Tenancy Status Identification of Parking Slots Using Mobile Net Binary Classifier

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Abstract: The inefficiency in accessing the tenancy status of the parking slots is mainly due to the results of irregular parking regulation/management. The effective parking management enables to avoid unwanted traffic jams and unnecessary fuel wastages. So an efficient parking is necessary for the developing smart cities that aim for a better way of living. So the paper uses the Mobile-Net Classifier to sort out the tenancy state of the parking slots in the cities to assist a proper parking regulation with better proficiency and perfect management. The Mobile-Net classifiers are a sort of light weight deep neural networks that help in identifying the parking slots available accurately based on the image mined from the live camera that feeds the status of the parking lot continuously. The mechanism put forth detects the patches of images form the live recorded video perfectly and determines the vacant slots. The laid out model was applied in an outdoor parking area to determine the systems effective working on the terms of detection accuracy, false positive and the false negative, true positive and the true negative rates along with the average speed of the in identifying the parking slots. As test case, two different mobile-Net network set up were compared to evaluate the swiftness in processing and the perfectness in detecting.

Keywords: Vacant Parking Slots, Tenancy Status, Light Deep Neural Network, Mobile Net Classifiers, Accuracy and Processing Speed.

1. Introduction

The day to day rapid increase in the number of vehicles on road has in turn led to the unsurprising increase in the vehicular congestion leading to demands in the parking slots. The continuous increase in the demands on the parking slots has further caused an ugly scenarios in the modern cities that are thriving to become smart. This demand has even resulted in cities “having a hard time taming. Specifically in the busy areas, these “disorganized parking additionally paves way to commuter’s woes”. 
Based on the studies appropriately more than five billion of the world’s eight billion population is has chosen their living space in the urban areas. This makes the traffic regulation of the cities significant as well as an urgent requisite. But managing the traffic and eluding the congestion to regulate a proper vehicle movement in the city limit is quiet challenging as the proper parking for the vehicles also has to be taken care. At this juncture it is necessary to add this notion, as the world is thriving to construct the smart cities the plan of building or implementing a smart parking system cannot be ignored.

1.1. Smart Parking

The latest information’s has revealed that the urban motorist spend nearly twenty to thirty minutes on road looking for the a vacant parking slot as the present parking regulation is ineffective. This kind of irregular and improper parking slot management leads to waste of time and fuel creating loss in the “productivity and the economic opportunity. Large amount of fuel gets wasted while searching the parking slot. This paves way for a parking frame work that is integrated. By this the vehicles could identify the proper parking slots even before they enter the premises. This type of system causes considerable saving in the time and the fuel as the slot where the vehicle has to be parked is known before hand.

Though multiple applications and digitized solutions has been provided by the researchers in assisting the vehicle drivers find the proper parking slot in the city limit and by collecting and delivering a real time parking information. The developed mobile applications helped in integrating the intelligent transport system with the parking regulation to manage the traffic flow properly. Yet the many conventional system proved to be time consuming and very slow in capturing the information’s about the vacant positions once again leading back to poor regulation of parking.

The conventional system faces complexity in understanding the patterns of the vacant slots in the parking spots. In order to have an proper remedy for the problems prevailing in the parking slots and to have a better management of resources many researchers have up with multitude of solutions such as the vision based sensors, “image recognition algorithm with the CNN that delivers a “state of art performances over the challenges incurred in the image net”, and many more higher complicated and deeper networks to improve the accuracy and the minimize the cost. Yet the real time tenancy status detection for the parking slots demands for the “image classification chores” to be performed on the embedded resource limited devices.

So the paper uses the Mobile-Net Classifier to sort out the tenancy state of the parking slots in the cities to assist a proper parking regulation with better proficiency and perfect management. The Mobile-Net classifiers are a sort of light weight deep neural networks that help in identifying the parking slots available accurately based on the image mined from the live camera that feeds the status of the parking lot continuously. The mechanism put forth detects the patches of images form the live recorded video perfectly and determines the vacant slots. The laid out model was applied in an outdoor parking area to determine the systems effective working on the terms of detection accuracy, false positive and the false negative, true positive and the true negative rates along with the average speed of the in identifying the parking slots. As test case, two different mobile-Net network set up were compared to evaluate the swiftness in processing and the perfectness in detecting.

The work put forward is arranged with the related works in section 2, proposed Mobile-Net based parking slot identification in section 3 results and discussion in section 4 and conclusion in section 5.
2. Related Works

Many researchers have up with multitude of solutions such as the vision based sensors, “image recognition algorithm with the CNN that delivers a “state of art performances over the challenges incurred in the image net”, and many more higher complicated and deeper networks to improve the accuracy and the minimize the cost. The section presents few previous findings to regulate the parking and resolve the problem caused due to improper parking. Suhr et al [1] in his paper has devised sensor fused system to monitor the vacant spaces in the parking. Lee et al [2] integrated the clustering technology in the three sixty degree tracking device to sort out the vacant spaces. Li et al [3] devised a standard, strategy that relies on training to identify the empty space using the machine vision and Zhang et al [4] made changes in the above mentioned system by engaging the deep convolutional neural networks.

Fan et al [5] performed the "Line Filter-Based Parking Slot Detection for Intelligent Parking Assistance System," Scheunert, et al [6] put forth the “Free space determination for parking slots using a 3D PMD sensor.” Li, et al [7] engaged the AVM device to automatically identify the left out empty spaces in the parking areas.Xu, et al [8] conducted the "Vision-guided automatic parking for smart car." Wu et al [9] presents the."PSDet: Efficient and Universal Parking Slot Detection."

Khan et al [10] put forward a Deep-Learning Based Vehicle Count and Free Parking Slot Detection System." and Zinelli, et al [11] devised an "A deep-learning approach for parking slot detection on surround-view images." Ahn et al [12] Duong et al [13] Hong et al [14] and Li et al [15] utilized the light weight based on the cascade- residual-paradigm, deep neural networks in the recognizing the faces and detecting the objects specifically for the devices with limited resources availability, the performances of the light weight network was better in the real time object detection.

Phan, et al [16] elaborates the utilization of the mobile binary network in distinguishing the images. The architecture of the mobile binary net used in classifying the images is depicted below in the figure.1

![Figure 1: MOBILENET Architecture in Image Classification](image)

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Ilhan, et al. [17] devised "A fully automated hybrid human sperm detection and classification system based on mobile-net and the performance comparison with conventional methods." Pandian, M. D. et al. [18] put forth the Sleep Pattern Analysis and Improvement Using Artificial Intelligence and Music Therapy. Journal of Artificial Intelligence. Bashar, A. et al. [19] devised a “Survey on Evolving Deep Learning Neural Network Architectures. Journal of Artificial Intelligence,” Bindhu, Vet al. [20] put forth the "Biomedical Image Analysis Using Semantic Segmentation."

Kumar, T. Senthil et al. [21] devised the. "A Novel Method for HDR Video Encoding, Compression and Quality Evaluation." Koresh, M. H. J. D et al. [22] proposed the “Computer Vision Based Traffic Sign Sensing for Smart Transport” from the papers surveyed using the related works it was found that the deep neural network based architectures demands bigger dimension frame works to attain a adequate small network to run on the platforms with the resource limitations. Providentially the mobile net binary classifiers that are a light weight deep neural networks could hypothetically satisfy the demands of the applications that requires low-latency. The work laid out in the paper puts forward the significance of the MOBILENET classifier in identify the tenancy status of the parking area in the outdoor parking spaces and further provides the “accuracy-speed” trade off various setup of the MOBILENET classifiers and tests the application of the classifiers on identifying the empty spaces in the parking areas for tenancy status identification for empty parking spaces with less delay on the embedded device.

3. Proposed Work

The system model is encompassed with three level of process to identify the tenancy status of the parking areas. The figure .2 shows the various unit encompassed in the system for assisting the parking identification.

![Figure 2 System Model](image)

The detection unit is comprised with the Internet protocol camera that is capable of receiving regulation information as well as sending images through the internet. These cameras are most commonly engaged in the surveillance, yet unlike closed circuit television cameras they don’t demand for a local recording device but for LAN connection to connect to the nearby base station and function as the “detection unit” in the level one process. The figure.3 depicts the architecture of the proposed parking slot tenancy status detection.
The images gathered with the help of the detection unit is processed in the extractor unit to extricate the patches of the images of the separate spaces in the parking areas according to the co-ordinates of the every parking space specified by the annotation file. The image patches acquired are physically segregated as vacant and non-vacant, and are used in training, the convolutional neural network using the MOBILENET Binary classifier, to acquire a framework trained to distinguish the images. The lightweight structure of the MOBILENET Binary classifier makes it compatible for the real-time appliances. The output of the classification framework in the form of “optimally adjusted weight values” is fed into the base station to enable the identification of the vacant spaces in the parking areas.

The MOBILENET Binary classifiers uses the supplementary “point wise convolution” at the three stages following the input, they are the pre-block, Mid-block and the Post-block as shown in figure.1. The block construction heightens the skip connectors allowing the separable layers in the CNN to preserve the original features. The “ReLU is replace with the PReLU to provide the “Convergence Stability.” The estimation of flow in every segment of the convolution starts with the “$INPUT \rightarrow Binary \ Activation \rightarrow Binary \ Convolution \rightarrow \text{PReLU} \rightarrow \text{Batch Normalization} \rightarrow \text{Output}$.” The pooling layers in the MOBILENET binary classifier in order to gradually shrink the spatial dimensions via network. The system prefers an average pooling to provide better empirical results [16].

The base Station simultaneously repossesses the images of the real lively parking spaces from the cameras fixed and affords to deliver a real-time tenancy status identification of the parking slots. For every images gathered the images patches of the vacant and the non-vacant spaces in the parking area are cropped and concurrently the classification between the vacant and the non-vacant is sorted out using the patches of images as input to the MOBILE NET binary classifier that incorporates the “preloaded image segregation model.”

4. Results and Discussion

4.1. Implementation

The system is encompassed with the “FOSCAM FI9800P” internet protocol camera mounted on the lamp posts in the parking area. The training images were gathered on the duration of three to four weeks, 24/7 from an outdoor parking area.
area with 50 layouts. The system uses the “label MG” as the annotating file to define the area of the separate parking areas. The patches of the images of fifty parking space are extricated using the OPENCV and are distinguished as vacant and non-vacant. More than 5K real life images are used in training the classifier to yield better results. The tensor flow does the training operation.

The “accuracy-speed” trade off various setup of the MOBILENET binary classifiers (baseline and the minimal) is done by fixing the width and the resolution multiplier to its corresponding minimal and maximal values. A uniform pruning of width multiplier is performed and the resolution multiplier is fixed base on the resolution of the input image fed. The two diverse MOBILENET binary classifiers are trained at the learning rate of .01 % with the number of iterations = 3000. The proposed work engages the “ASUS tinker board as base station and the embedded computer with Quad Processor (1.8GHz) and 2 GB RAM. The DBOS (TINKER) is installed in the Tensor module compatible with the ARM. The MOBILENET Binary classifier is fed as application written in python into the tensor flow frame work. The JSON format is utilized to display the tenancy status of the spaces in the parking area.

4.2 Result Analysis

The qualitative analysis of the proposed framework is performed by determining the performance metric like false positive, false negative, true positive and true negative. Using the performance metric the average accuracy of the detection is determined as shown in equation 1

\[
Detection \ Accuracy = \frac{True \ positive + True \ negative}{True \ positive + True \ negative + False \ positive + False \ negative}
\]  

(1)

The table 1 below depicts the identification accuracy of the two MOBILENET Binary Classifiers and also present the false positive ratio and the false negative ratio of the proposed model for the testing images acquired through the IP camera for a week.
Table 1: Identification Accuracy, False Positive Rate, False Negative Rate

The figure represents the time taken by the two diverse MOBILENET Binary Classifiers to classify the patches of images. The proposed model affords to deliver the tenancy status within 2.0 seconds.

Figure 4: Time Utilized.

As a summary from the results observed, minimal MOBILENET Binary Classifiers prove to be the choicest framework for reflecting the results of identification incurring a less delay, without compromising on the accuracy in identification, the result observed in table 1 shows the accuracy of the same. So the frame put forward in the paper is capable of delivering accurate results with speedy processing. Moreover, the proposed system affords to deliver low-cost solutions. Compared to other standard approaches, unlike convolutional neural network methods, the MOBILENET Binary classifier does not demand a manual network customization.

5. Conclusion

The effective parking management enables to avoid unwanted traffic jams and unnecessary fuel wastages. So an efficient parking is necessary for the developing smart cities that aim for a better way of living. So the paper uses the Mobile-Net Classifier to sort out the tenancy state of the parking slots in the cities to assist a proper parking regulation with better proficiency and perfect management. The Mobile-Net classifiers are a sort of light weight deep neural networks that help in identifying the parking slots available accurately based on the image mined from the live camera that feeds the status of the parking lot continuously. The mechanism put forth detects the patches of images form the live recorded video perfectly and determines the vacant slots. The laid out model was applied in an outdoor parking area to determine the systems effective working on the terms of detection accuracy, false positive and the false negative, true positive and the true negative rates along with the average speed of the in identifying the parking slots.
Based on the evaluation of “accuracy-speed” trade off various setup of the MOBILENET binary classifiers (baseline and the minimal) the proposed system proved to deliver a low cost solution with the better accuracy and speedy processing.

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