



Challenges in Time-Stamp Aware Anomaly

Detection in Traffic Videos

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Computer Science & Engineering





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Introduction and Motivation

- The main intension of this paper is to find Timestamp aware anomaly detection in traffic videos.
- Anomalies can be due to car crashes or stalled vehicles.
- Now a days, it became very difficult to know that an accident has occurred and to locate the













Track-1. City-scale multi-camera vehicle tracking



Track-2. City-scale multi-camera vehicle re-identification



Track-3. Traffic Anomaly Detection





Track 3 Traffic Anomaly Detection Dataset



Train Videos: 100 videos (approx. 15 min duration each) Test Videos : 100 videos (approx. 15 min duration each)













• To detect stalled vehicle there are multiple challenges with multi view cameras they provided data



Illumination problem





Cars in parking area

Patch Problem



Slow vehicle problem







Construction vehicle background



Anomaly detection Framework



Output

















Object Detection Model

Algorithm

Anomaly Time stamp



Background Model



The proposed deep background estimation network



Object Detection Model



The proposed one-stage object detector for anomalous object localization and classification.





Algorithm 1

Input: Vehicle detection response in background image. *Vid* contains the set of normal (no detection) and abnormal (some detection) label of a video.

L: length (*Vid*) N(Win_X): Frequency of normal instances in Win_X A(Win_X): Frequency of abnormal instances in Win_X

Output:

end

Step1: for i in L Win_10 = Vid [i-5:i+5] If (N(Win_10)>A(Win_10)) Vid [i]=normal end end Step2: for i in L Win_20 = Vid [i:i+20] if (N(Win_20)<5) Vid [i:i+20]=abnormal elif (A(Win_20)<5) Vid [i:i+20]=normal end</pre>

Step3:

```
steps:
for i in L
    Win_5 = Vid [i:i+5]
    if (N(Win_5)==1)
        Vid [i:i+5]=abnormal
    elif (A(Win_5)==1)
        Vid [i:i+5] = normal
    end
end
Initial Anomaly Timestamp:
for i in L
    if (Vid [i]==abnormal)
        InitialTime = i*3.3 sec
        Break;
    end
end
```





Qualitative and Quantitative Analysis

- Our method achieved 0.2641 S3-score on track-3 test videos of NVIDIA AI city challenge.
- It achieved 0.3838 F1-score and 93.61 RMSE respectively. T
- he lowest S3-score is 0.0162





Qualitative and Quantitative Analysis



Sample Correct Results





Qualitative and Quantitative Analysis



Sample False Positive Results





- This paper presents a 3-stage pipeline for time-stamp aware anomaly detection in road/traffic videos..
- A two-stage method was proposed consisting of deep background modelling and one stage object detection, An algorithm for post processing was proposed to remove temporally inconsistent false positives to certain degree.
- An intuitive approach to anomaly detection was proposed and the challenges to solve the problem of NVIDIA AI city challenge track-3 were discussed.





Thank

