



CS 351 MACHINE LEARNING

Mini Project Assignment Prepared By: Shreyas Bhat (181ME175)
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INTRODUCTION

- Since I had prior experience with Deep Learning I decided to try and write the code from scratch to get a better understanding of the Dataset and the various problems that may arise
- I used Tensorflow 2.x as the library to implement the CNN
- I used the Kaggle Platform to train my model. There are 2 main reasons for this:
 - Kaggle comes with a pre built environment to run python code hence eliminating installation hassles
 - Kaggle boasts a Nvidia P100 GPU which will help reduce training times and let me experiment with model
- After training the inference script was run on my local machine. This was so I can get a taste of solving installation issues while not compromising on speedy training.
- Source Code: <https://github.com/DarthRoco/CS351-ML>

STEPS INVOLVED

- 01 DATASET PREPROCESSING
- 02 INPUT PIPELINE
- 03 MODEL DEFINITION AND TRAINING
- 04 INFERENCE

01 DATASET PREPROCESSING

- Dataset was downloaded from [this link](#). It consists a random subsample of ~5000 images from original dataset
- The labels are originally suited for multi categorical classification. However as per sir's instruction I had to convert it to a binary classification problem
- Thus I used the Pandas library to create a new column with a simple condition based filter on the labels already provided
- As seen in code snippet below if the label was "No Finding" rename to Healthy else rename to "Diseased"
- After which we sort and only take first element (since some rows have multiple labels)

```
df['bin']=df['Finding Labels'].apply(lambda s: ["Diseased" if 1 not
in ['No Finding'] else "Healthy" for 1 in str(s).split('|')])
df['bin']=df['bin'].apply(lambda s:sorted(s) )
df['bin']=df['bin'].apply(lambda s:s[0] )
df.head()
```


03 MODEL DEFINITION AND TRAINING

- For this phase I decided to attempt 2 ways of selecting model.
- In the First Model(Model A) I tried to add a few dense layers after a pretrained EfficientNet V4 backbone and fitted the dataset
- In Model B I used the architecture provided in source and trained entire model from scratch.I chose this model as final since it gave better scores
- The metric used for comparison in both cases was ROC-AUC score which is commonly used for imbalanced classification problems
- The results are summarised below.

MODEL A

- Used concepts of transfer learning
- Backbone consisted of EfficientNet V4
- Trained for ~60 epochs
- Best AUC score:~0.7

MODEL B

- Trained Model from scratch
- Architecture inspired from original source provided by sir
- Trained for ~60 epochs
- Best AUC score:~0.96

04 INFERENCE

```
Anaconda Prompt (Anaconda)

(tf-gpu) D:\Desktop\CSMINOR>python SingleInference.py trial1.png

===
Loading Model
===

===
Model Loaded in 0.57 seconds
===

Prediction of model is:
Diseased, with a sureity of [0.22837567]
Time Taken:0.33 seconds

(tf-gpu) D:\Desktop\CSMINOR>
```

- Once the model was trained I downloaded it as a .h5 file which is native to TensorFlow. With this file I can run the model in inference mode on any device without access to original source.
- I created an Anaconda Environment on my local machine and installed all the necessary libraries
- I wrote a simple python script that takes the path of a single image in the form of a single command line argument as input and predicts whether it is "Diseased" or "Healthy"(threshold 0.5)
- The script loads the image as per path and performs resize and normalisation operation on it. Then loads the model and performs prediction.
- For full source refer [this](#)



THANK YOU

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