

PBEL Virtual Internship

Project Title: Image Classification of Cats and Dogs using CNN

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Declaration

I hereby declare that this project report titled "Image Classification of Cats and Dogs using CNN" is a result of my own work carried out during the IBM PBEL Virtual Internship. The project is original, and no part of it has been copied or submitted elsewhere for any other course or internship.

Acknowledgement

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1. Introduction

The goal of this project is to develop a **binary image classifier** that can distinguish between cats and dogs. A lightweight pretrained CNN model — **MobileNetV2** — was used to build an efficient and fast classifier. This project also includes a **web-based interface using Gradio**, allowing users to upload or capture images and get instant predictions.

2. Technologies Involved

Technology Purpose

Python Core language

TensorFlow / Keras Deep learning model building

TensorFlow Datasets (tfds) Dataset loading and preprocessing

Google Colab Development and training environment

Gradio UI for model deployment

Matplotlib Visualization of training graphs

MobileNetV2 Pretrained CNN for efficient feature extraction

3. Problem Faced & Solution Implementation

During the development of this project, several challenges were encountered and addressed effectively. One of the major issues was **overfitting**, which started to appear when the model was trained for too many epochs. To resolve this, the number of epochs was limited to 5, and **MobileNetV2's pretrained base was frozen** to prevent unnecessary parameter updates. Another problem was that the model sometimes predicted images that were not cats or dogs as one of the classes. This was mitigated by adding a **confidence threshold** in the prediction logic—if the prediction probability was too close to 50%, the image was marked as "Invalid." Additionally, since Google Colab environments reset after every session, **saving and loading the model to Google Drive** was implemented to retain training progress. Finally, for easier and interactive model deployment, **Gradio** was used to build a clean web-based interface, which simplified testing and user input through webcam or file upload.

4. Dataset Description

• Source: Oxford-IIIT Pet Dataset via TensorFlow Datasets

• Classes: 37 pet breeds

• Simplified to:

 \circ Cat \rightarrow label < 12

 \circ **Dog** \rightarrow label ≥ 12

• **Split**: 80% Training / 20% Validation

5. Model Design

- **Base Model**: MobileNetV2 (frozen)
- Custom Head:
 - o GlobalAveragePooling2D
 - o Dense(128, activation='relu')
 - Dense(1, activation='sigmoid')
- **Optimizer**: Adam
- **Epochs**: 5 (can be increased)
- Saved Model: Google Drive

6. Gradio Interface

- Accepts both webcam and uploaded images
- Resizes, preprocesses, and classifies the image
- Uses a **confidence check** to filter out non-cat/dog inputs
- Output format:
 - o Cat (95.3% confidence)
 - o Dog (88.7% confidence)
 - o Invalid image (not a cat or dog)

7. Conclusion

This project showed how to build a cat vs dog image classifier using a pre-trained model (MobileNetV2). The model was fast and worked well, even on limited hardware. Gradio made it easy for users to try out the model, and saving it to Google Drive helped keep the model for future use.

Through this project, I improved my skills in:

- Using CNNs and transfer learning
- Preprocessing image data with TensorFlow Datasets (tfds)
- Testing and evaluating ML models
- Creating a complete machine learning project from training to deployment