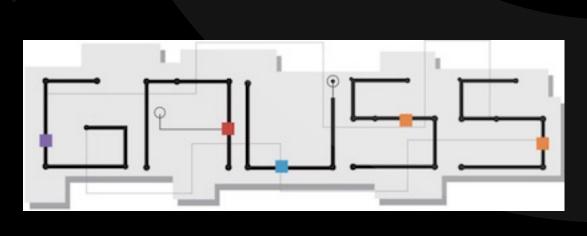
Towards Anomaly Detectors that Learn Continuously











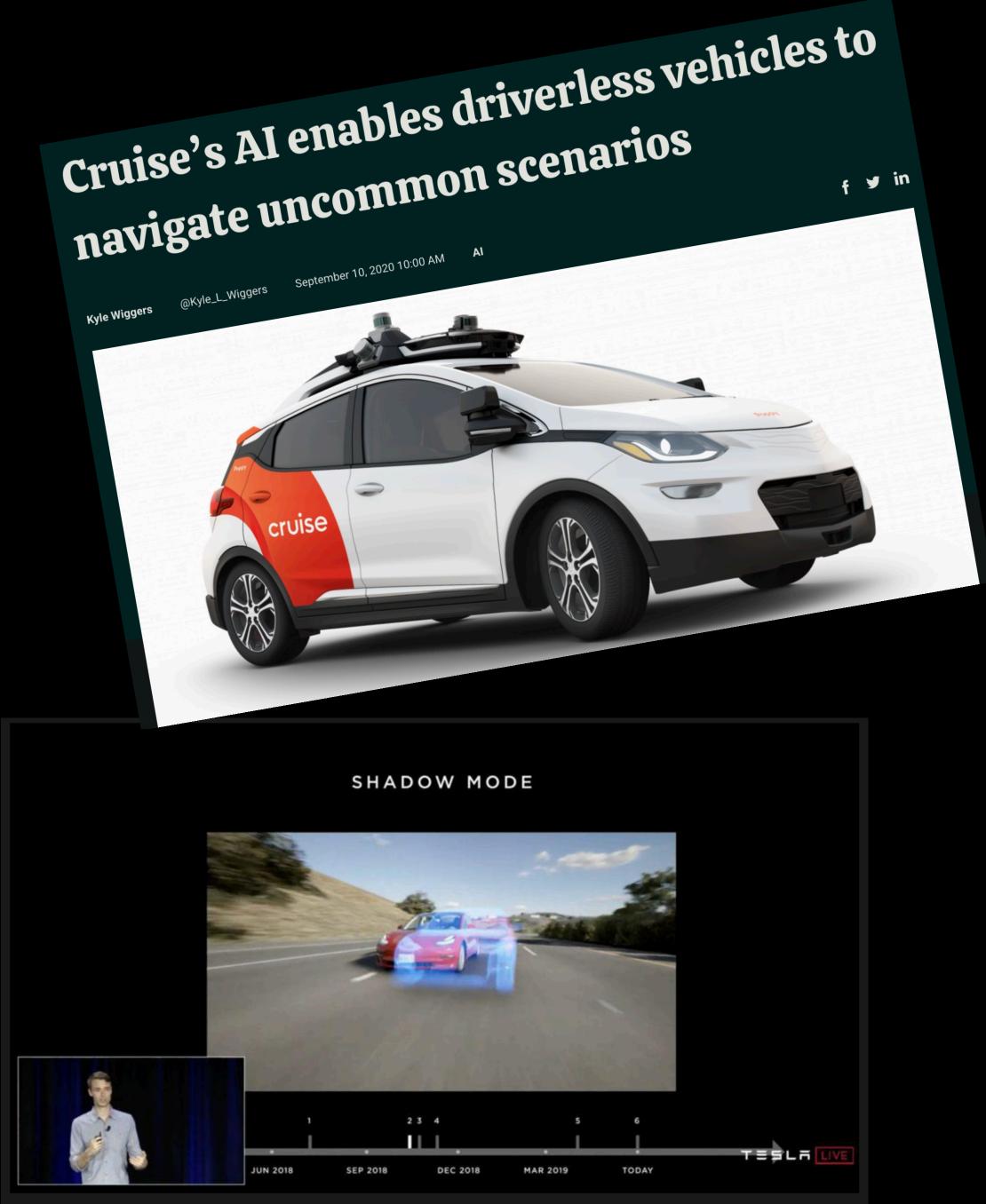
Paolo Tonella





Cruise uses Continual Learning to predict intent during on-road driving

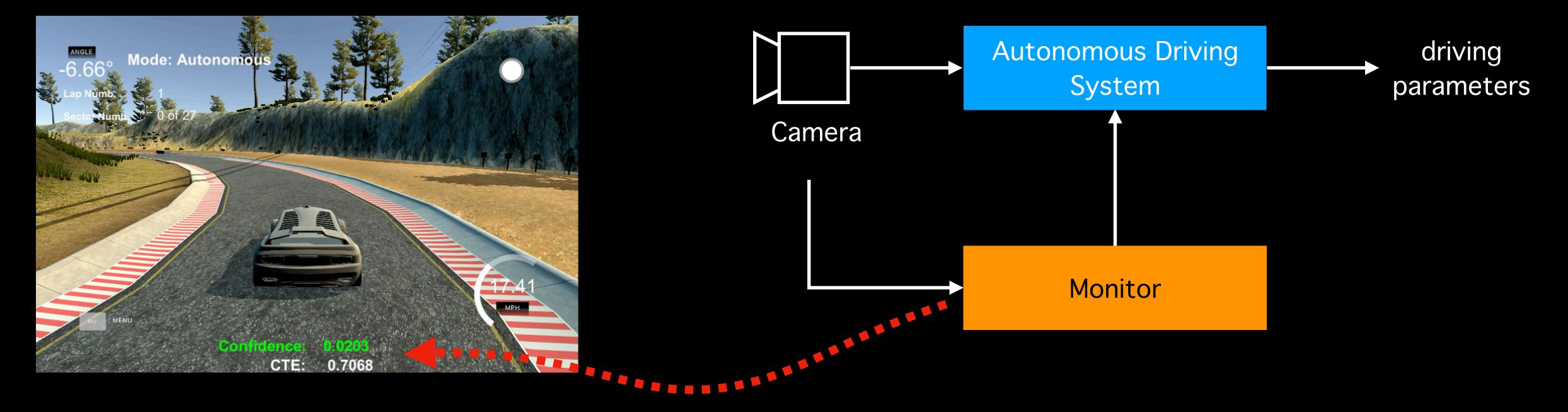
Tesla continually records in-field data to run shadow testing of their autopilot



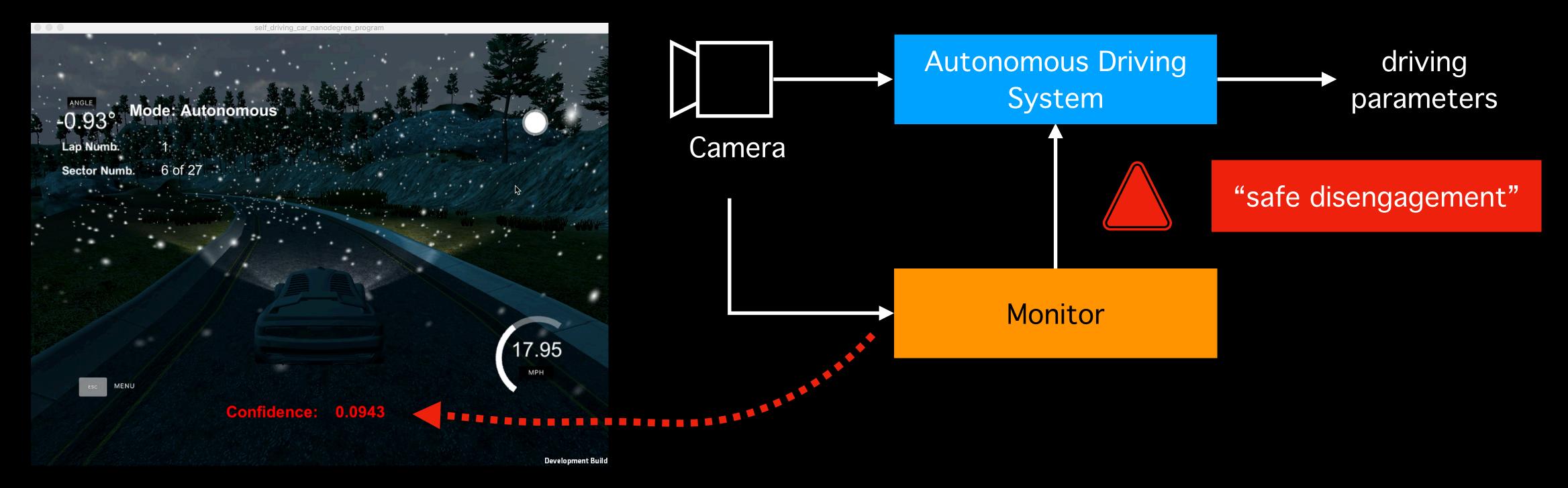
Tesla visualizes their "shadow mode" where the test software drives the car on a different path than... [+]

TESLA PRESENTATION

Sunny (nominal conditions)



Night + Snow (unseen conditions)



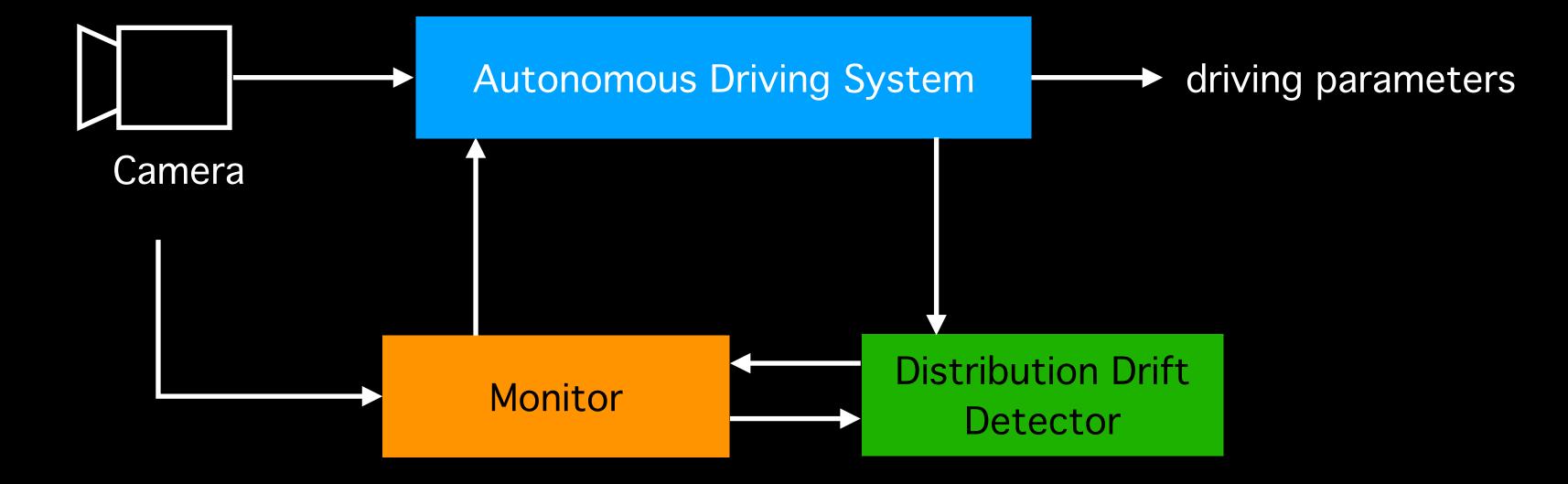


underrepresented inputs (bridge)



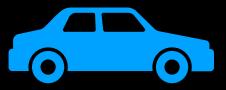
novel classes of data (light rain)

FP = context is nominal && anomaly detector raises an alarm



Can we use in-field data to train a better monitor





Dave-2





Monitor (SelfOracle)



Distribution Drift Detector

Predictive Uncertainty (Unc)

Lateral Position (CTE)

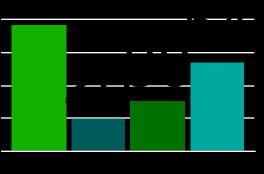
LFP = Unc/CTE is good &&
anomaly detector raises an alarm

Two Experiments

Novelty Detection



Class Imbalance



Metric

% detected LFP





Dave-2





Monitor (SelfOracle)



Distribution Drift Detector

Predictive Uncertainty (Unc)

Lateral Position (CTE)

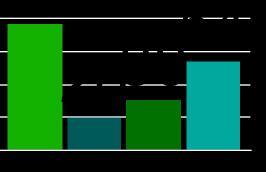
LFP = Unc/CTE is good &&
anomaly detector raises an alarm

Two Experiments

Novelty Detection



Class Imbalance



Metric

% detected LFP



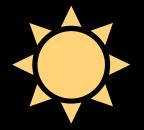


Dave-2





Monitor (SelfOracle)



Distribution Drift Detector

Predictive Uncertainty (Unc)

Lateral Position (CTE)

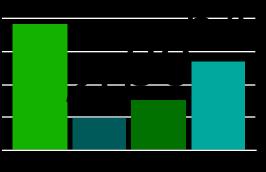
LFP = Unc/CTE is good &&
anomaly detector raises an alarm

Two Experiments

Novelty Detection



Class Imbalance



Metric

% detected LFP



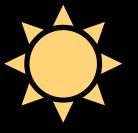


Dave-2





Monitor (SelfOracle)



Distribution Drift Detector

Predictive Uncertainty (Unc)

Lateral Position (CTE)

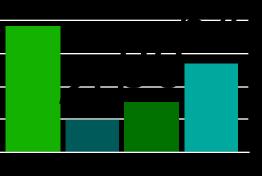
LFP = Unc/CTE is good &&
anomaly detector raises an alarm

Two Experiments

Novelty Detection

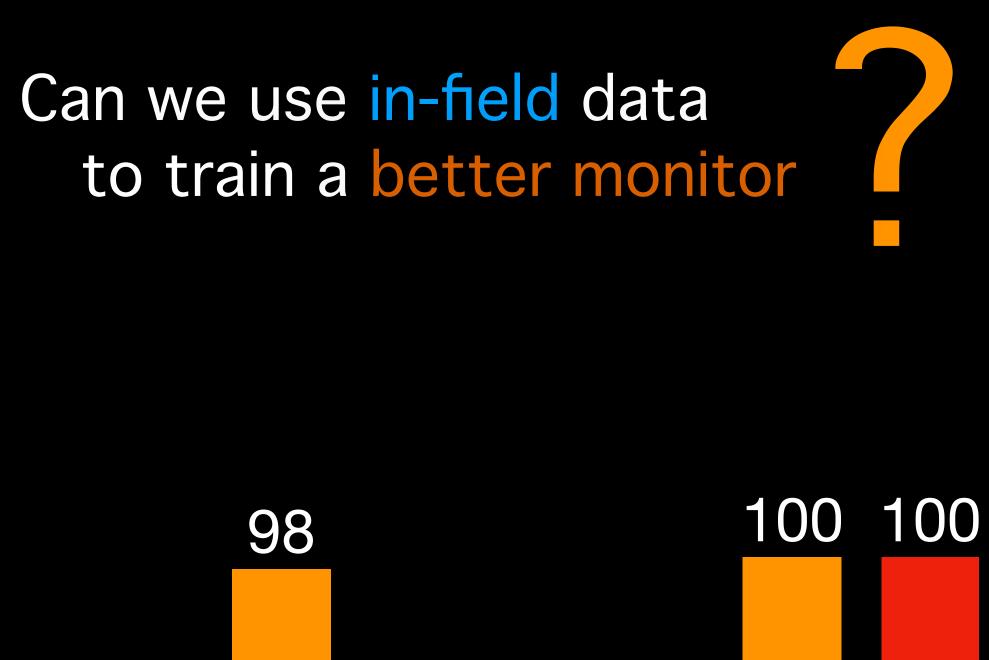


Class Imbalance



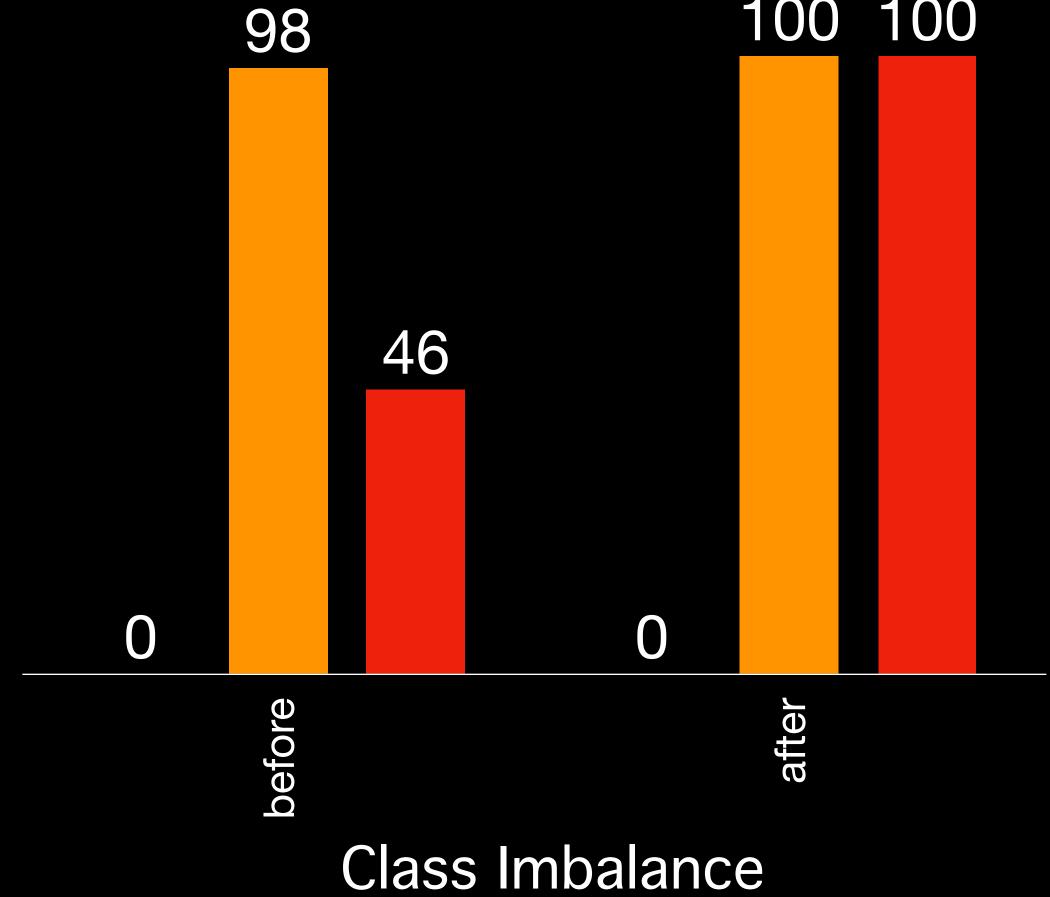
Metric

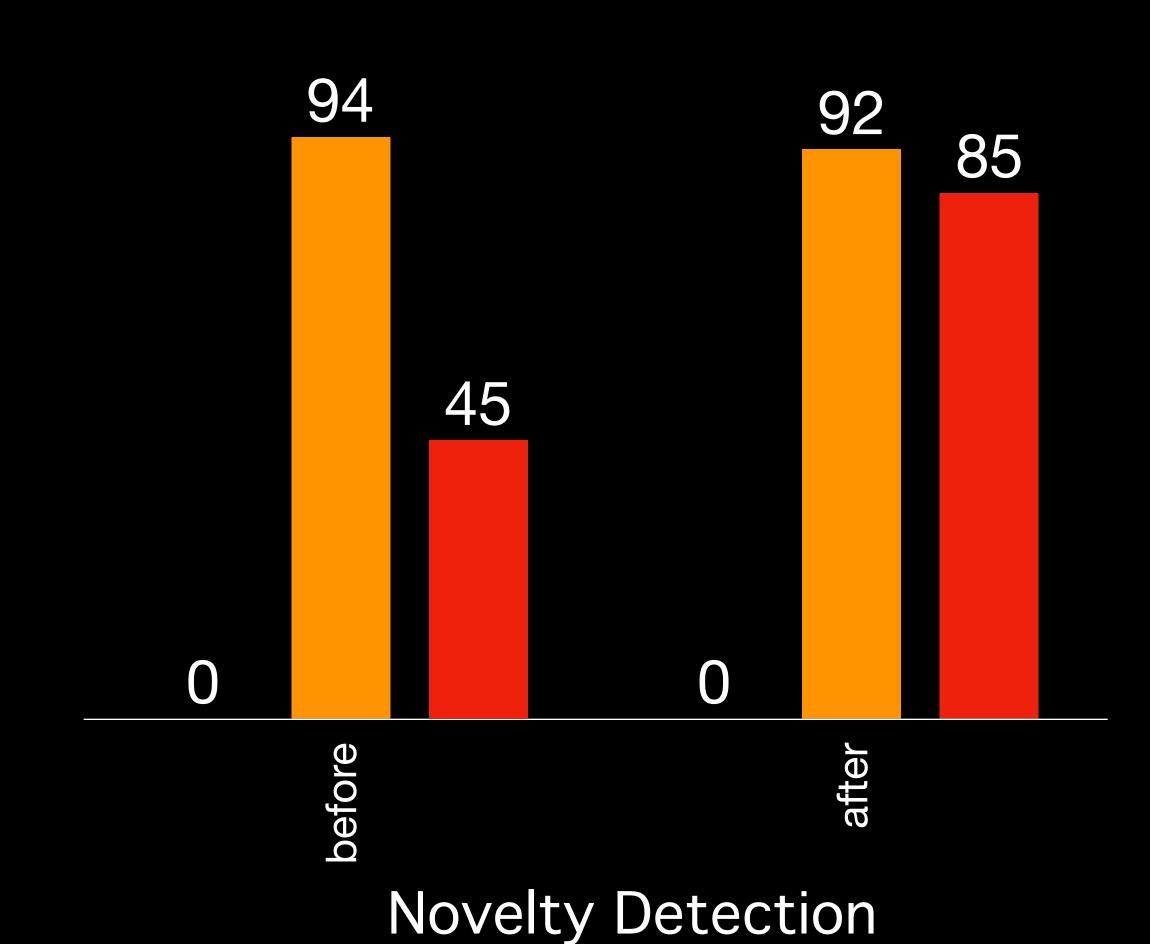
% detected LFP



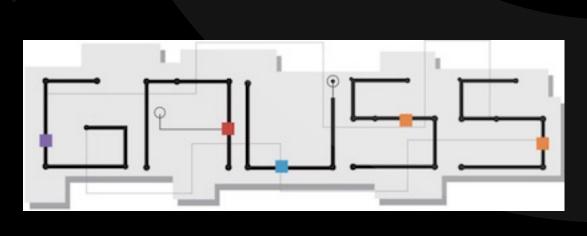








Towards Anomaly Detectors that Learn Continuously











Paolo Tonella

