# Video Analytics in Smart Transportation for the AIC'18 Challenge



Al City Challenge CVPR 2018 Workshop

https://www.aicitychallenge.org/

# Ming-Ching Chang, Yi Wei, Nenghui Song, Siwei Lyu

Contact: mchang2@albany.edu University at Albany, State University of New York, NY, USA



### Introduction

### **Motivation:**

- · Ubiquitous street cameras generate enormous amount of data.
- · Al technologies are readily applicable to help analyze real-world videos.
- · Automatic traffic data analysis can improve congestion control, safety, accident recovery, and transportation infrastructure planning.



### **Summary:**

- . In Track 1 challenge, we demonstrated automatic traffic flow analysis using vehicle detection and tracking with robust speed estimation.
- In Track 2 challenge, we developed a reliable anomaly detection pipeline that can recognize abnormal stalled vehicles with precise location and time.
- In Track 3 challenge, we presented early results of vehicle re-identification by combining vehicle detection/tracking with the deep triplet-loss features that can match vehicles across 4 sites from long hours of videos.

### AlCity'18 Challenge

### **Datasets**

- · 35+ hours of street intersection and highway data from CA and lowa
- . Track 1: 4 CA locations, 27 minutes of videos
- Track 2: 20+ Iowa locations, 25 hours of all-weather, day/night videos
- . Track 3: 4 CA locations, 15 hours of videos



### Evaluation

- Track 1: S1 = DR \* (1 NRMSE of speed)
- Track 2: S2 = F1 \* (1 NRMSE of stall time)
- Track 3: S3 = 0.5 \* (Target Detection Rate + Localization Precision)

### **Vehicle Detection**

- model from Google Object Detection API.
- DETRAC dataset.



Background subtraction + averaging

(1) Flow Density Estimation

### Track 1. Traffic Flow Speed Analysis

### Camera Calibration



**Vehicle Tracks with Speed Estimation** 

Camera PMat

Ground plane projection

**Online Tracking** Ground plane tracking using confidence-aware Kalman filtering

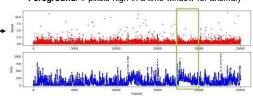
Offline speed denoising & post-processing

# **Track 2. Anomaly Detection**

### (3) Patch-based Feature Extraction



Speed: optical flow low in a time window for anomaly Foreground: # pixels high in a time window for anomaly



### (2) ROI Extraction

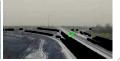




## (4) Thresholding & Cleaning for Anomaly Output







### Track 3. Multi-Camera Vehicle Re-identification

### **Assumption**

· Vehicles of interest must appear at all 4 sites at least once.

### **Pipeline**

- · Vehicle detection & tracking (by applying Track 1 method).
- Extract deep triplet loss features for re-identification.
- · Pairwise matching across videos and sites and keep top 200 matching pairs.
- · Extend the matching pairs across three sites, keep 300 triplets of matches.
- Extend the matching pairs across all four sites, fine select top 100 quadruplets as output.



### Conclusions

- Track1. Real-time online traffic speed
- Track2. Abnormal stalled vehicle detection and alert. Results rank #3 in the AIC'18 challenge.
- Track3. Large-scale, multi-camera vehicle detection, tracking, and re-identification. Result ranks #4 in the AIC'18 challenge.
- Future work include continue refinement of the methods on live, real-world use cases.



- · Faster R-CNN inception ResNet v2 atrous
- · Trained on COCO dataset, fine-tuned on UA-

# · New results using Mask-RCNN.