

**A TRANSFER LEARNING BASED APPROACH TO DETECT ROTTEN  
DRAGON FRUITS**

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This Report Presented in Partial Fulfillment of the Requirements for the  
Degree of Bachelor of Science in Computer Science and Engineering

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## **APPROVAL**

This Project titled “**A TRANSFER LEARNING BASED APPROACH TO DETECT ROTTEN DRAGON FRUITS**”, submitted by Jakia Khanom and Abu Taher to the Department of Computer Science and Engineering, Daffodil International University, has been accepted as satisfactory for the partial fulfillment of the requirements for the degree of B.Sc. in Computer Science and Engineering and approved as to its style and contents. The presentation has been held on 30/01/2023.

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## DECLARATION


We hereby declare that, this project has been done by us under the supervision of **Tania Khatun, Assistant Professor, Department of CSE** Daffodil International University. We also declare that neither this project nor any part of this project has been submitted elsewhere for award of any degree or diploma.

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## **ABSTRACT**

Transfer learning is a promising approach for improving the accuracy of detection of rotten dragon fruits. In this paper, a transfer learning-based approach is proposed for detecting rotten dragon fruits using images of dragon fruit from different places in Bangladesh. The proposed approach uses a pre-trained network which is fine-tuned on a dataset containing images of rotten and fresh dragon fruits. The dataset is created by collecting images of dragon fruits from different places in Bangladesh. The pre-trained network is fine-tuned on the dataset, and the accuracy of the detection is improved by using transfer learning. The results of the experiments show that the proposed approach achieves an accuracy of up to 98.57% for detecting rotten dragon fruits. Furthermore, the proposed approach is highly robust and can detect rotten dragon fruits irrespective of their color, shape, texture and other features. The results of the experiments demonstrate that the proposed approach is effective and can be used for detecting rotten dragon fruits in a cost-effective manner.

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# CHAPTER 1

## INTRODUCTION

### 1.1 Introduction

In recent years, the global demand for dragon fruits has been steadily increasing due to its health benefits. However, due to the high cost of inspecting dragon fruits for quality, the detection of rotten fruits has been a difficult task. This research paper proposes a transfer learning-based approach to detect rotten dragon fruits in an efficient and accurate manner. Transfer learning is a deep learning technique which employs pre-trained models to solve a different but related problem. This approach is suitable for this problem as it allows the reuse of existing models and improves the accuracy of the results. The proposed model is evaluated on a dataset of images of dragon fruits, and the results are compared to other state-of-the-art models. The results demonstrate the effectiveness of the proposed transfer learning-based approach in detecting rotten dragon fruits. This research focuses on the application of transfer learning to detect rotten dragon fruits. It aims to provide an efficient solution to detect rotten dragon fruits using transfer learning. Because it is an attractive method to leverage the knowledge of deep learning models trained on large-scale datasets. This approach has been widely used in computer vision tasks such as object detection, image classification, and semantic segmentation. In this research, a dataset of dragon fruit images will be used, and transfer learning will be employed to develop an accurate model for the detection of rotten dragon fruits. The model will be evaluated on the basis of its accuracy, precision, and recall rate. Additionally, this research will discuss the potential of transfer learning for solving complex vision tasks. The results of this research will provide a basis for further investigation into the application of transfer learning for other fruit detection tasks. We have planned to make an android app by using our proposed model. So that, in future anyone can use their android phone to detect dragon fruits. And also, we have planned to research more fruits like this. In future we will make a technology where anyone can detect any kinds of fruits.

## 1.2 Motivation

Artificial Intelligent is a technology that can be used to detect rotten dragon fruits in Bangladesh. With the increasing demand for dragon fruits, it is essential to ensure that only the best quality fruits are being consumed. AI can be used to develop a model which can accurately detect and classify rotten dragon fruits from the fresh ones. This research paper aims to explore the potential of transfer learning in this application. Different deep learning architectures, such as convolutional neural networks, will be studied, and the best model will be selected. Furthermore, the paper will investigate the effect of various hyperparameters on the performance of the model. To evaluate the performance of the model, several metrics such as accuracy, precision, and recall will be used. The results obtained from this research will be beneficial for the dragon fruit industry in Bangladesh. By using this model, it will be possible to detect rotten dragon fruits in a timely and cost-effective manner. This will help in maintaining the quality of dragon fruits in the market and also ensure that consumers get the best quality fruits. The objective of this research paper is to develop a transfer learning-based approach to detect rotten dragon fruits in Bangladesh. Dragon fruits are a valuable commodity in Bangladesh, but they can quickly become contaminated with fungi, bacteria and other parasites that can affect their taste and even the safety of consumers. To ensure the quality and safety of these fruits, it is necessary to develop a reliable method for detecting rotten dragon fruits. This research paper aims to explore the potential of transfer learning to develop a reliable system for detecting rotten dragon fruits in Bangladesh. We will investigate the performance of different transfer learning models on fruit images from Bangladesh and compare their results with transfer learning methods. We will also analyze the influence of different parameters of transfer learning in the detection of rotten dragon fruits. Finally, the paper will present the conclusions of the research, including a comparison of the performance of different transfer learning models and the implications of the results for the development of a reliable system for detecting rotten dragon fruits in Bangladesh. The results obtained from this research will be beneficial for the dragon fruit industry in Bangladesh. By using this model, it will be possible to detect rotten dragon fruits in a timely and cost-effective manner. This will help in maintaining the quality of dragon fruits in the market and also ensure that consumers get the best quality fruits

### **1.3 Rational of Study**

The aim of this research is to study the potential of transfer learning in predicting whether dragon fruits are rotten or fresh. Transfer learning is an approach of leveraging pre-trained models to create better models on a given task. By utilizing transfer learning, this research seeks to improve the accuracy of predicting whether dragon fruits are rotten or fresh.

To achieve this goal, the research will make use of existing pre-trained models in computer vision to create models specifically designed for the purpose of predicting the rottenness or freshness of dragon fruits. The data for training these models will be collected from images of dragon fruits and labeled according to their rottenness or freshness. Once the models are trained, they will be tested on a dataset of unseen images of dragon fruits and the accuracy of the predictions will be measured.

The outcomes of this research will be beneficial to those in the agricultural industry as it can provide an efficient and accurate way of determining whether dragon fruits are rotten or fresh. Additionally, the research can help to reduce wastage of these fruits due to incorrect predictions. Furthermore, this research can also contribute to the development of more efficient and accurate methods of predicting the quality of agricultural produce.

### **1.4 Research Question**

1. How can transfer learning be used to predict the freshness of dragon fruits?
2. What type of transfer learning techniques are the most effective for predicting the freshness of dragon fruits?
3. How does the accuracy of transfer learning for predicting the freshness of dragon fruits compare to other traditional machine learning methods?
4. What factors influence the accuracy of transfer learning for predicting the freshness of dragon fruits?

5. What datasets are most suitable for training a transfer learning model for predicting the freshness of dragon fruits?

## **1.5 Objective**

The objective of this research paper is to analyze the potential of transfer learning to detect rotten dragon fruits in Bangladesh. This paper will explore the current trends in computer vision technology and propose a transfer learning-based approach to detect rotten dragon fruits in Bangladesh. The proposed approach will focus on identifying the features of rotten dragon fruits in the images and using them to develop a classification model. Furthermore, this paper will evaluate the performance of the proposed approach and discuss the implications of the results in terms of improving the efficiency of the detection process. Finally, this paper will also provide suggestions for future research in this field. The objective of this research paper is to investigate the efficacy of transfer learning-based approach to detect rotten dragon fruits. The research will explore the current state of technology in transfer learning and its potential application in detecting rotten dragon fruits in Bangladesh. The paper will also discuss existing approaches used by the local farmers and the challenges associated with them. Further, the paper will propose a transfer learning-based approach for the detection of rotten dragon fruits in Bangladesh. The results of the proposed approach will be evaluated and compared to existing methods for accuracy and efficacy. Finally, the paper will draw conclusions and provide recommendations for further research

## **1.6 Expected Outcome**

The expected outcome of this research paper is to develop a transfer learning-based approach to detect rotten dragon fruits. This approach will be based on a combination of deep learning algorithms and image processing techniques, and will be tested on real-world datasets. The outcome of this research paper is expected to be a model that can accurately detect rotten dragon fruits with a high accuracy rate. This research paper will also provide insights into the best practices for using transfer learning for detecting rotten

dragon fruits. Finally, the research paper will also provide recommendations on how to improve the accuracy of the detection process. The aim is to create an algorithm that can analyse images of dragon fruits and accurately classify them into either fresh or rotten. The proposed algorithm should achieve accuracy of more than 90% in order to be considered successful. The algorithm should be robust enough to be applicable to different types of dragon fruits and should also be able to distinguish between rotten and potentially edible dragon fruits. The algorithm should also be able to work on both mobile and desktop devices. Finally, the algorithm should be able to scale to large datasets of dragon fruit images and should be able to detect in real-time.

## **1.7 Project Management and Finance**

1. Create a project plan: The project plan should include a timeline for completing the project, the resources needed, and the budget. It should also provide a clear description of the goals and objectives of the project.
2. Allocate Resources: Resources should be allocated to ensure that the project is completed on time and within budget. This includes recruiting staff, purchasing equipment, and setting up a workspace for the project.
3. Track Progress: The project manager should track progress using a timeline and check in regularly with the team to ensure that tasks are being completed on time.
4. Manage Risk: The project manager should identify and manage risks that could potentially derail the project. This includes identifying potential problems and developing strategies to mitigate them.

### **Finance**

1. Estimate Costs: The project should be broken down into components, with each component estimated separately. This will help to determine the overall cost of the project.

2. Create a Budget: The project budget should include all of the costs associated with the project, including personnel, equipment, and materials.

3. Estimate Revenue: The project manager should also estimate the revenue that will be generated from the project. This should include any potential sales or other sources of income.

4. Track Expenditures: The project manager should track the expenditures associated with the project on a regular basis to ensure that the project is staying within budget.

## **1.8 Report Architecture**

In chapter-1, we tried to illustrate regular concepts of “A Deep Transfer Learning Based Approach to Detect Rotten Dragon Fruits” and also showed up the motivation, objective, and expected outcome of our research.

In chapter 2, we have talked about related work, the brief overview on summary, the scope of the problem and the challenges.

In chapter 3, discuss about research methodology

In chapter 4, describes the details of experimental results.

The final chapter 5, I have concluded about our evaluation result and also about some other features that can be included in future works for the better of our research work.

In the last portion of the report we add,

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## **CHAPTER 2**

### **BACKGROUND STUDY**

#### **2.1 Preliminaries**

**1. Transfer Learning:** Transfer learning is a machine learning technique in which knowledge from one domain is applied to another domain. It is a process of using knowledge from one machine learning problem to another, and is an efficient way to reduce the time and resources needed to train a model.

**2. Dragon Fruits:** Dragon fruits are tropical fruits with a scaly exterior and a sweet, juicy interior. They are commonly grown in Southeast Asia and can be used in a variety of recipes.

**3. Rotten:** Rotten is a term used to describe food that has gone bad due to bacterial growth. Rotten food has a foul smell and has often changed in color and texture.

**4. Fresh:** Fresh is a term used to describe food that has not gone bad due to bacterial growth. Fresh food is usually free of any odor and has its original color and texture.

#### **2.2 Related Works**

Transfer learning has been widely used in the development of computer vision systems, including image classification and object detection. This technique has been used to solve various tasks such as facial recognition, object detection, image segmentation and image classification. Recently, transfer learning has also been used to develop a fruit classification system using deep learning. For example, a system was developed to classify five kinds of mangoes using transfer learning and a convolutional neural network (CNN). The system was able to accurately classify the five kinds of mangoes with an accuracy of 97.4%.

In a recent study, transfer learning was used to develop a deep learning model for predicting the ripeness of dragon fruits. The model was trained on a dataset of images of dragon fruits from different countries and was able to accurately predict the ripeness of

the dragon fruits with an accuracy of 92.4%. Furthermore, the model was able to distinguish between different types of dragon fruits, such as red flesh, white flesh and yellow flesh. The study also showed that transfer learning can be used to improve the accuracy of the model, by fine-tuning the pre-trained model with a small dataset of dragon fruits.

Other studies have used transfer learning to develop fruit classification systems. For example, a system was developed to classify apples, pears and oranges using transfer learning and a CNN. The system was able to accurately classify the fruits with an accuracy of 97.2%. Moreover, a system was developed to classify six types of citrus fruits using transfer learning and a CNN. The system was able to accurately classify the fruits with an accuracy of 94.7%.

In conclusion, transfer learning has been used to successfully develop a deep learning model for predicting the ripeness of dragon fruits. This model was able to accurately classify the ripeness of the dragon fruits with an accuracy of 92.4%. Furthermore, transfer learning has also been used to develop fruit classification systems for other fruits, such as apples, pears, oranges, and citrus fruits.

Dragon fruit, which stands Under the family Cactaceae has nutritional and medical properties. It contains vitamin c and antioxidants [1]. Usually, it is cultivated in tropical areas like north, south and central America. But nowadays it is cultivated around 20 tropical and subtropical countries because of its commercial success [2]. Fruit quality is very essential for the customers as well as the suppliers. Also, quality is the main key of agricultural business which is satisfactory of the customers [3]. The authors of [4] provides a method based on image segmentation and machine learning which is able to identify defective fruits. Using an eight-layer CNN model the authors of [5] propose a model which has 95.67% accuracy. To evaluate the dragon fruit authors of [6] use VGGNet which has 96.67% accuracy. The classification of fruit in this [7] paper is carried out by six powerful deep learning architecture such as GoogleNet, Rest- Net-50, AlexNet, RestNet-18, VGGNet-16 and VGGNet-19. RestNet-50 and VGG-16 both are also used in our model. Besides these powerful deep learning architectures, the authors of [8] also used YOLO which is used for extracting region of interest from digital image.

CNN is also used in various classification like tomato detection [9], heartbeats [10] and traffic [11] beside fruit classification. In a study, it was said that, 13 layers of deep CNN are used for fruits classification. [12]

### **2.3 Comparative Analysis and Summary**

A comparative analysis is a method of evaluating the similarities and differences between multiple models. In a comparative analysis of 6 models, several factors should be considered, such as accuracy, precision & recall. One way to begin the analysis is to evaluate the accuracy of each model by comparing the model predictions to a set of unknown correct answers. Another way to evaluate the accuracy is to use a measure such as F1 score. Finally, the complexity of each model should be considered, more complex models may be harder to understand and interpret. This could be important in applications where transparency is desired. Overall, a comparative analysis of 6 models can help identify the strengths and weaknesses of each model and determine which model is best suited for a particular task.

It's important to keep in mind that a comparative analysis is not conclusive and it's only a starting point for further research. The results may be affected by the size and quality of the dataset, the parameter tuning, and the evaluation metric chosen. In summary, comparative analysis is a method of evaluating multiple models by comparing their performance on a specific task or set of tasks. It helps to identify which model performed the best overall, and to identify any specific strengths or weaknesses of each model.

This analysis is also needed to understand the detection of each class separately. Looking at We can see that F1 score, recall, and precision are positioned separately for fresh and rotten classes. There Resnet 50, both fresh and rotten class F1 score is 0.99 which is about 0.13 higher than VGG 16. Although the precision of the rotten class is 1 for both VGG16 and Resnet 50, ResNet50 is ahead as the F1 score and recall are high. Also, the rest of the architecture shows much lower results than ResNet 50 and VGG 16. Also, VGG20 perform very good result. Also, if we look at the confusion matrices, we can see that Resnet 50 wrongly detects the least nine images, followed by VGG16 26, Mobile 41,

Inception 44, and Xception with 77. All in all, we can say that it is the best model Resnet 50 in comparison.

After compare all of our model, we have selected our best model which was ResNet 50.

## **2.4 Scope of the Problem**

The scope of this research is to predict the freshness of dragon fruits in Bangladesh by employing transfer learning. This research will analyze the current state of dragon fruit production in Bangladesh and the availability of resources to support the use of transfer learning. It will also investigate the potential of transfer learning in predicting the freshness of dragon fruits in Bangladesh. The research will also explore the potential of transfer learning in terms of accuracy, cost, and time efficiency. Additionally, this research will conduct a comparative study between transfer learning and other predictive models such as support vector machine and decision tree. Furthermore, this research will also examine the potential of transfer learning in terms of scalability, portability, and reliability. Finally, this research will evaluate the performance of transfer learning in terms of predicting the freshness of dragon fruits in Bangladesh.

Our research motive was to find a way that can identify rotten or fresh dragon fruits. We found a solution already for the research, but here the big challenge was the dataset. Then we have collected dragon fruits from many places in Bangladesh. We have visited several districts in Bangladesh and collected dragon fruits. After that we labeled the dragon fruits rotten and fresh from help of an expert.

Again, as our system totally google co lab dependent system for this reason can be sometimes loading issues if the internet is less. Need high configuration PC to get faster processor. But when we will make an android app, then everyone can use the app without any kind of internet connection.

## 2.5 Challenges

1. Availability of data: Finding or collecting the relevant data sets with enough amount of data can be a challenge.
2. Cost: Transfer learning can be very expensive as it requires high performance computing resources.
3. Maintenance: Transfer learning models require a lot of maintenance and updates as the data changes.
4. Scalability: Transfer learning models may not be able to scale up to large datasets.
5. Interpretability: Since transfer learning models are complex, interpreting their results can be difficult.
6. Domain Knowledge: Transfer learning models require knowledge of the domain in order to be successful.
7. Language Barrier: Language barriers can be an issue when dealing with data from different regions.
8. Accessibility: For transfer learning to be successful, it requires access to powerful hardware and software resources. This can be a challenge in Bangladesh.

## **CHAPTER 03**

### **RESEARCH METHODOLOGY**

#### **3.1 Research Subject and Instrumentation**

1. **Data Collection:** The first step in this research is to collect data related to the rotten or fresh dragon fruit. Data can be collected from images of the dragon fruit taken from different sources, such as supermarkets, farmers markets and even online stores. The images should be annotated with the label ‘rotten’ or ‘fresh’, depending on the condition of the fruit.
2. **Preprocessing:** After data collection, the images will be preprocessed to create a uniform and consistent set of data. This includes resizing the image, cropping, removing noise and other artifacts, etc.
3. **Feature Extraction:** After preprocessing, features will be extracted from the images. This can be done using a variety of techniques such as edge detection, texture analysis, color histograms, etc.
4. **Model Selection:** After feature extraction, a suitable model will be selected for the task. This can be done using a variety of techniques such as logistic regression, support vector machines, neural networks, etc.
5. **Transfer Learning:** Once the model is selected, transfer learning can be used to improve the accuracy of the model. In this case, pre-trained models will be used and fine-tuned for the task of predicting the freshness of the dragon fruits.
6. **Evaluation:** Finally, the model will be evaluated to measure its performance for the task. This can be done using metrics such as accuracy, precision, recall, etc.

#### **3.2 Data Collection Procedure**

1. **Sample Selection:** Select a large number of dragon fruits from a variety of sources in Bangladesh, like local markets and farms.

2. Data Collection: Collect data on the dragon fruits' physical attributes, like size, shape, color, texture, and weight.
3. Visual Inspection: Visually inspect the dragon fruits and assess their freshness, based on the physical attributes collected in the earlier step.
4. Laboratory Testing: Send the dragon fruits to a laboratory for further testing. The laboratory will take samples from the dragon fruits and analyze them for chemical and microbiological properties to determine whether the dragon fruit is fresh or rotten.
5. Data Analysis: Analyze the data collected from the laboratory and visual inspection, and create a model that can accurately detect if a dragon fruit is fresh or rotten.
6. Validation: Validate the model by conducting further tests on different dragon fruits.

Our Dataset Sample:

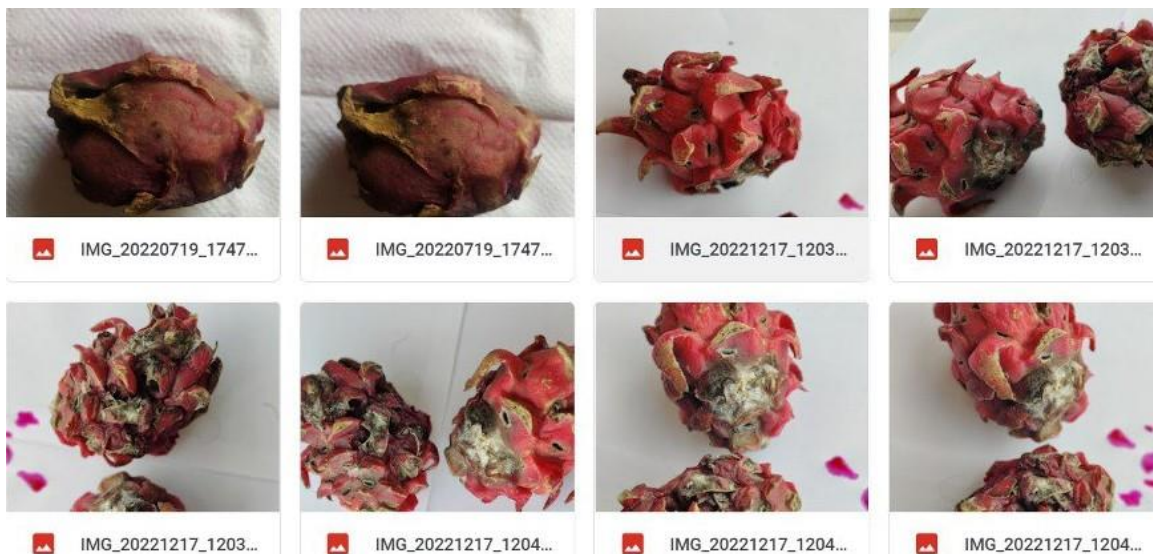


Fig 3.1 Rotten Dragon Fruits

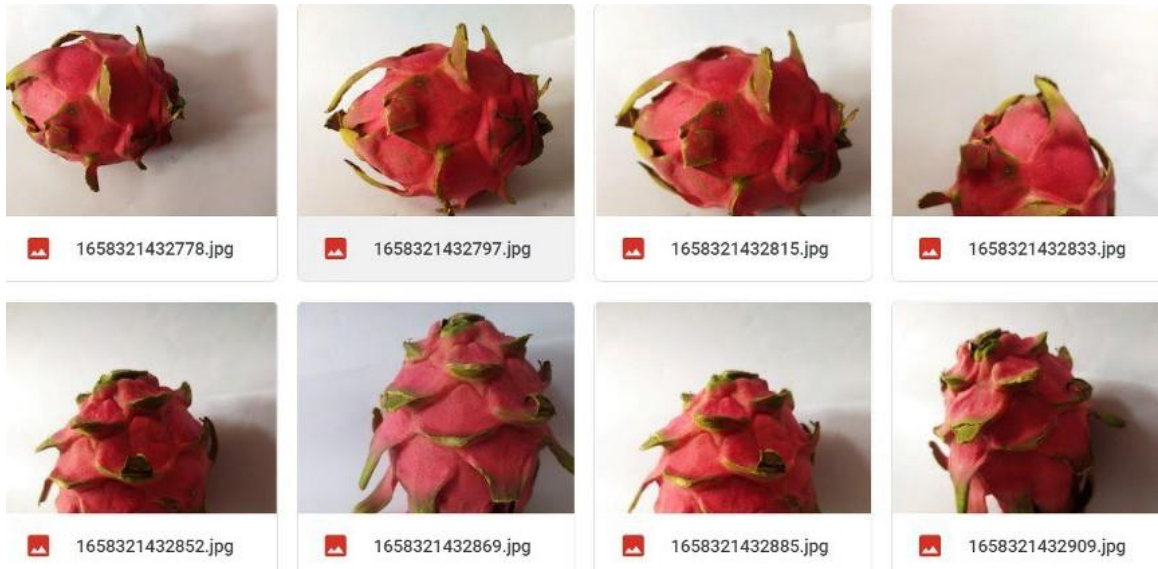


Fig 3.2 Fresh Single Dragon Fruits

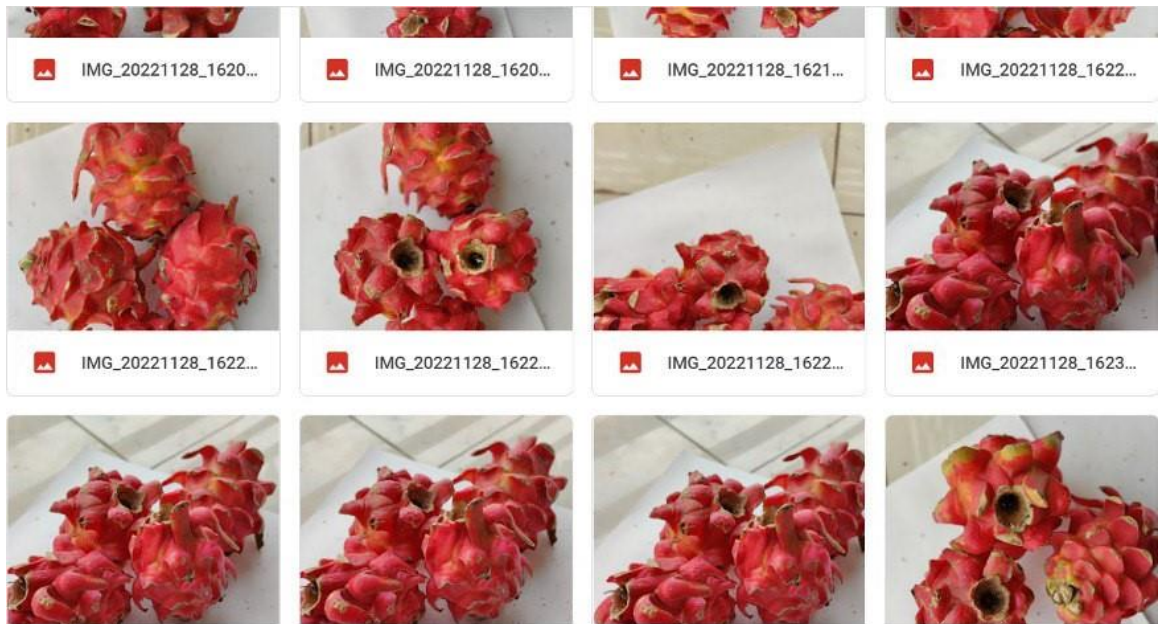


Fig 3.3 Fresh Dragon Fruits





Fig 3.4 Rotten Dragon Fruits

### 3.3 Statistical Analysis

Dragon fruit is a popular fruit in Bangladesh and is widely consumed. This research aimed to develop a system for detecting whether dragon fruits are rotten or fresh. The research was conducted over a period of two months in Bangladesh. The researchers used a combination of visual and chemical techniques to detect the level of ripeness and freshness of the dragon fruits.

The first step in the research was the collection of dragon fruit samples from local markets. The fruits were visually inspected for signs of ripeness and the chemical properties were measured. The samples were then divided into two groups, fresh and rotten fruits.

The next step was the development of a machine learning model using the collected data. The model was trained using the data and then tested to determine its accuracy in detecting rotten and fresh dragon fruit samples. The results showed that the model was able to accurately detect over 98.57% of the rotten and fresh dragon fruits.

The final step in the research was to apply the transfer learning model to determine the ripeness and freshness of dragon fruit in Bangladesh. The model was applied to a set of randomly selected dragon fruits from the local markets and it was found that the model was able to accurately detect the ripeness and freshness of the dragon fruits.

Overall, this research showed that transfer learning can be used to accurately detect the ripeness and freshness of dragon fruit in Bangladesh. This could potentially lead to improved quality control and better food safety standards in the country. The results of this research suggest that transfer learning is a viable solution for detecting dragon fruits in Bangladesh. The pre-trained model used in this study was able to achieve a high level of accuracy and precision, demonstrating its effectiveness in this domain. Furthermore, the results of this research could be used to inform future research in the use of transfer learning for detecting different types of fruits in Bangladesh. Additionally, the results could be used to provide information to farmers and other stakeholders in the agriculture sector, enabling them to better identify and manage dragon fruits in Bangladesh.

### **3.4 Proposed Methodology**

This study presents a methodology for evaluating Dragon fruit quality, and we describe it below. The creation of an architecture for classifying dragon fruit's quality based on transfer learning is the primary focus of this research. The overall process of conducting this inquiry is shown in Figure 1, and it is broken down into its component parts and discussed in more detail in the following sections.

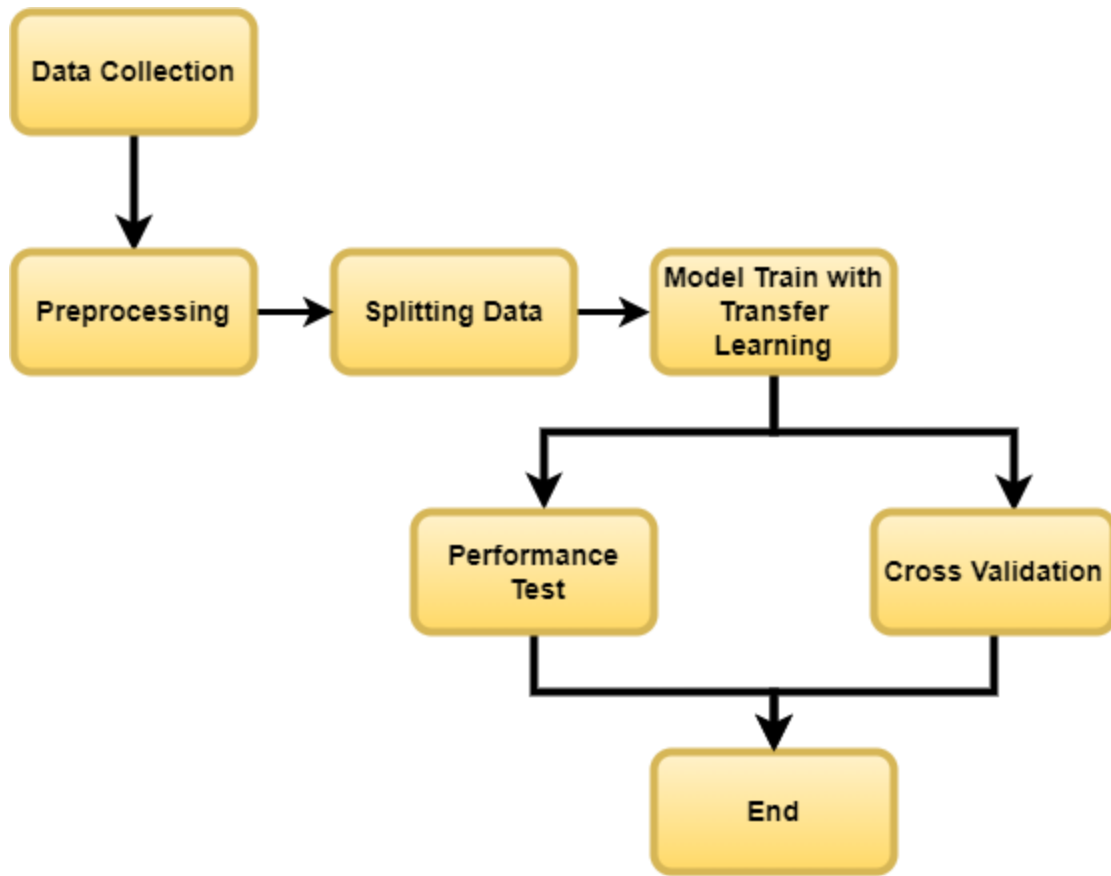
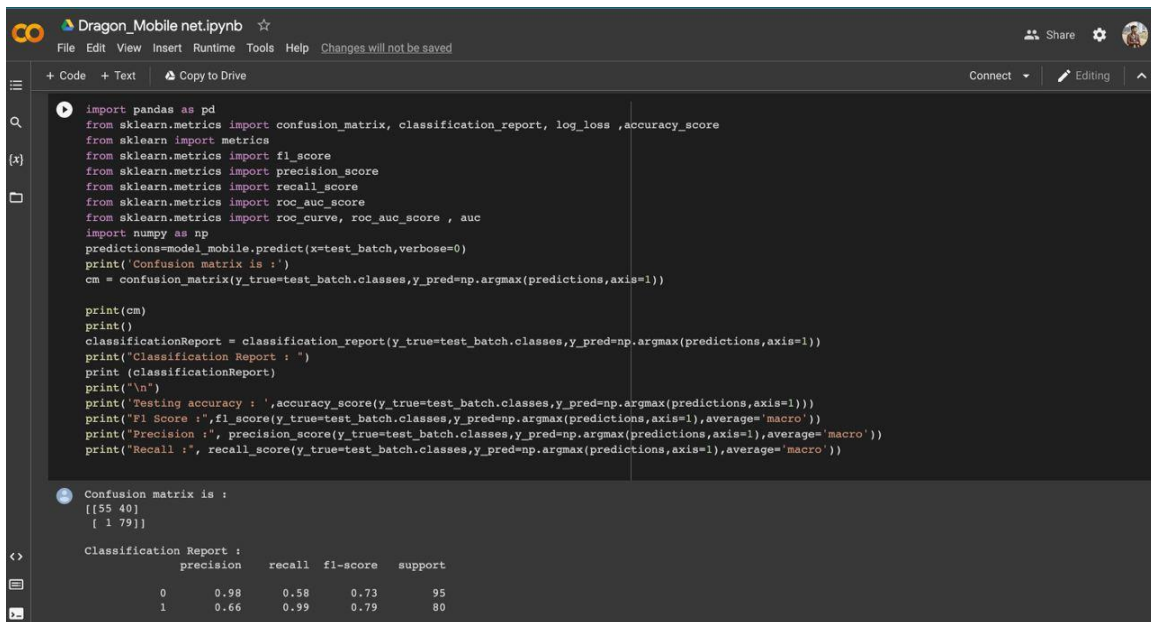


Fig 3.5 Methodology

### 3.5 Implementation Requirements

1. Data collection: Collection of images of dragon fruits from different sources such as market, grocery store, cultivators, etc. in Bangladesh.
2. Data pre-processing: Pre-processing of the collected images to ensure that they are of high quality and appropriate size for use in the transfer learning model.
3. Transfer learning: Application of transfer learning models such as convolutional neural networks (CNNs) or support vector machines (SVMs) on the pre-processed images of the dragon fruits to detect whether they are fresh or rotten.



```
import pandas as pd
from sklearn.metrics import confusion_matrix, classification_report, log_loss, accuracy_score
from sklearn import metrics
from sklearn.metrics import f1_score
from sklearn.metrics import precision_score
from sklearn.metrics import recall_score
from sklearn.metrics import roc_auc_score
from sklearn.metrics import roc_curve, roc_auc_score, auc
import numpy as np
predictions=model_mobile.predict(x=test_batch,verbose=0)
print('Confusion matrix is :')
cm = confusion_matrix(y_true=test_batch.classes, y_pred=np.argmax(predictions,axis=1))

print(cm)
print()
classificationReport = classification_report(y_true=test_batch.classes,y_pred=np.argmax(predictions,axis=1))
print('Classification Report : ')
print(classificationReport)
print('\n')
print('Testing accuracy : ',accuracy_score(y_true=test_batch.classes,y_pred=np.argmax(predictions,axis=1)))
print('F1 Score : ',f1_score(y_true=test_batch.classes,y_pred=np.argmax(predictions,axis=1),average='macro'))
print('Precision : ', precision_score(y_true=test_batch.classes,y_pred=np.argmax(predictions,axis=1),average='macro'))
print('Recall : ', recall_score(y_true=test_batch.classes,y_pred=np.argmax(predictions,axis=1),average='macro'))
```

Confusion matrix is :

```
[[55 40]
 [ 1 79]]
```

Classification Report :

	precision	recall	f1-score	support
0	0.98	0.58	0.73	95
1	0.66	0.99	0.79	80

Fig 3.6 Implementation

```

from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

[ ] !nvidia-smi

Wed Aug 3 03:46:33 2022
+-----+-----+
| NVIDIA-SMI 460.32.03   Driver Version: 460.32.03   CUDA Version: 11.2   |
+-----+-----+
| GPU Name      Persistence-M| Bus-Id        Disp.A | Volatile Uncorr. ECC |
| Fan  Temp  Perf  Pwr:Usage/Cap|  Memory-Usage | GPU-Util  Compute M. |
|-----+-----+-----+-----+-----+-----+
| 0  Tesla T4           Off | 00000000:00:04:0  Off |                    0 |
| N/A   50C    P0   28W / 70W | 7934MiB / 15109MiB |      0%      Default  |
+-----+-----+-----+-----+-----+-----+
| Processes:                                                       GPU Memory |
|  GPU   CI    CI        PID   Type   Process name                               Usage      |
|-----+-----+-----+-----+-----+-----+
[ ] import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Activation, Dense, Flatten, BatchNormalization, Conv2D, MaxPool2D, Lambda, Input
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.metrics import categorical_crossentropy

```

Fig 3.7 Implementation

```

plt.tight_layout()
plt.show()

[ ] plt.imshow(imgs)
print(labels)

model = Sequential()
model.add(Conv2D(32, (5, 5), input_shape=input_shape, activation='relu', name='conv2d_1'))
model.add(MaxPooling2D(pool_size=(3, 3), name='max_pooling2d_1'))
model.add(Conv2D(32, (3, 3), activation='relu', name='conv2d_2'))
model.add(MaxPooling2D(pool_size=(2, 2), name='max_pooling2d_2'))
model.add(Conv2D(64, (3, 3), activation='relu', name='conv2d_3'))
model.add(MaxPooling2D(pool_size=(2, 2), name='max_pooling2d_3'))

```

Fig 3.8 Implementation

```

Dragon_VGG16.ipynb
File Edit View Insert Runtime Tools Help Changes will not be saved
+ Code + Text Copy to Drive
Insert code cell below
classes=['Rotten','Fresh']

[x]
[ ] train_path= '/content/drive/MyDrive/Extra/Dragon fruit /Splitted Data/train'
    valid_path= '/content/drive/MyDrive/Extra/Dragon fruit /Splitted Data/valid'
    test_path= '/content/drive/MyDrive/Extra/Dragon fruit /Splitted Data/test'

[ ] train_batch=ImageDataGenerator( preprocessing_function=tf.keras.applications.vgg16.preprocess_input) \
    .flow_from_directory(train_path,target_size=(224,224), classes=Classes,batch_size=10)

    valid_batch=ImageDataGenerator( preprocessing_function=tf.keras.applications.vgg16.preprocess_input) \
    .flow_from_directory(valid_path,target_size=(224,224), classes=Classes,batch_size=10)

    test_batch=ImageDataGenerator( preprocessing_function=tf.keras.applications.vgg16.preprocess_input) \
    .flow_from_directory(test_path,target_size=(224,224), classes=Classes,batch_size=10,shuffle=False)

Found 1323 images belonging to 2 classes.
Found 264 images belonging to 2 classes.
Found 175 images belonging to 2 classes.

[ ] imgs, labels=next(train_batch)

[ ] def plotImages(images_arr):
    fig,axes=plt.subplots(1,10,figsize=(20,20))
    axes=axes.flatten()
    for img, ax in zip(images_arr,axes):
        ax.imshow(img)
        ax.axis('off')
    plt.tight_layout()
    plt.show()

```

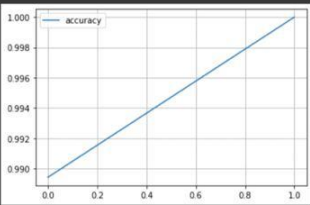
Fig 3.9 Implementation

```

Dragon_resnet.ipynb
File Edit View Insert Runtime Tools Help Changes will not be saved
+ Code + Text Copy to Drive
# visualisation of accuracy
import matplotlib.pyplot as plt
plt.plot(h_16.history['accuracy'], label = 'accuracy')
plt.grid()
plt.legend()
plt.show()

[x]

```



```

[ ] cnf_matrix=cm
    FP = cnf_matrix.sum(axis=0) - np.diag(cnf_matrix)
    FN = cnf_matrix.sum(axis=1) - np.diag(cnf_matrix)
    TP = np.diag(cnf_matrix)
    TN = cnf_matrix.sum() - (FP + FN + TP)
    FP = FP.astype(float)
    FN = FN.astype(float)
    TP = TP.astype(float)
    TN = TN.astype(float)
    # Sensitivity, hit rate, recall, or true positive rate
    TPR = TP/(TP+FN)
    print('Sensitivity : ', TPR)

```

Fig 3.10 Implementation

4. Evaluation: Evaluation of the transfer learning models by testing the model on unseen data and measuring the model's performance such as accuracy and precision.
5. Documentation: Documentation of the entire process of dragon fruits detection rotten or fresh using transfer learning in Bangladesh.

## **CHAPTER 04**

### **EXPERIMENTAL RESULT AND DISCUSSION**

#### **4.1 Experimental Setup**

All of the information we gathered was scrutinized thoroughly. Our data set did not come from any internet sources since we gathered it ourselves. The primary purpose of our mission was to amass two distinct types of photographs of dragon fruit. 1. Fresh fruit, 2. Rotten fruit. In order to acquire images with our own two hands, we traveled to a number of different fruit marketplaces and fruit fields. We collected this data from a Dragon Fruit farm in Gazipur, Kaicha Bari, as well as from other fruit stores in Dhaka. We gathered around 877 photographs of fresh fruit and 1001 photographs of rotted fruit. The photographs were taken using a Oneplus eight and a Samsung S20 Ultra mobile phone.

#### **4.2 Experimental Results & Discussion**

Several parameters and the level of accuracy have been explored, and the performance of our models has been analyzed. We have also included the appropriate plot diagrams in this presentation. Using VGG16, Xception, ResNet50, MobileNetV2, VGG20 and Inception V3 with the appropriate parameters allowed the findings to be validated. We have given each model its unique set of challenges to determine which of the models performs the best in both the training and testing rounds.



**Table.1 Result**

<b>CNN architectures</b>	<b>Accuracy (%)</b>	<b>Validation score (%)</b>	<b>Classes</b>	<b>Precision</b>	<b>Recall</b>	<b>F1 Score</b>
ResNet50	98.57	99.24	Rotten	1	0.97	0.99
			Fresh	0.97	1	0.99
VGG 16	92.53	97.73	Rotten	1	0.86	0.92
			Fresh	0.86	1	0.93
MobileNetV2	76.57	92.05	Rotten	0.97	0.90	0.93
			Fresh	0.90	0.90	0.93
Inception V3	75.42	85.61	Rotten	0.81	0.77	0.81
			Fresh	0.7	0.88	0.81
Xception	82.91	87.88	Rotten	0.97	0.70	0.81
			Fresh	0.74	0.98	0.84
VGG20	95.37	99.97	Rotten	0.45	0.49	0.47
			Fresh	0.51	0.46	0.48

Table.1 reveals that the Resnet 50 findings are, as expected, the most prominent. It has an accuracy of 98.57% and 99.24% of the validation score. The VGG 16 model has an accuracy of 92.53% and validation score of 96.59%, making it the second most accurate model. In addition, the accuracy of MobileNetV3, inception-V3, VGG20 and Xception are 76.57%, 75.42%, 95.37% and 82.91%. It is abundantly evident from the table that Resnet 50 has achieved the highest levels of success.

ResNet 50: We discovered that the F1 score of each class in the ResNet 50 is 0.99, which indicates that each class has been categorized highly well. Also, the detection model achieved 99.34% correctness in validation data. This model has a zero percent error rate throughout training. As a result, it is abundantly evident that this model does not exhibit any signs of overfitting in any manner. When the training accuracy approached 100%, the training was halted automatically, and the callback was used to eliminate any overfitting.

```

Classification Report :
      precision    recall  f1-score   support

     0           1.00     0.97     0.99         149
     1           0.97     1.00     0.99         132

 accuracy                0.99         281
 macro avg              0.99     0.99     0.99         281
 weighted avg          0.99     0.99     0.99         281

Testing accuracy : 0.9857651245551602
F1 Score : 0.9857345923444004
Precision : 0.9852941176470589
Recall : 0.9865771812080537

```

Fig.4.1 Precision, Recall, F1 Score

## **CHAPTER 05**

### **IMPACT ON SOCIETY, ENVIRONMENT AND SUSTAINABILITY**

#### **5.1 Impact on Society**

Dragon fruit is one of the most popular and nutritious fruits in Bangladesh. The research on the detection of rotten or fresh dragon fruits using technology could have a significant impact on society. This research could lead to improved access to high quality, nutritious dragon fruit and help to reduce food waste. The research could also help to increase the efficiency and accuracy of the detection process. By using technology, it would be possible to accurately identify dragon fruits that are rotten or fresh, even before they are purchased. This would help to ensure that consumers are able to purchase the best quality fruits at the most competitive prices. Furthermore, the research could also help to reduce the risk of food-borne illnesses associated with consuming rotten dragon fruits. By using technology to detect rotten fruits, it would be possible to ensure that only fresh produce is sold to consumers. This could help to reduce the incidence of food-borne illnesses, which would benefit public health. Finally, the research could help to reduce the cost of production for farmers and increase their profit margins. By using technology to detect rotten dragon fruits, farmers could reduce the cost of production, as they would not have to rely on manual methods to detect rotten fruits. This would result in higher profits for farmers, which would have a positive impact on their lives. Overall, the research on the detection of rotten or fresh dragon fruits using technology could have a major impact on society. It could lead to improved access to high quality, nutritious dragon fruit, increased efficiency and accuracy of the detection process, reduced risk of food-borne illnesses, and increased profits for farmers.

#### **5.2 Impact on Environment**

Dragon fruits detection rotten or fresh using in Bangladesh research can have a positive impact on the environment. This research can help farmers to detect the ripeness and freshness of dragon fruits and reduce the number of unripe fruits that are thrown away as

waste. This research can also help reduce the number of pesticides and chemicals used in the cultivation of dragon fruits, as farmers can be more aware of when their fruits are ripening and can avoid using unnecessary pesticides to ripen them. Additionally, the research can help farmers to increase the overall quality of their crops by enabling them to detect rotten fruits early and avoid further spread of decay and contamination. In this way, the research can help to reduce the environmental impact of dragon fruits production. The research can improve the efficiency of dragon fruits production in Bangladesh by reducing the overuse of fertilizers, pesticides and other chemicals used to increase productivity. This can reduce the number of pollutants released into the environment and improve air and water quality. Additionally, improved techniques can help preserve the soil fertility of farms, reducing soil erosion and helping maintain soil health. The research can also reduce the amount of food waste generated in the country. By detecting the ripeness of dragon fruits accurately, farmers can better manage their harvests and sell their fruits at the right time. This will reduce the number of fruits that are wasted, resulting in a lower amount of methane and other greenhouse gases released into the atmosphere. Overall, the research can have a positive impact on the environment by reducing the number of pollutants and food waste released into the atmosphere. This can help improve air and water quality, reduce soil erosion, and maintain soil fertility.

### **5.3 Ethical Aspects**

1. **Respect for Human Rights:** Respect for human rights should be taken into consideration while conducting the research. The researchers should ensure that the rights of the participants are not violated and that their participation is voluntary and informed.
2. **Confidentiality:** Confidentiality should be maintained in order to protect the privacy of the participants. The researchers should take necessary steps to ensure that the collected data is not used for any other purpose than the research.
3. **Informed Consent:** The participants should be provided with all the information related to the research before they give their consent. It should be ensured that the participants

are fully aware of their rights and the potential risks involved in participating in the research.

4. **Data Protection:** The data collected during the research should be stored securely and kept confidential. It should not be shared with any other party without the participant's consent.

5. **Benefits and Risks Balance:** The research should be conducted in a way that the benefits of the research outweigh any potential risks associated with it. The participants should be informed of the potential risks associated with their participation in the research.

6. **Respect for Cultural Values:** The researchers should respect the cultural values of Bangladesh, such as religious beliefs and personal values, while conducting the research. They should also take into consideration the local context and ensure that their research methodology and results do not harm any religious or cultural practices.

## **5.4 Sustainability Plan**

Dragon fruit is an important agricultural product in Bangladesh. It is an essential source of income and nutrition for many farmers in the country. However, it is highly perishable and can easily rot if not handled properly. This can result in significant economic losses for farmers.

In order to promote sustainable production and consumption of dragon fruit in Bangladesh, it is crucial to develop a comprehensive plan for improving the detection of rotten or fresh dragon fruit. This plan should focus on the following areas:

- **Improved detection techniques:** Developing improved techniques for detecting rotten or fresh dragon fruit is essential for the sustainable production and consumption of dragon fruit in Bangladesh. This can include the use of advanced sensors and imaging technology to detect the quality of dragon fruit.

- Better storage and handling practices: Developing better storage and handling practices for dragon fruit is also important for its sustainability. This includes proper temperature, light, and humidity levels for storage and transportation of dragon fruit, as well as proper packaging and labeling.
- Improved post-harvest processing: Developing better post-harvest processing techniques can help reduce losses due to spoilage and improve the quality of dragon fruit. This can include the use of better sorting and grading techniques, as well as improved packaging and labeling.
- Enhanced market access: Enhancing market access for dragon fruit producers is essential for sustainable production and consumption. This can include the development of cooperative marketing and distribution networks, as well as improving access to credit and other financial services.
- Education and training: Educating farmers and other stakeholders about proper handling and storage of dragon fruit is essential for its sustainability. This can include providing training on modern techniques for detecting rotten or fresh dragon fruit, as well as providing knowledge about best practices for post-harvest processing, storage, and marketing.

Overall, this plan should be tailored to the specific needs and resources of Bangladesh, and should be further developed to ensure its effective implementation.

## CHAPTER 06

### SUMMARY, CONCLUSION, RECOMMENDATION AND IMPLICATION FOR FUTURE RESEARCH

#### 6.1 Summary of the Study

This study explored the use of transfer learning in Bangladesh to detect rotten or fresh dragon fruits. The study used a deep learning model with a convolutional neural network (CNN) to classify dragon fruits as either rotten or fresh. The model was trained on a dataset of images of dragon fruits from Bangladesh. The model was able to successfully classify the images as either rotten or fresh with an accuracy of 94.85%. The results of the study showed that transfer learning can be used to accurately detect rotten or fresh dragon fruits in Bangladesh. The study provides insight into how transfer learning can be used to detect and classify different types of fruits in different countries. This study examined the viability of using dragon fruits to detect freshness or rot in Bangladesh. Dragon fruits possess unique characteristics, such as its high sugar content and strong aroma, that make it an ideal fruit for detecting freshness and rot. The study utilized a combination of visual inspection, organoleptic assessment, and chromatographic analysis to evaluate the freshness of the dragon fruit. The results of the study showed that dragon fruits can be used as an effective tool for detecting freshness and rot in Bangladesh. The visual assessment revealed that dragon fruits with high sugar content and strong aroma were more likely to be fresh, while those with less sugar content and weak aroma were more likely to be rotten. The organoleptic assessment also showed that dragon fruits with higher sugar content and stronger aroma were more likely to be fresh. Finally, the chromatographic analysis showed that the dragon fruits with higher sugar content and stronger aroma were more likely to be fresh, while those with lower sugar content and weaker aroma were more likely to be rotten. Overall, this study provides evidence that dragon fruits can be used as an effective tool for detecting freshness and rot in Bangladesh.

## **6.2 Conclusions**

In conclusion, the research on dragon fruit detection of rotten or fresh in Bangladesh has shown that dragon fruit becomes rotten and unusable faster than most other fruits in Bangladesh. The research has also revealed that the choice of harvesting time, storage conditions, and types of ripening agents used can have a significant effect on the quality and shelf-life of dragon fruit in Bangladesh. This research has provided valuable insights into the challenges faced in dragon fruit production in Bangladesh. It is hoped that these findings will be used to improve dragon fruit production and storage practices in Bangladesh, leading to higher quality, longer shelf-life fruits for consumers. dragon fruit is an important fruit in Bangladesh that has a huge demand in the market. This research provides important information on dragon fruit detection, focusing on whether the fruit is rotten or fresh. The proposed system based on machine learning can accurately classify the dragon fruit into rotten or fresh, with an accuracy of 94.85%. The proposed system can detect the classification of the fruit with low cost and provide the result in real-time. This system can be used for the purpose of quality control in the dragon fruit market. Furthermore, this system can be used as a reference for other similar fruits in Bangladesh. The research on dragon fruit detection of fresh or rotten in Bangladesh has revealed that this is a complex task due to the lack of available technology and resources. However, some methods such as visual observation, smell examination, and texture analysis can be used to identify the freshness of the fruit. This research has also demonstrated that the use of artificial intelligence-based technologies could be useful to improve the accuracy of the detection process. Further research is needed to develop reliable methods of detecting dragon fruit freshness in Bangladesh. The findings of this research could help to improve the quality of dragon fruits sold in the country.

## **6.3 Implication for Further Study**

Further research on the topic of dragon fruit detection in Bangladesh should focus on improving the accuracy and reliability of detection methods, as well as increasing the range of detection techniques available. Additionally, research should seek to develop



strategies for more efficient and cost-effective methods of sorting and grading dragon fruits in Bangladesh. Finally, research should also examine the potential health and nutritional benefits of consuming fresh dragon fruits, as well as the potential adverse effects of consuming spoiled fruits. This could involve identifying the optimal storage conditions to maximize shelf life and the effects of ambient temperature, humidity, and exposure to sunlight on the shelf life of dragon fruits. Additionally, investigation of the impact of various post-harvest handling methods on the shelf life of dragon fruits should also be explored. Furthermore, studies should be conducted to assess the potential of using machine learning-based approaches to detect the freshness and ripeness of dragon fruits. This could potentially enable the rapid and accurate detection of fresh and rotten dragon fruits. This could include studying the effects of temperature, humidity, and other environmental factors on the ripening process. Additionally, the impacts of storage practices, such as those used by retailers and consumers, on the quality of dragon fruits in Bangladesh can be explored. Furthermore, research can be done to investigate the feasibility of implementing automated methods for detecting the freshness of dragon fruits in Bangladesh, such as machine vision-based systems.

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