

## **Parking Occupancy Detection from CCTV Images**

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GEOM90038 - Advanced Imaging

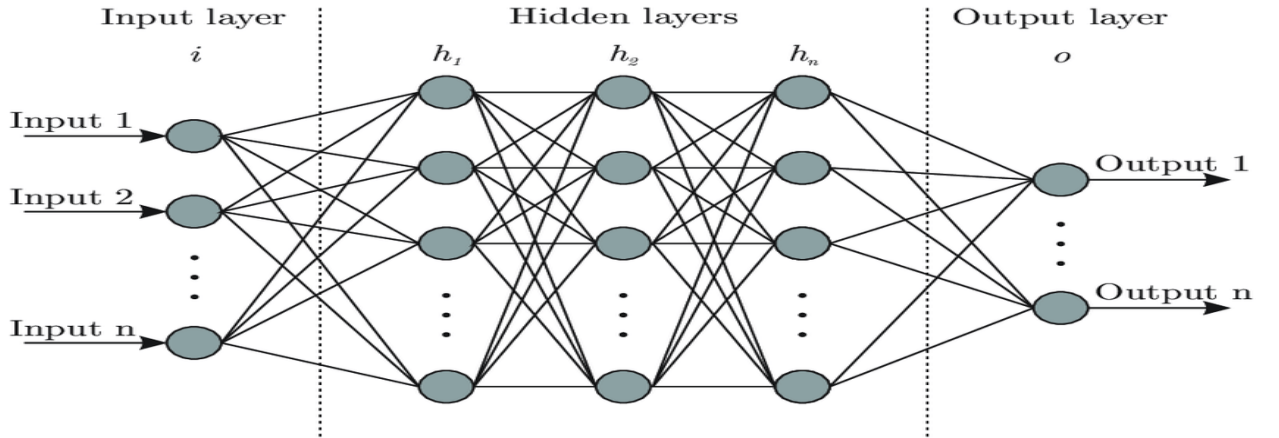
# Parking Occupancy Detection from CCTV Images

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## Introduction

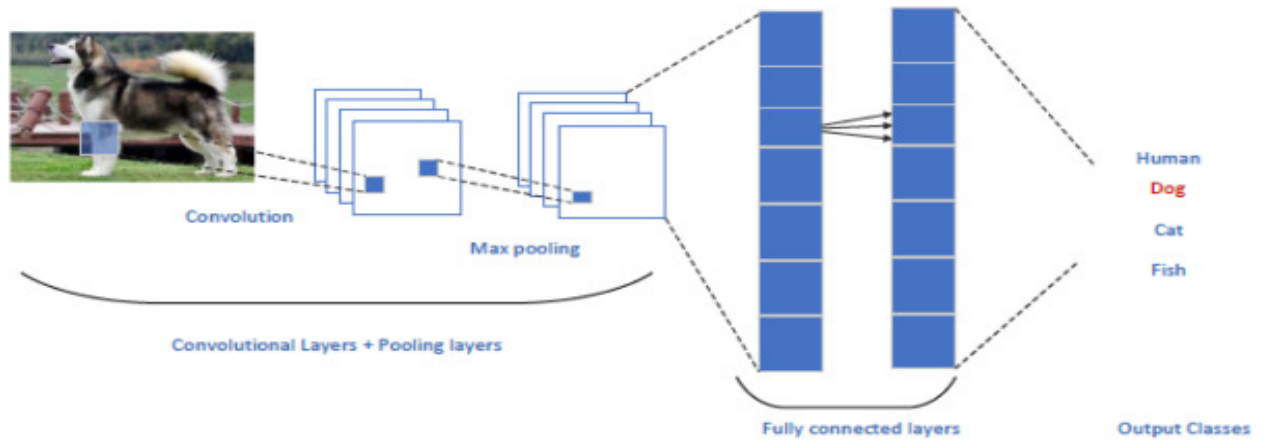
Recent developments in data has undergone significant changes from data processing into machine learning due to increased volumes of readily available data (Khan and Al-Habsi (2020)). The introduction of image-based neural network architecture has allowed the field of computer vision to flourish. Computer vision leverages the power of machine learning to derive meaningful information from visual data, which allows them to take actions and recommendations when they detect issues. Powerful advances in machine learning such as the widely popular transformer model that Chat-GPT is based on has also been adapted to be trained on visual information (El-Nouby et al. (2021)). Figure 1 provides a representation of a neural network, which consists of multiple layers. The input layer is the initial data, and hidden layers calculate the weights of each of these data and propagates them forward to other hidden layers. Finally, the output layer provides probabilities of the desired output (e.g., some classification label).



**Figure 1**

*Visual Representation of Layers That Makes Up a Neural Network (Shukla (2019))*

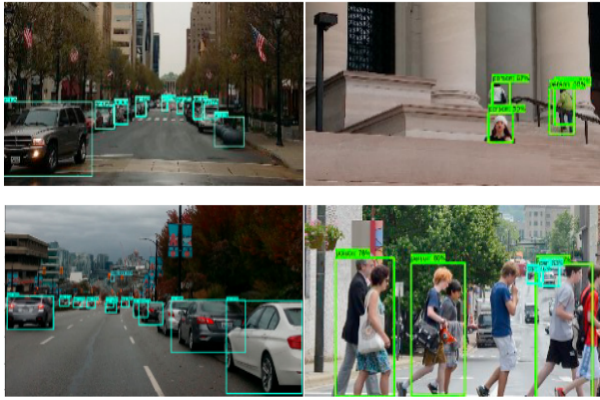
Figure 2 provides a simplified neural network architecture for visual data. The visual data that is placed into the input layer depends on the specific architecture and design. In the case of figure 2, a section of the image is placed into the input layer. It is then propagated forward towards additional layers before finally obtaining the results of the output layer, which in this case, attempts to classify the image into four possible species.



**Figure 2**

*Sample Architecture of Machine Learning Being Used in Computer Vision (Chai et al. (2021))*

The use of computer vision has been applied to wide range of different fields. For example, Esteva et al. (2021) discussed the way medical imaging applications in multiple medical areas such as pathology and dermatology has enhanced the level of care provided. Another field in which machine learning has enhanced computer vision is in the automation and digitization of fruit quality measuring and maintainence (Rathnayake P et al. (2022)). Figure 3 shows how computer vision can be used to detect cars and humans in images. These detections are generated post-training, where multiple images are provided for training the neural network. The trained neural network can then be given new images or even real time video to detect what they were trained to. However, as seen in figure 3, such neural networks are not perfect, and can misclassify or not detect parts of the image that they are supposed to (e.g., the model did not successfully classify one of the humans walking).



**Figure 3**

*Sample Image Detection Results of Cars and People Using Deep Learning and RCNN Models(Khan and Al-Habsi (2020))*

In this report, I aim to

## Methods and Results

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Automatic Delineation of Parking Spaces

Evaluation

Improving Accuracy

## Discussion

Accuracy Evaluation

Improvement Based on Assumptions

Challenges and Shortcomings

Scopes of Improvement

## Conclusions and Future Directions

## Appendix

## References

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