

## Model Development Phase Template

|               |  |
|---------------|--|
| Date          | 28 November 2024                                       |
| Team ID       | 739996   |
| Project Title | Deep Fruit Veg: Automated Fruit And Veg Identification |
| Maximum Marks | 10 Marks   |

### Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include a summary and training and validation performance metrics for multiple models, presented through respective screenshots.

### Initial Model Training Code (5 marks):

Paste the screenshot of the model training code

```
class LR_ASK(keras.callbacks.Callback):
    def __init__(self, model, epochs, ask_epoch):
        super(LR_ASK, self).__init__()
        self.set_model(model)
        self.ask = ask_epoch
        self.epochs = epochs
        self.ask = True
        self.lowest_vloss = np.inf
        self.best_weights = self.model.get_weights()
        self.best_epoch = 1
        self.plist = []

    def get_list(self):
        return(self.plist)

    def on_train_begin(self, logs=None):
        if self.ask_epoch==0:
            print('you set ask_epoch=0, ask epoch will be set to 1',flush=True)
            self.ask_epoch=1
        if self.ask_epoch >=self.epochs:
            print('ask_epoch>=epochs, will train for',epochs,'epochs',flush=True)
        if self.epochs==1:
            self.ask=False
        else:
            print('Training will proceed until epoch',ask_epoch,'then you will be asked to')
            print('enter H to halt training or enter an integer for how many more epochs to run then be asked again')
            self.start_time.time()
```

```
def on_train_end(self,logs=None):
    print('loading model with weights from epoch',self.best_epoch)
    self.model.set_weights(self.best_weights)
    tr_duration=time.time()-self.start_time
    hours=tr_duration//3600
    minutes=(tr_duration-(hours*3600))//60
    seconds=tr_duration-((hours*3600)+(minutes*60))
    msg=f'training elapsed time was{str(hours)} hours,{minutes:4.1f} minutes,{seconds:4.2f} seconds'
    print(msg,flush=True)
    def on_epoch_end(self,epoch,logs=None):
        v_loss=logs.get('val_loss')
        if epoch>0:
            deltav=self.lowest_vloss-v_loss
            pimprov=(deltav/self.lowest_vloss)*100
            self.plist.append(pimprov)
        else:
            pimprov=0.0
        if v_loss<self.lowest_vloss:
            self.lowest_vloss=v_loss
            self.best_weights=self.model.get_weights()
            self.best_epoch=epoch+1
            print(f'\n validation loss of {v_loss:7.4f} is {pimprov:7.4f} % below lowest loss, saving weights from epoch {str(epoch+1):3s} as best weights')
        else:
            pimprov=abs(pimprov)
            print(f'\n validation loss of {v_loss:7.4f} is {pimprov:7.4f} % above lowest loss of {self.lowest_vloss:7.4f} keeping weights from epoch{str(self.best_epoch)}')
            if self.ask:
                if epoch+1==self.ask_epoch:
                    print(f'\n validation loss of {v_loss:7.4f} is {pimprov:7.4f} % below lowest loss, saving weights from epoch {str(epoch+1):3s} as best weights')
                else:
                    pimprov=abs(pimprov)
                    print(f'\n validation loss of {v_loss:7.4f} is {pimprov:7.4f} % above lowest loss of {self.lowest_vloss:7.4f} keeping weights from epoch{str(self.best_epoch)}')
                    if self.ask:
                        if epoch+1==self.ask_epoch:
                            print('\n enter H to end training or an integer for the number of additional epochs to run then ask again')
                            ans=input()
                            if ans=='H' or ans=='h' or ans=='0':
                                print('you entered',ans,'training halted on epoch',epoch+1,'due to user input\n',flush=True)
                                self.model.stop_training=True
                            else:
                                self.ask_epoch+=int(ans)
                                if self.ask_epoch>self.epochs:
                                    print('\n you entered maximum number of epochs as',self.epochs,'cannot train for',self.ask_epoch,flush=True)
                                else:
                                    print('You entered ',ans,'Training will continue to epoch',self.ask_epoch,flush=True)
                                    lr=float(tf.keras.backend.get_value(self.model.optimizer.lr))
                                    print(f'current LR is {lr:7.5f}')
                                    ans=input()
                                    if ans=='':
                                        print(f'keeping current LR of {lr:7.5f}')
                                    else:
                                        new_lr=float(ans)
                                        tf.keras.backend.set_value(self.model.optimizer.lr,new_lr)
                                        print('changing LR to',ans)
```

## Model Validation and Evaluation Report (5 marks):

| Model   | Summary  | Training and Validation Performance Metrics  |
|---------|--|--|
| Model 1 | <pre>[ ] epochs ask_epoch= ask_val_epoch= ask_train_epoch= ask_val_epoch= ask_test_epoch=  [ ] Start coding or generate with AI.  [ ] Import tensorflow as tf print(tf.__version__)  [ ] 2.12.4  [ ] model.compile(optimizer=Adamax(learning_rate=0.001),loss='categorical_crossentropy',metrics=['accuracy']) history=model.fit(train_gen,epochs=epochs,verbose=1,callbacks=callbacks,validation_data=validation_gen,validation_steps=validation_steps,shuffle=False,initial_epoch=</pre> | <pre>model.compile(optimizer=Adamax(learning_rate=0.001),loss='categorical_crossentropy',metrics=['accuracy']) history=model.fit(train_gen,epochs=epochs,verbose=1,callbacks=callbacks,validation_data=validation_gen,validation_steps=validation_steps,shuffle=False,initial_epoch=  Epoch 1/3 180/180 [#####] - 2768s 15s/step - loss: 1.7998 - accuracy: 0.5244 - val_loss: 1.5487 - val_accuracy: 0.7956 Epoch 2/3 180/180 [#####] - 2641s 15s/step - loss: 1.5386 - accuracy: 0.8158 - val_loss: 1.3179 - val_accuracy: 0.8778 Epoch 3/3 180/180 [#####] - 2678s 15s/step - loss: 1.2845 - accuracy: 0.8842 - val_loss: 1.1768 - val_accuracy: 0.9133</pre> |

