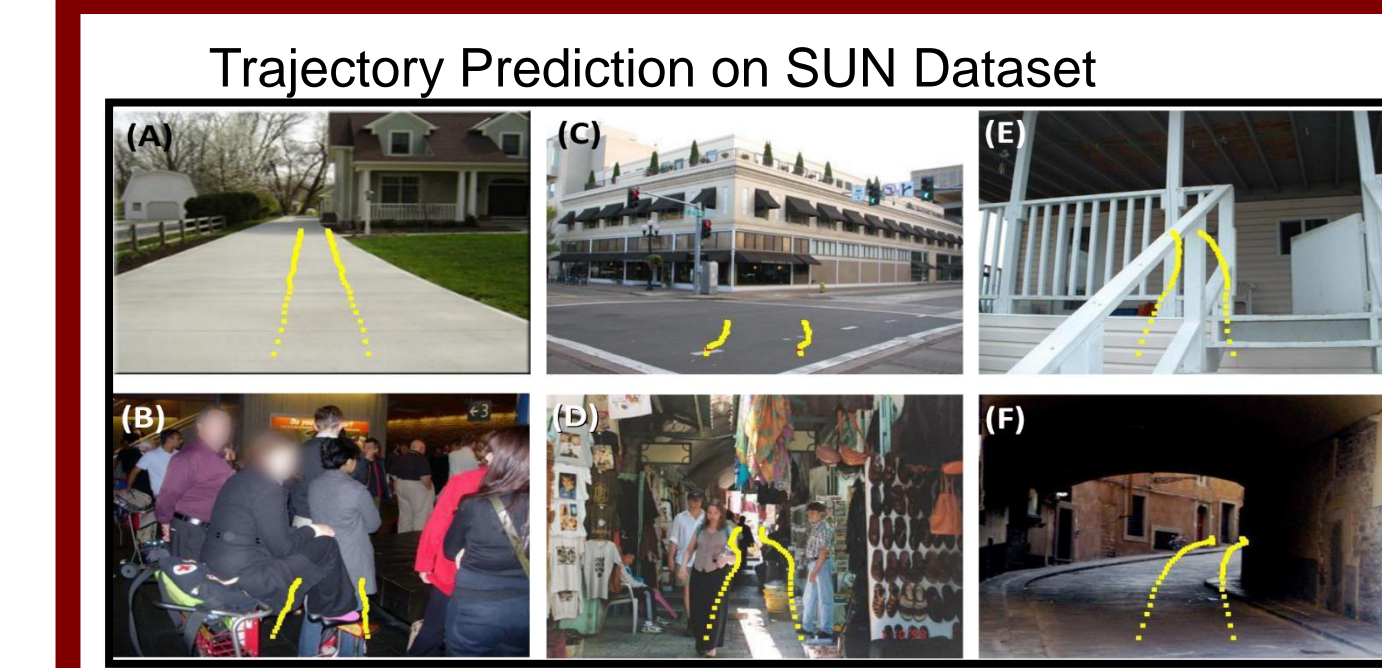


Trajectory Prediction

Trajectory is predicted by averaging trajectory of top 10 nearest neighbors in deep feature space (Pool-5 MIT Places Hybrid-CNN).

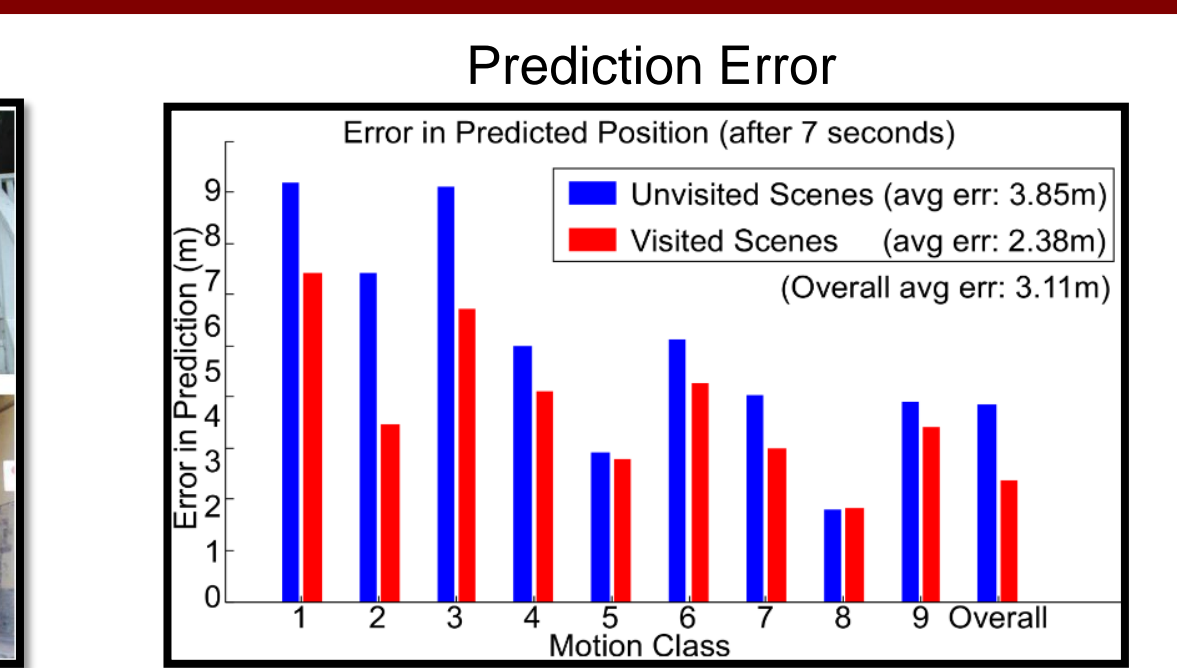
Ground-truth	Predicted Traj	Top 10 Neighbors	
			Predict Turn
			Predict straight at novel place
			Predict stop while eating
			Predict Stop at intersection
			Predict stop in front of car at novel place
			Predict turn
			Predict stop at novel bus stop

Trajectory Prediction on SUN Dataset



Prediction Error

Error in Predicted Position (after 7 seconds)



Unvisited Scenes (avg err: 3.85m)

Visited Scenes (avg err: 2.38m)

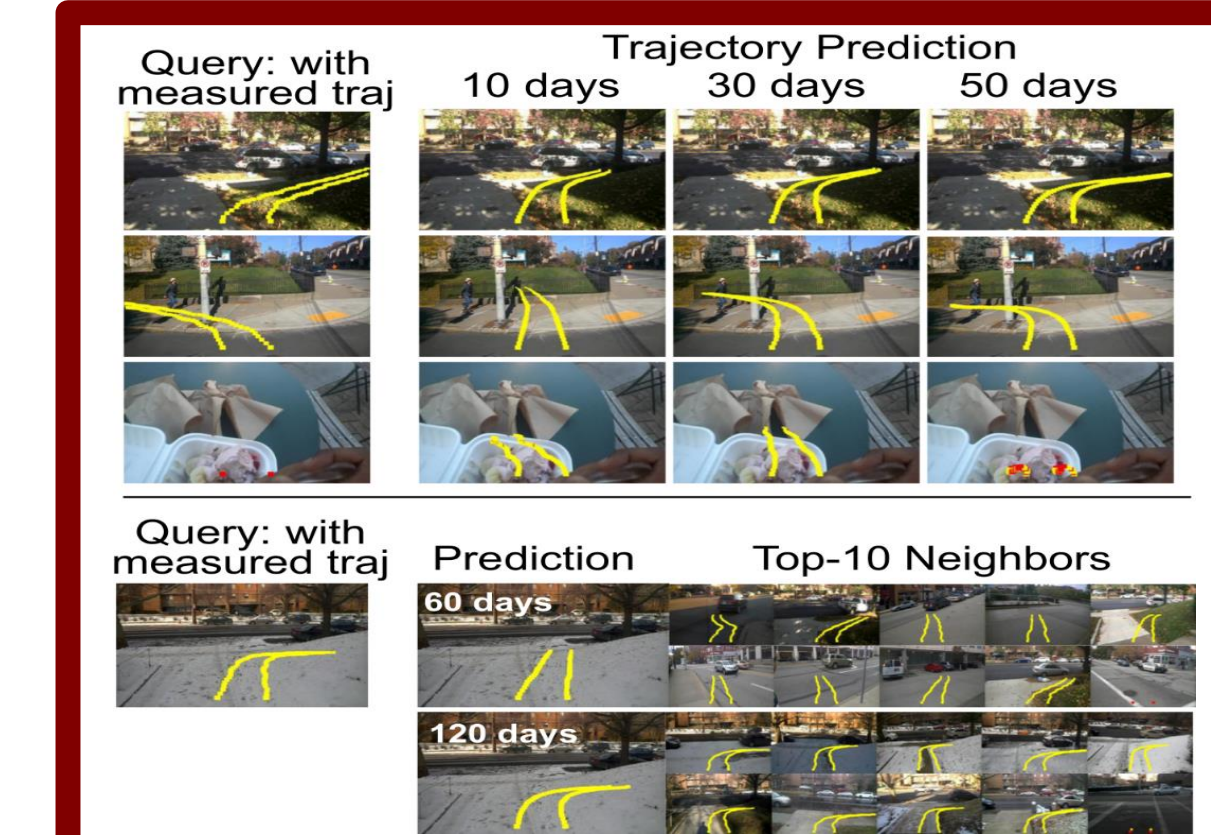
(Overall avg err: 3.11m)

Value of Extended Recording

Query: with measured traj

Trajectory Prediction

10 days 30 days 50 days

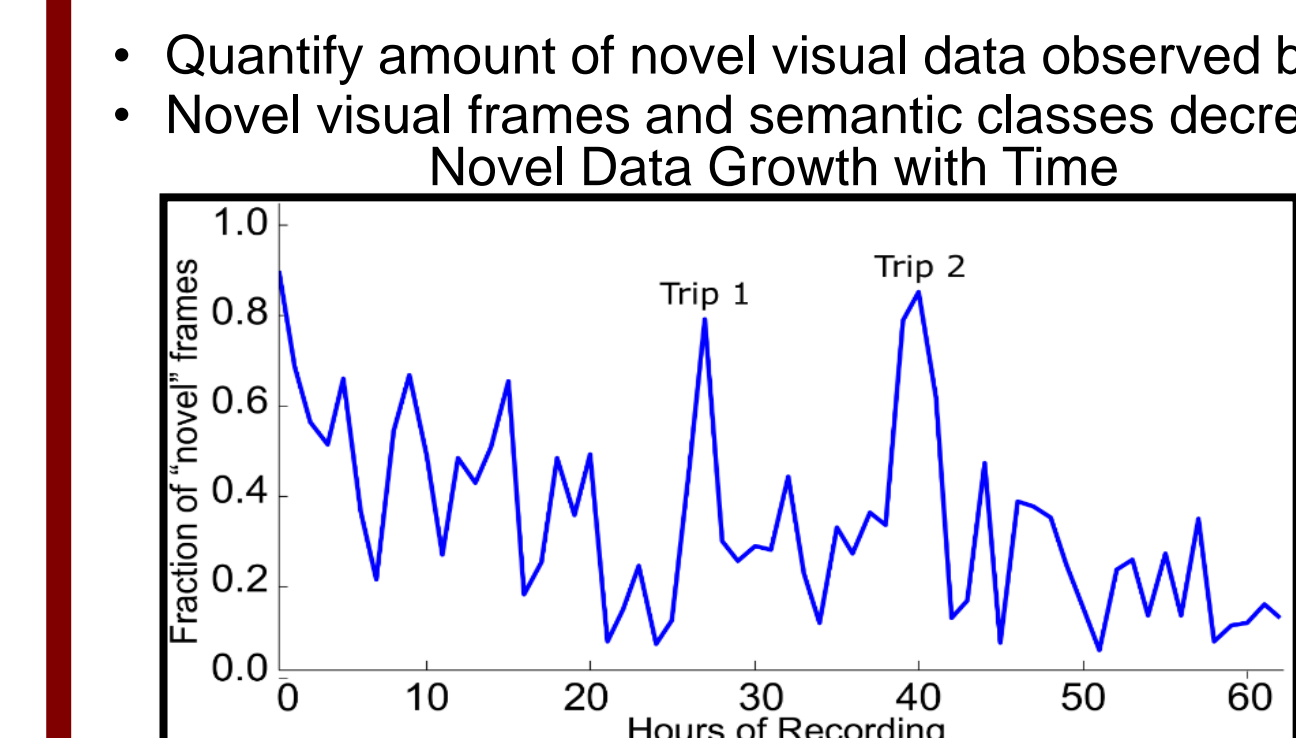


- Longer recording is needed to adequately sample rare events:
 - 10 days for common turn.
 - 30 days for less common turn.
 - 50 days for rare eating event.
- Not until four months of recording had occurred did snowing days begin to appear in the dataset, making prediction robust to seasonal change.
- Using only 50% and 25% of the training data, decreases motion class prediction accuracy relatively by 29% and 51% respectively.

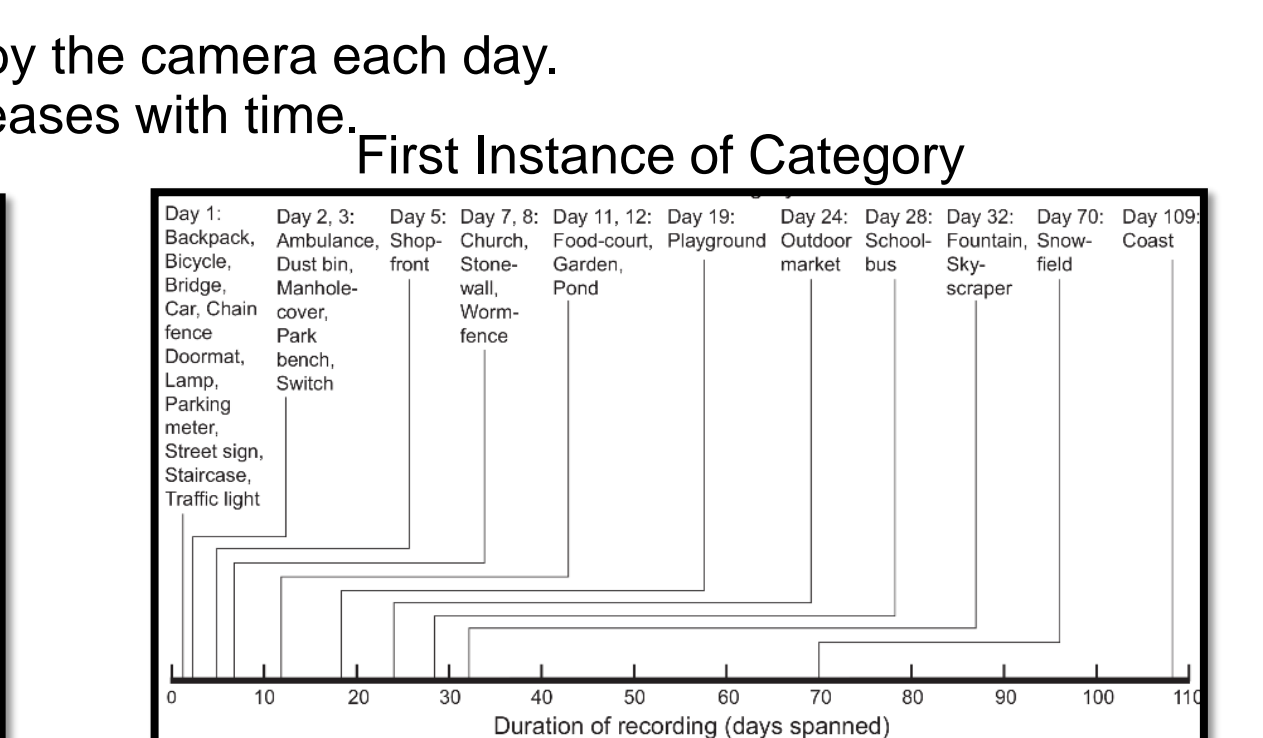
Novel Data Growth

- Quantify amount of novel visual data observed by the camera each day.
- Novel visual frames and semantic classes decreases with time.

Novel Data Growth with Time

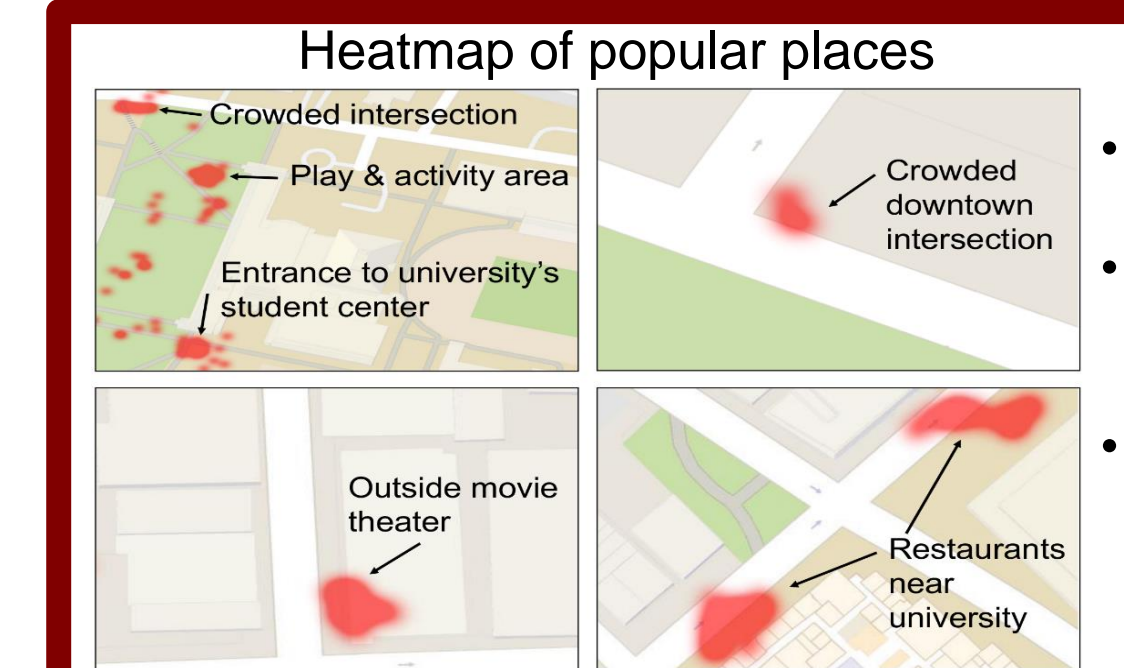


First Instance of Category



Popular Places

Heatmap of popular places



- 17% of the dataset contains at least one human.
- Popular places are found correlating pedestrian detections with GPS measurements.
- Red regions indicate locations where (on average) more than four people are present in images.