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Vehicle Density Estimation Using Density-Aware Attention Network for Traffic Surveillance System

The surveillance system is widely deployed in several applications for observing the characteristic of target density because it has an advantage in low installation and maintenance costs. In a traffic surveillance system, this characteristic is usually applied in a traffic light control system (TLCS) from the number of vehicles. It helps to control the period of traffic lights and prevent any traffic accidents from vehicles. Besides, the traffic characteristic is useful information for road users to avoid crowded regions, including vehicles, people, and others. Since this system utilizes computer vision-based techniques, it is sensitive to outdoor environments (e.g., lighting conditions, traffic viewpoints, and so on). In addition, hardware with low performance usually is deployed in the surveillance system. It means that the system should be compact and requires low computational cost for operation. Therefore, density estimation cannot be operated properly under any circumstances without concerning these conditions.

Recently, there are several related studies on density estimation. It is well known that density is originally estimated by detecting and counting their regions in the input image. The counting target regions were classified by their visual features (appearance and motion features). Instead of relying on a detection-based approach, recent studies focused on end-to-end learning methods relied on regression-based approaches. These approaches concern holistic features which are visual features extracted from whole input images predicted by a prediction model. The holistic features are utilized for mapping dense regions into density maps which represent the change of object density, where their ground truth was prepared by a convolution between Gaussian distribution and target coordinates corresponding to the target sizes. Nowadays, the task of vehicle counting via a regression-based approach is riddled with a scale-aware model to handle scaling problems. Prediction models are usually designed by several stacked CNNs in parallel, which are called multi-column network architectures. In short, one stack of CNN is used to predict density maps on a specific size of the vehicle. However, these techniques have problems related to practical application as follows:

- With the variation of camera viewpoints, the target sizes can be calculated accurately resulting in the misclassification for density map formulation in the ground truth of an estimation network.
- Recent regression models with multi-column network architectures have a large number of model parameters and high computational costs using several network architectures for various target sizes.

The purpose of this study is to examine the only one stack of CNN, which is called a single-column network architecture, and reduce computational costs and model parameters while keeping similar counting accuracy to multi-column network architectures. It also should not suffer from a scaling problem by avoiding the utilization of target sizes. It found that the target size can be categorized by their local densities which are related to distances of their neighbor target for preparing the ground truth. The proposed prediction model chooses to investigate the connections

between feature maps from different layers, where holistic features of small and large objects can be extracted from feature maps in shallow and deep layers, respectively. The connection can be done by skip connections to integrate feature maps from shallow layers with another in deeper layers. This process, which is called forward connections, can recover the holistic features of small objects. On the other hand, the backward connections are introduced by extracting feature maps from deep layers to combine with the shallower layer. It can be expected that information on a deep layer can help to optimize the shallow layer, where the performance in an earlier stage can be improved. Considering the quality of a density map, feature maps in every layer should have the same resolution to prevent information loss from adjusting their resolutions. Then, pooling layers are replaced with a dilated convolutional layer. Therefore, the contributions of this dissertation can be summarized as follows:

- Instead of relying on the target size, the proposed density map utilizes average distances among target samples where it is designed for visualizing the difference pattern of various vehicle densities (high and low density regions).
- To reduce computation cost, a single column network architecture for vehicle density estimation is designed by including skip connections and dilated convolutions to extract holistic features from intermediate convolutional layers and keep semantic information for density map estimation, respectively.

Since vehicle density estimation is mainly focused on this research, all models are evaluated by the common criteria for vehicle density accuracy. The state-of-the-art of regression-based approach was implemented for comparison. In addition, well-known CNNs with skip connections (e.g., U-net, Resnet, and Densenet) were also applied for analysis. The empirical results show that the proposed prediction model with backward connections achieved a vehicle density accuracy 92.47 % which is close to accuracy of state-of-the-art (93.33 %). From this result, the achievement is summarized as follows:

- In the density map configuration, the target size estimated by an average distance of the target is insensitive to camera viewpoints.
- From the point of view of counting accuracy, the proposed method with a single-column network achieves a promising result which is close to the vehicle counting accuracy from a multi-column network or network with a large number of model parameters.

Moreover, the proposed network satisfies with the minimum requirement of TLCS and it is effective to reduce under-counting errors. However, there is a room for improvement in other datasets consisting of overlapping target regions or crowd counting datasets.

Keyword: surveillance system, vehicle density estimation, regression model, skip connection, dilated convolution, traffic analysis.