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# FLOWER RECOGNITION USING CNN



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# OUR GROUP MEMBERS



Subhasrere  
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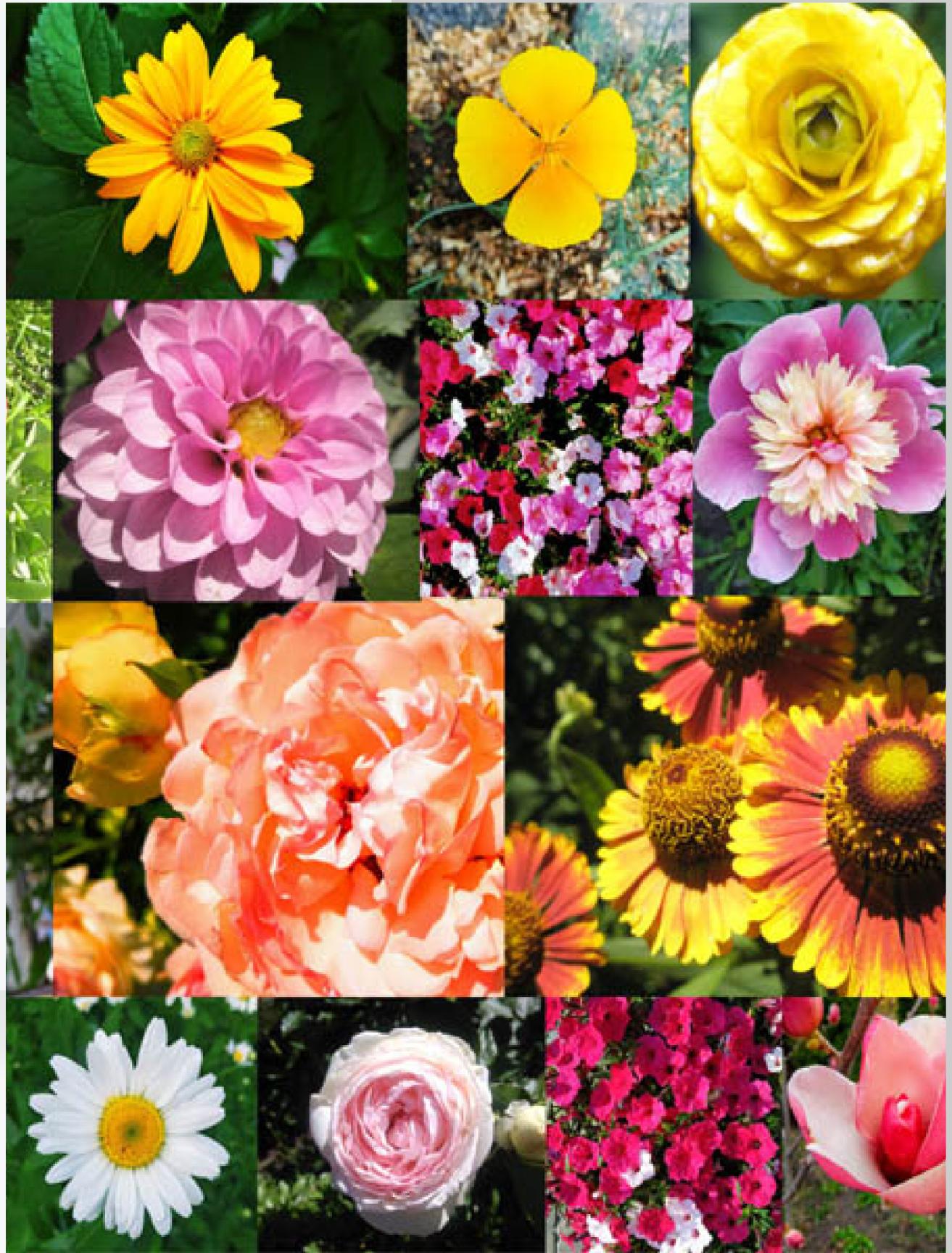
Sagnika  
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# PROBLEM STATEMENT

THE PRIMARY GOAL OF THIS PROJECT IS TO DEVELOP A CONVOLUTIONAL NEURAL NETWORK (CNN) MODEL CAPABLE OF AUTOMATING THE PROCESS OF FLOWER CLASSIFICATION. THIS INVOLVES TRAINING THE MODEL TO ACCURATELY CLASSIFY IMAGES OF FLOWERS INTO PREDEFINED CATEGORIES CORRESPONDING TO DIFFERENT SPECIES OR CLASSES.



# PROJECT INTRODUCTION

## Botany

### Species Identification



## Agriculture

To monitor crop health and growth stages by identifying flowering plants

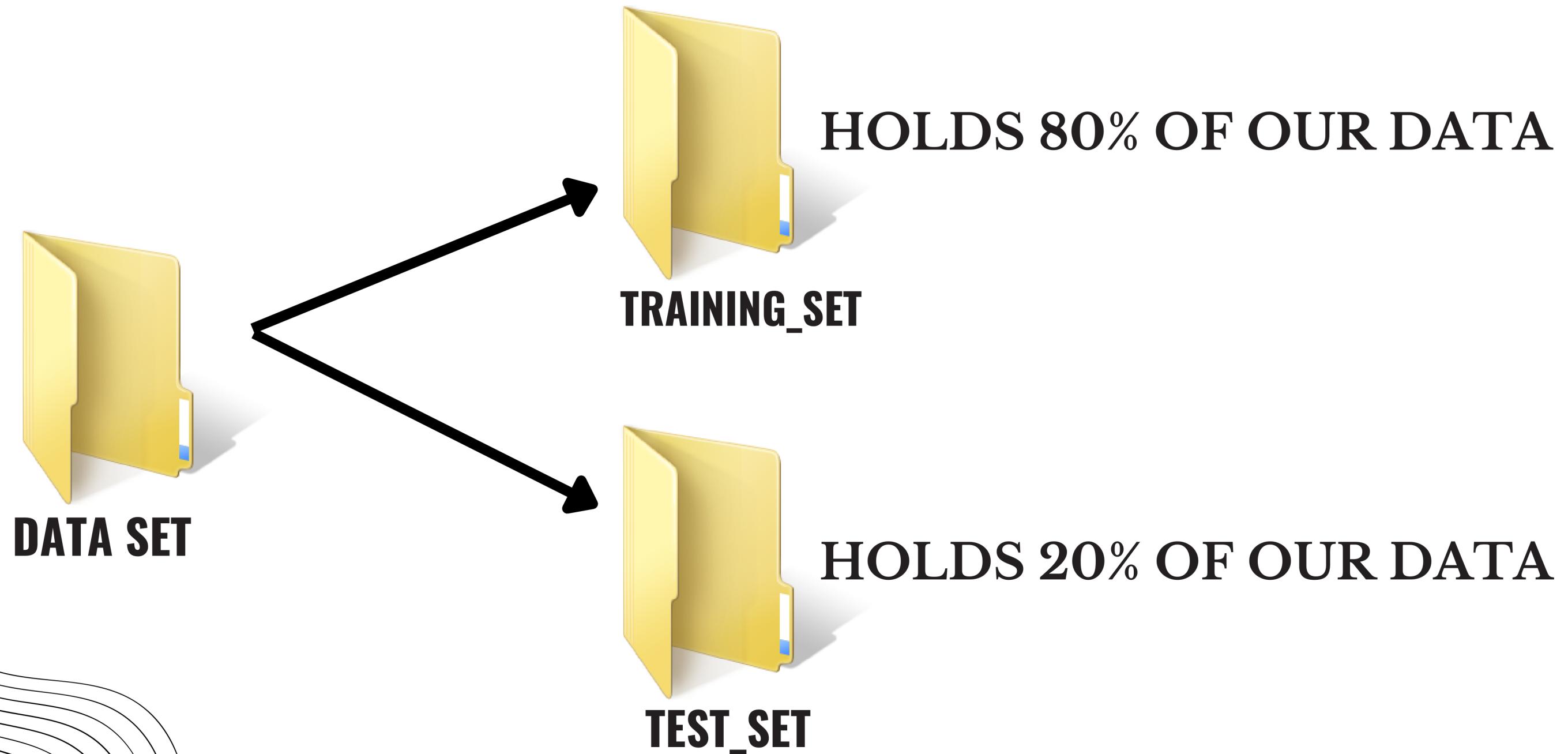


## Environmental monitoring

This project helps keep an eye on nature's well-being by automatically identifying flowers. It helps us understand if ecosystems are healthy, where pollinators live, and if any harmful plants are invading.



# DATA COLLECTION



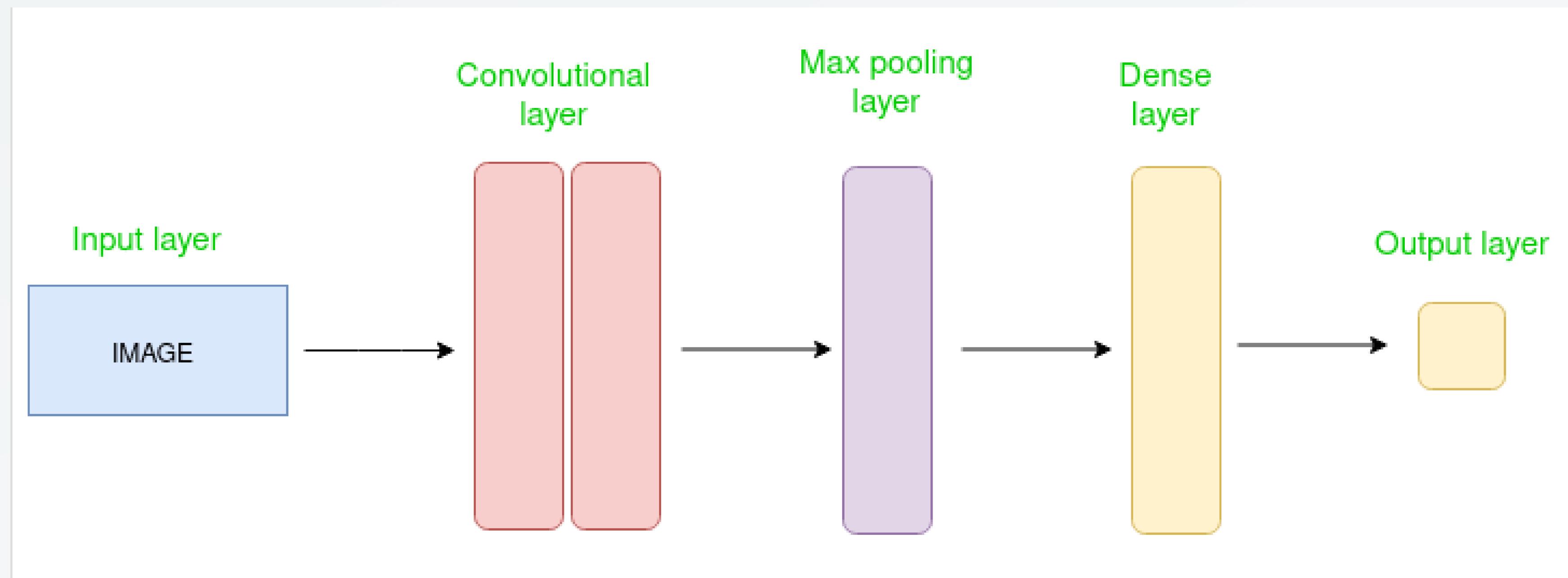
# DATA PREPROCESSING

```
# Data Augmentation and Preprocessing  
  
train_datagen = ImageDataGenerator(  
    rescale=1./255,  
    shear_range=0.2,  
    zoom_range=0.2,  
    horizontal_flip=True)  
training_set = train_datagen.flow_from_directory(  
    'training_set',  
    target_size=(64, 64),  
    batch_size=32,  
    class_mode='categorical')  
  
test_datagen = ImageDataGenerator(rescale=1./255)  
test_set = test_datagen.flow_from_directory(  
    'test_set',  
    target_size=(64, 64),  
    batch_size=32,  
    class_mode='categorical')
```

## TECHNIQUES INCLUDE:

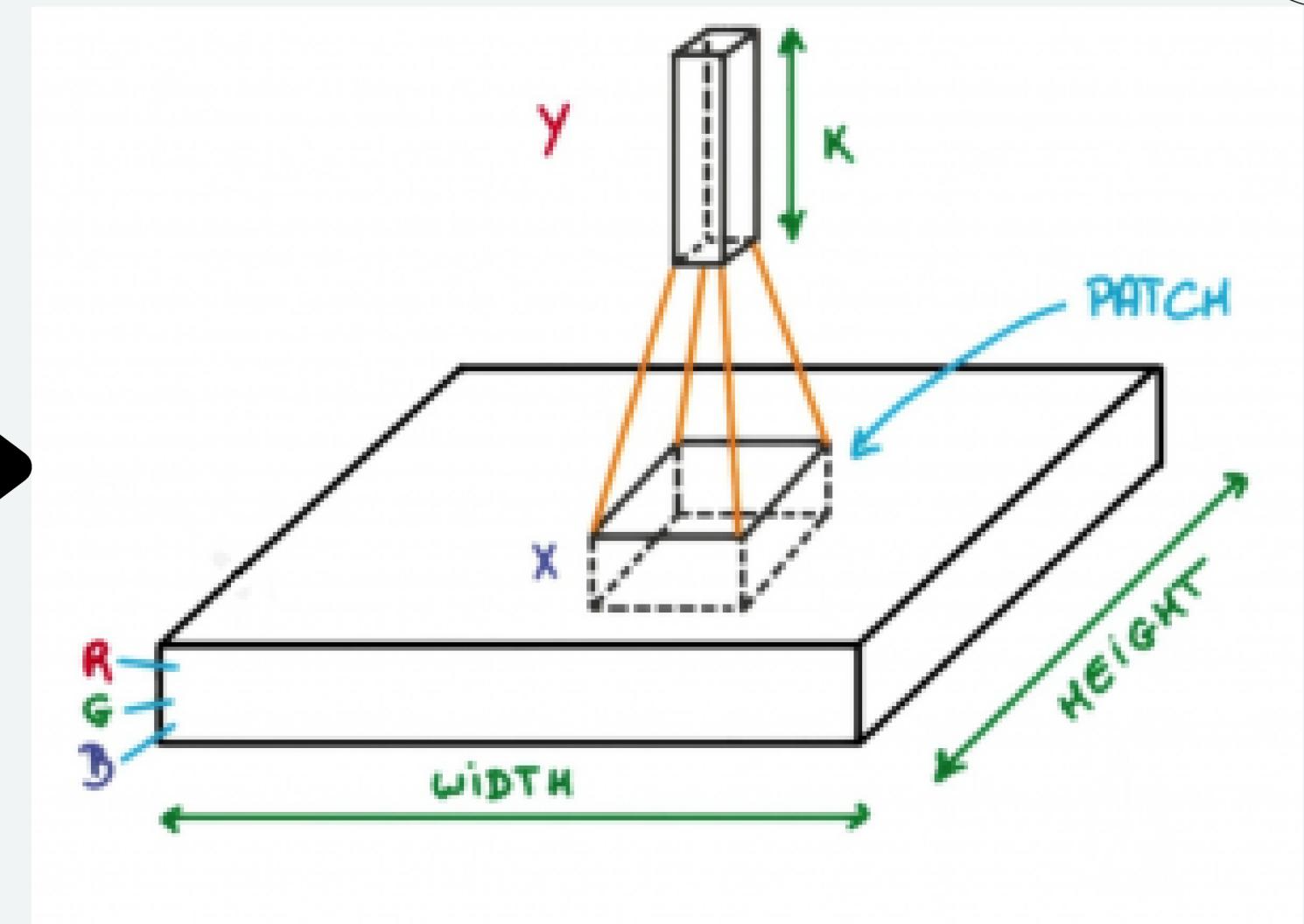
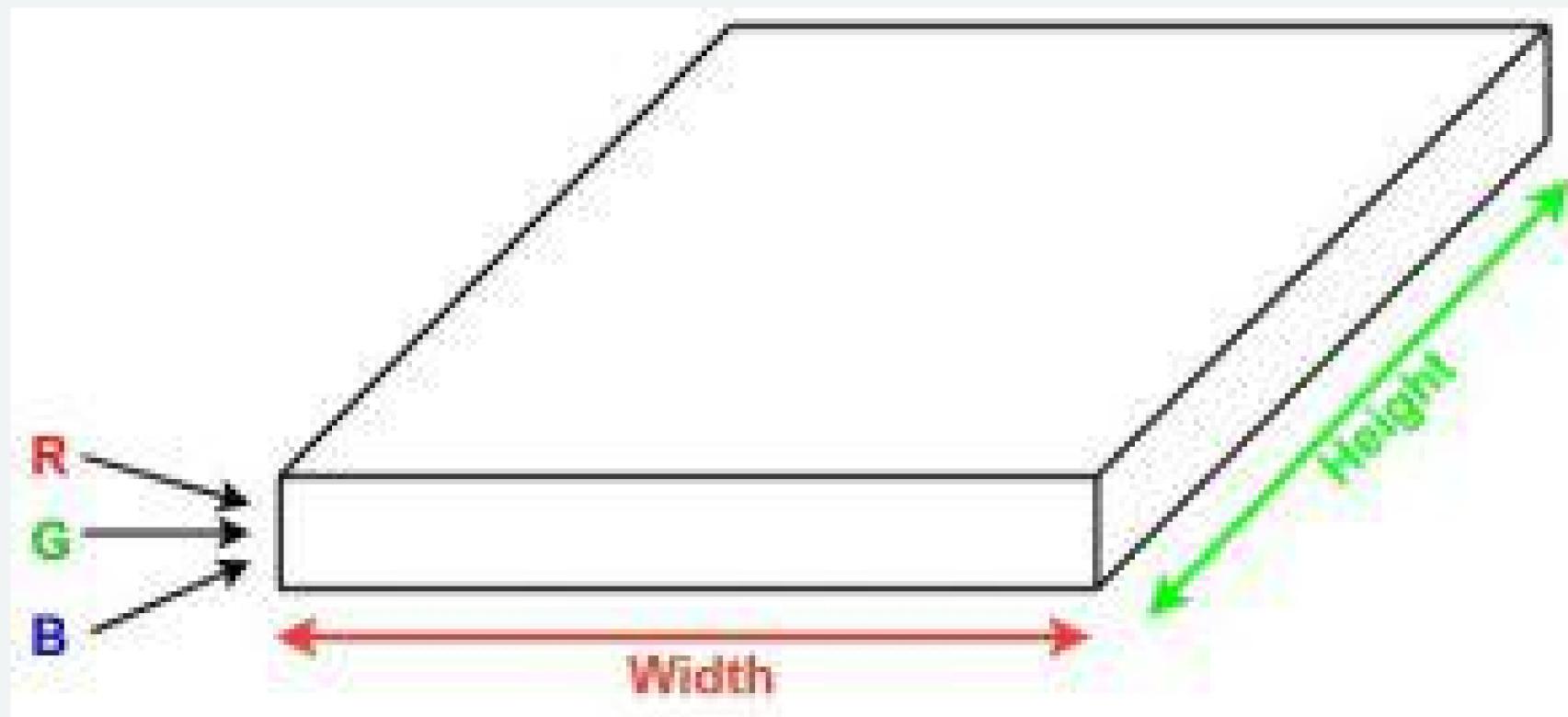
- Random shearing (shear\_range=0.2).
- Random zooming (zoom\_range=0.2).
- Horizontal flipping (horizontal\_flip=True).

# MODEL ARCHITECTURE

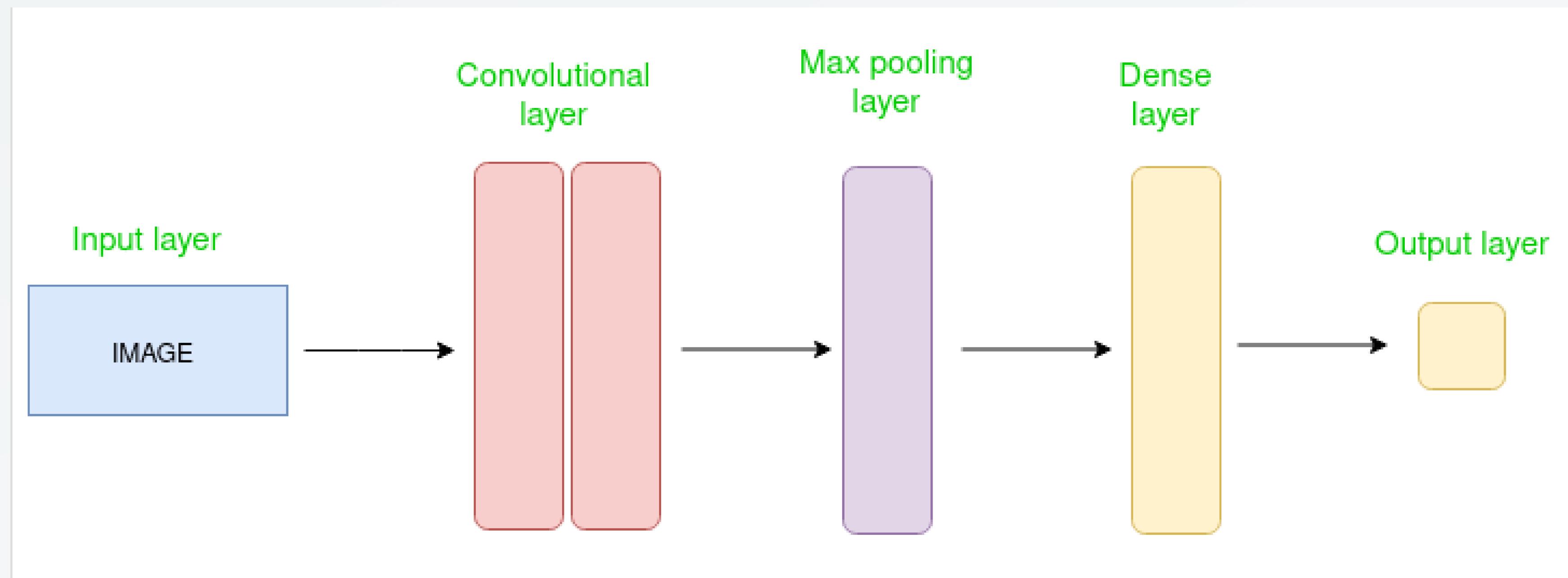


**Convolutional Neural Network (CNN)**

# CONVOLUTIONAL LAYERS



# MODEL ARCHITECTURE



**Convolutional Neural Network (CNN)**

# MODEL EVALUATION

python

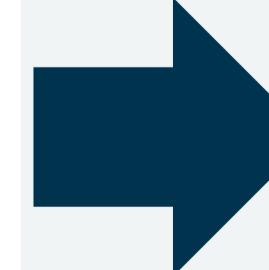
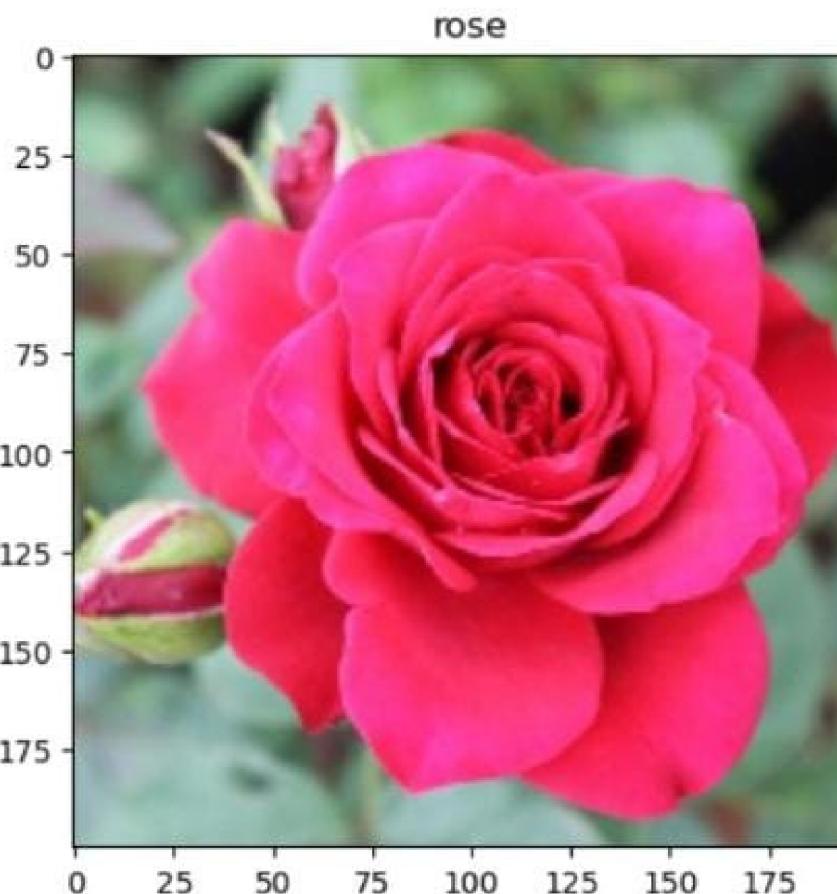
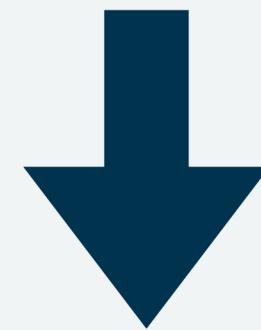
 Copy code

```
cnn.fit(x=training_set, validation_data=test_set, epochs=30)
```

- The model is trained using data from the **training\_set**, with parameters updated based on the optimization algorithm and loss function specified during compilation.
- The **validation\_data** argument allows the model to evaluate its performance on a separate set of data (test\_set) after each epoch. This evaluation is crucial for monitoring the model's generalization ability and detecting overfitting.

# SAMPLE PREDICTIONS

```
In [26]: import matplotlib.pyplot as plt  
predicted_class = categories[np.argmax(result)]  
test_image = tf.keras.utils.load_img('prediction/r.jpg',target_size=(200,200))  
plt.imshow(test_image)  
plt.title(predicted_class)  
plt.show()  
  
#test_image = tf.keras.utils.img_to_array(test_image)  
#test_image = np.expand_dims(test_image, axis=0)
```



```
In [22]: category=["Daisy", "Dandeline", "Rose", "Sunflower", "tulip"]  
if result[0][0]==1:  
    print('Daisy')  
elif result[0][1]==1:  
    print('Dandelion')  
elif result[0][2]==1:  
    print('Rose')  
elif result[0][3]==1:  
    print('SunFlower')  
elif result[0][4]==1:  
    print("Tulip")
```

Rose

# FUTURE SCOPE

01

02

03

04

## DATASET EXPANSION

Augment the existing dataset with additional images of various flower species to enhance model generalization and performance

## FINE-TUNING

Explore hyperparameter tuning and optimization techniques to further improve model accuracy and robustness.

## USER INTERFACE ENHANCEMENT:

Design an intuitive user interface for seamless interaction with the flower recognition system, facilitating user engagement and usability.

## USER INTERFACE ENHANCEMENT:

Develop methods for deploying the trained model in real-time applications, such as mobile or web-based flower recognition tools.

# CONCLUSION

*In conclusion, the flower recognition project has successfully developed and trained a convolutional neural network (CNN) model capable of accurately identifying different species of flowers from images. The model achieved 87% accuracy on the test dataset, demonstrating its effectiveness in flower classification tasks. Through comprehensive data preprocessing, model construction, and training, we have established a robust framework for automated flower recognition.*



THANK  
YOU

From team: **FLOWER ID**