

# Lung and Colon Cancer Detection

# Project Overview

- **Objective:**
  - Detect lung and colon cancer using advanced image classification techniques.
- **Dataset:**
  - Source: Kaggle
  - Total Images: 25,000 colored images
  - Classes:
    1. Colon adenocarcinoma
    2. Colon benign tissue
    3. Lung adenocarcinoma
    4. Lung benign tissue
    5. Lung squamous cell carcinoma

# Project Overview

- **Approach:**
  - Developed and evaluated four models using CNN and DenseNet architectures.
- **Team Composition:**

Ahmed Said Emam	CNN ‘Gray Images’
Mohammed Ahmed	CNN ‘Colored Images’
Ahmed Mohammed Elsayed	DenseNet ‘Gray Images’
Mohammed Ahmed Ismail	DenseNet ‘Colored Image’
Ahmed Maher	MLOps
Mohammed Abdalkader	Deployment ‘Flask’

# Introduction

- **Background:**
  - Cancer remain a leading cause of mortality globally, with lung and colon cancer being among the most common.
- **Lung Cancer Statistics:**
  - Leading cause of cancer deaths worldwide, approximately 1.8 million deaths in 2020.
  - 5-year survival rate is around .
- **Colon Cancer Statistics:**
  - Third most common cancer globally, with over 1.9 million new cases in 2020.
  - 5-year survival rate is approximately
- **Importance of the Project:**
  - Leveraging machine learning and deep learning for early detection.
  - Utilize image data from scans to develop robust models for classifying cancer types.
  - Enhancing diagnostic accuracy and reducing the burden on healthcare systems.

# Model Architectures

- **Models Developed:**
  - **Convolution Neural Network:**
    - Grayscale images.
    - Colored images.
  - **DenseNet pre-trained:**
    - Grayscale (Trained from scratch).
    - Colored (Fine-tuned pre-trained model).
- **Input Shape:**
  - 200x200 pixels for DenseNet models.
  - 200x200 pixels for CNN models

# Model 1: CNN on Colored Images

- **Architecture:**

- Convolution layers, pooling, ReLU activation, full connected layers, softmax for output.

Model: "sequential"

Layer (type)	Output Shape
conv2d (Conv2D)	(None, 200, 200, 32)
max_pooling2d (MaxPooling2D)	(None, 100, 100, 32)
conv2d_1 (Conv2D)	(None, 100, 100, 64)
max_pooling2d_1 (MaxPooling2D)	(None, 50, 50, 64)
conv2d_2 (Conv2D)	(None, 50, 50, 128)
max_pooling2d_2 (MaxPooling2D)	(None, 25, 25, 128)
conv2d_3 (Conv2D)	(None, 25, 25, 256)
max_pooling2d_3 (MaxPooling2D)	(None, 12, 12, 256)
conv2d_4 (Conv2D)	(None, 12, 12, 512)
max_pooling2d_4 (MaxPooling2D)	(None, 6, 6, 512)
flatten (Flatten)	(None, 18432)
dense (Dense)	(None, 256)
dense_1 (Dense)	(None, 5)

Total params: 6,288,709 (23.99 MB)

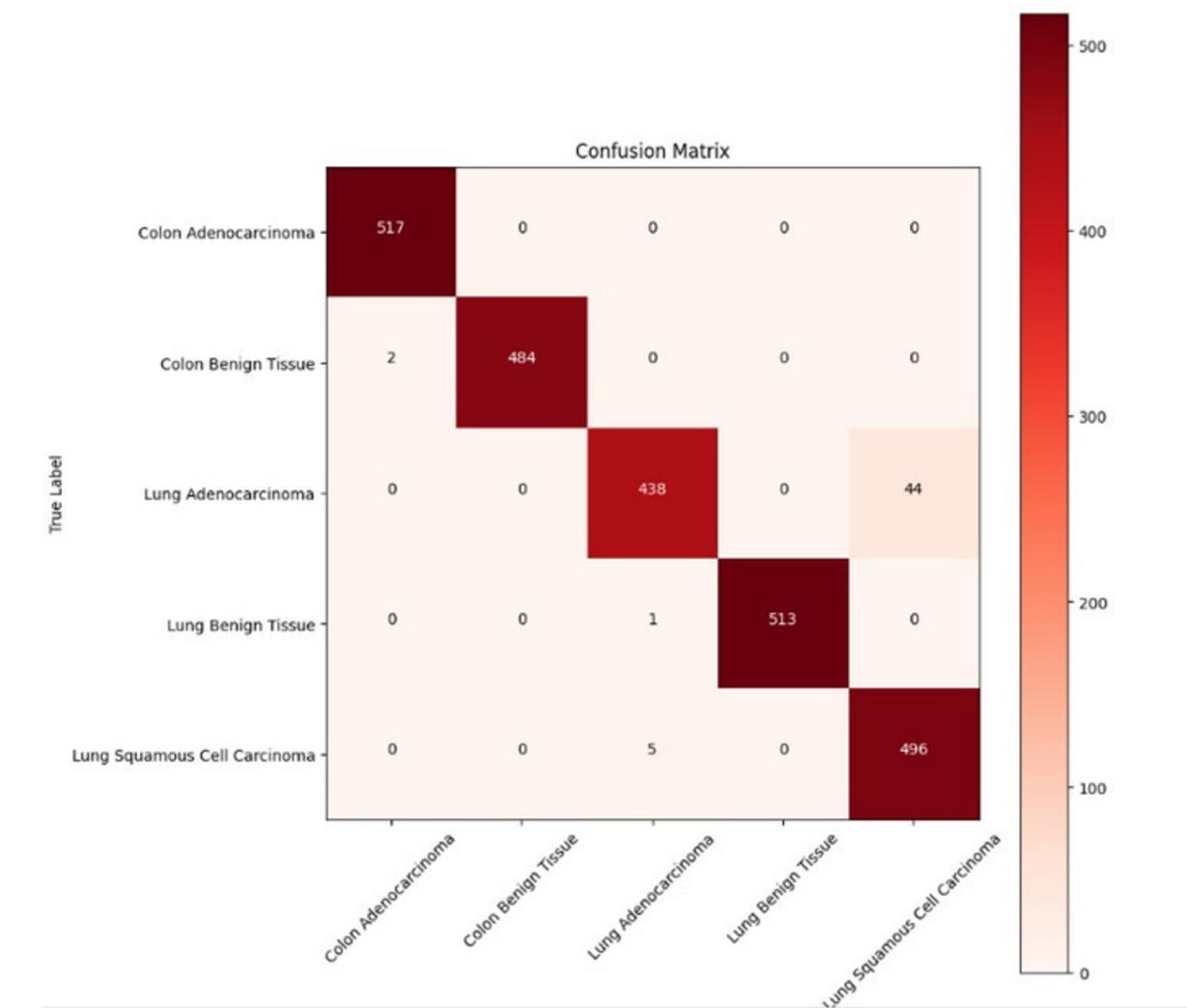
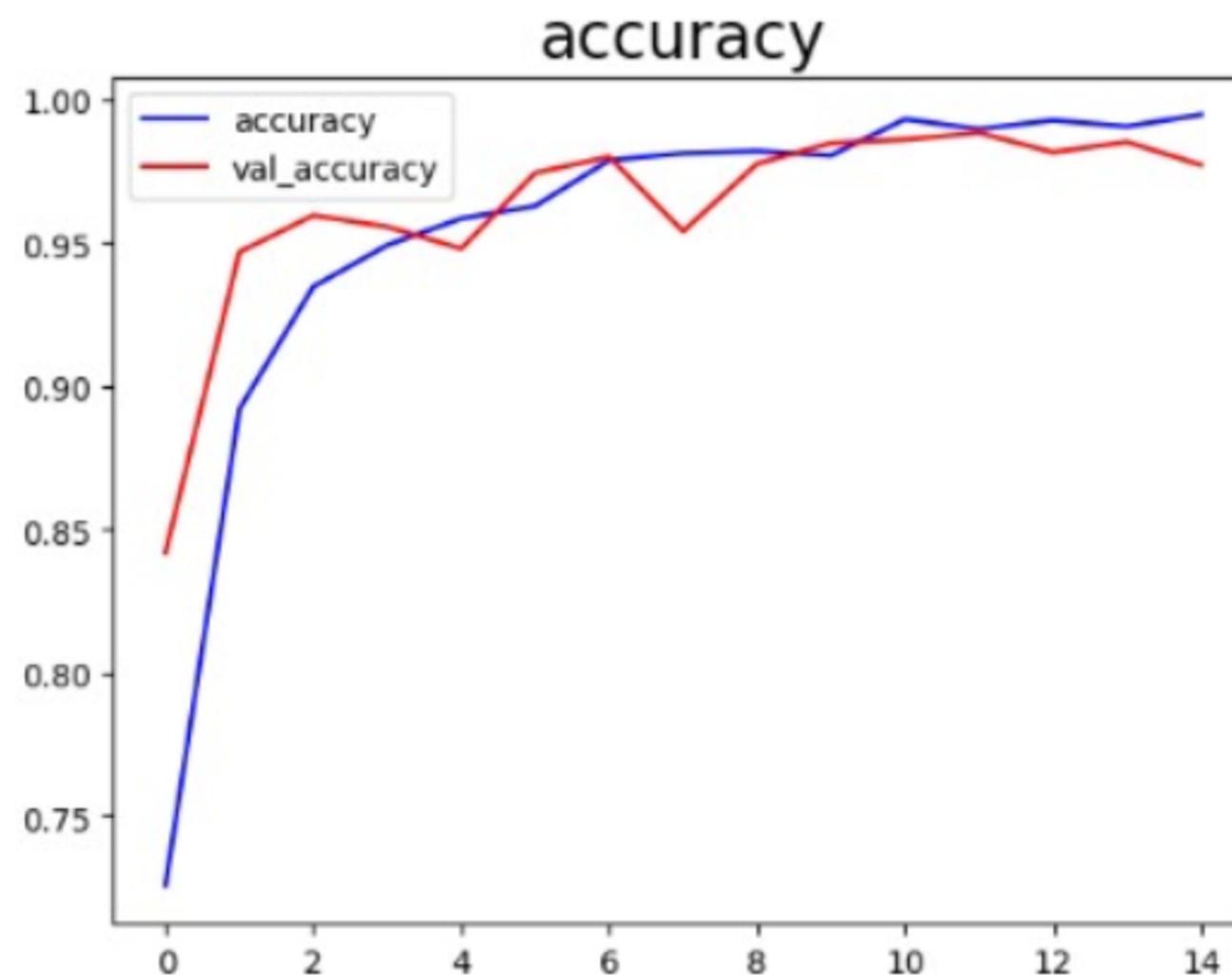
Trainable params: 6,288,709 (23.99 MB)

Non-trainable params: 0 (0.00 B)

# Model 1: CNN on Colored Images

## ■ Performance Metrics:

- Accuracy: 98% on test data.



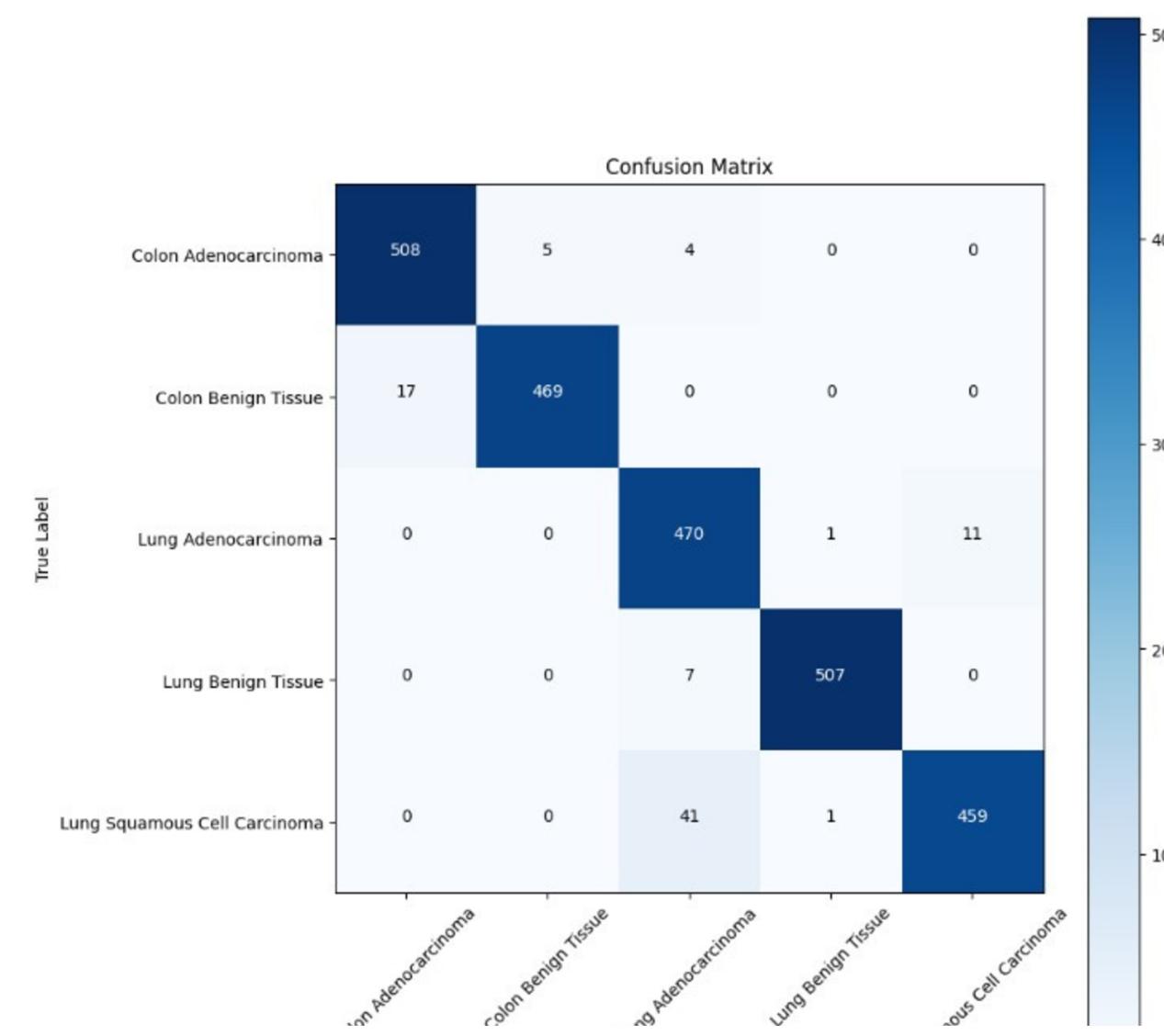
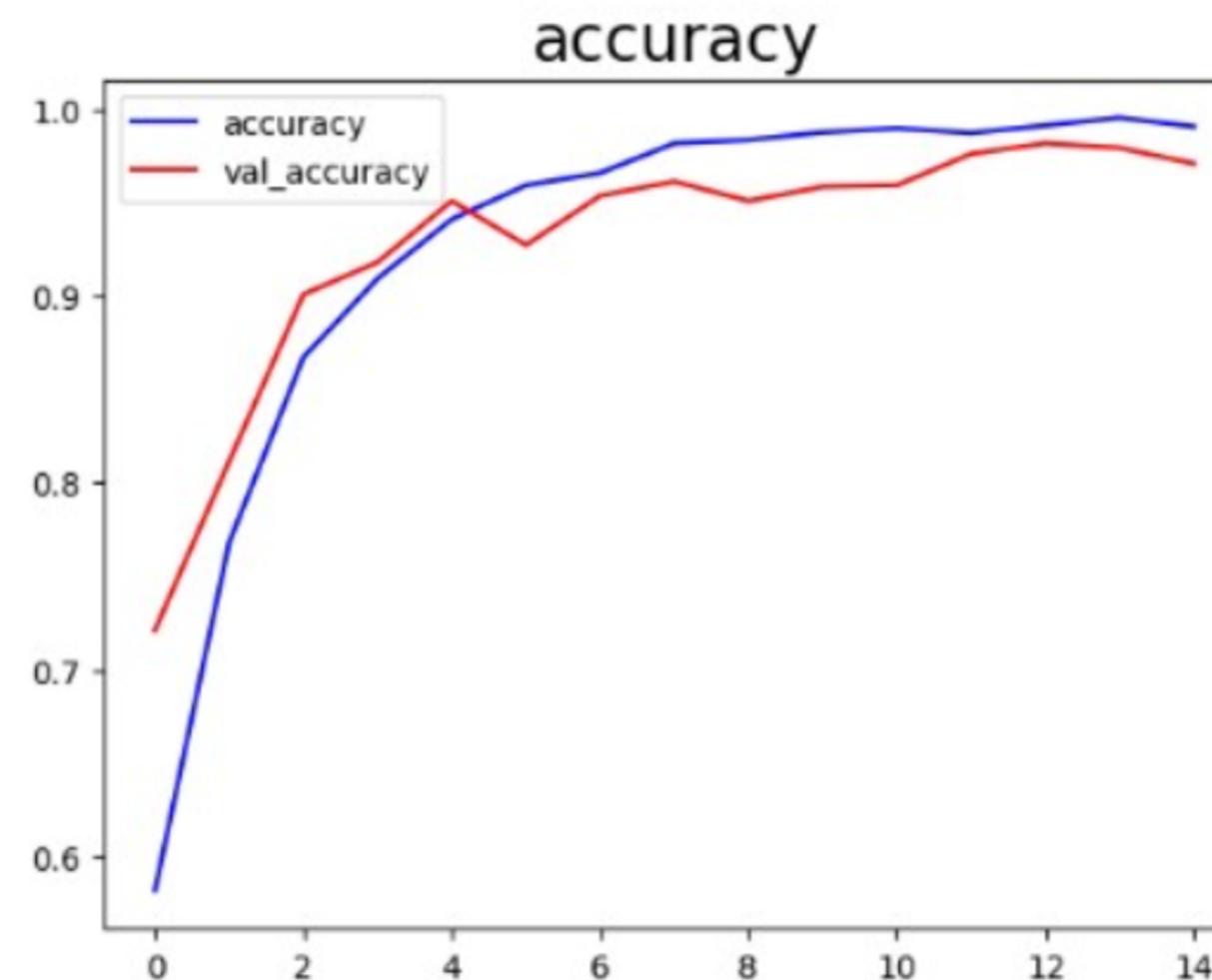
# Model 2: CNN on Grayscale Images

- **Architecture:**
  - Similar to Model 1 but adapted for grayscale images.
  - Grayscale input (derived from colored images).
- **Key Difference:**
  - Processing grayscale channels in input layers.

# Model 2: CNN on Grayscale Images

## ■ Performance Metrics:

- Accuracy: 96% on test data.



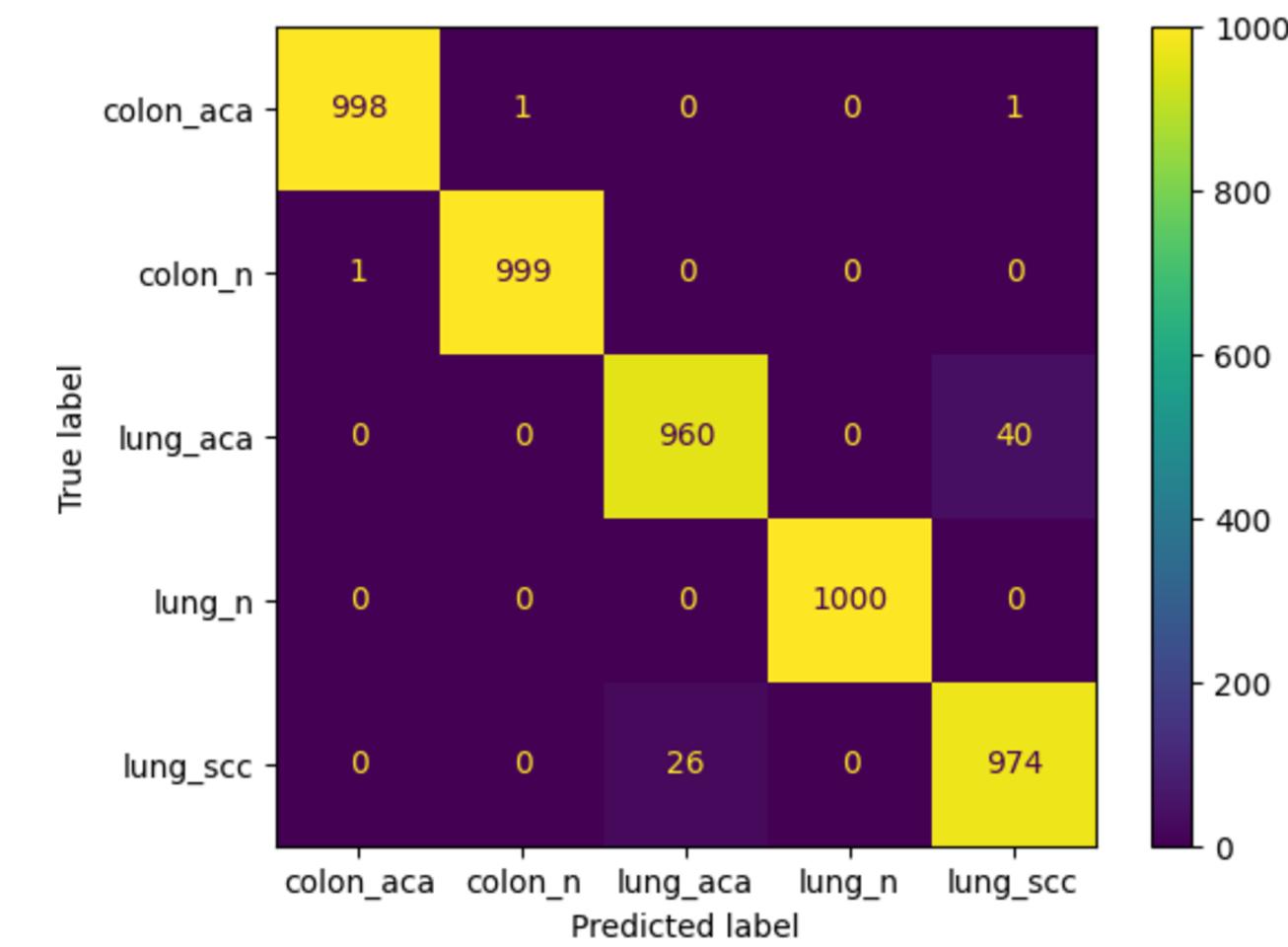
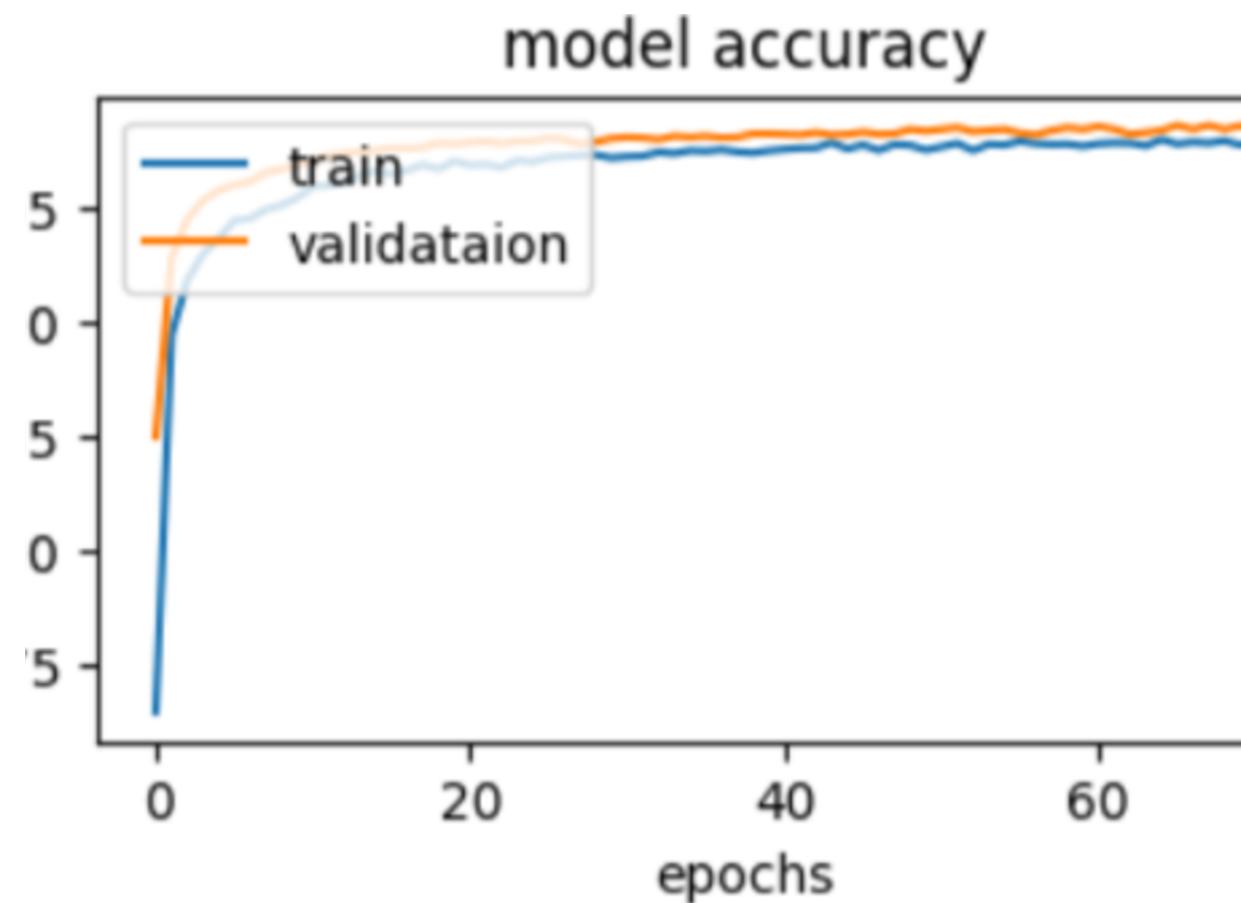
# DenseNet Model on Colored Images

- **Architecture:**
  - DenseNet pre-trained model (imageNet) fine-tuned for colored data.
- **Preprocessing:**
  - Images resized to pixels
  - Rescaling:  $1./255$
- **Optimizer:** Adamax (learning rate = 0.001)
- **Epochs:** 100 (Early Stopping applied)

# DenseNet Model on Colored Images

- **Performance Metrics:**

- Accuracy: 98% on test data.



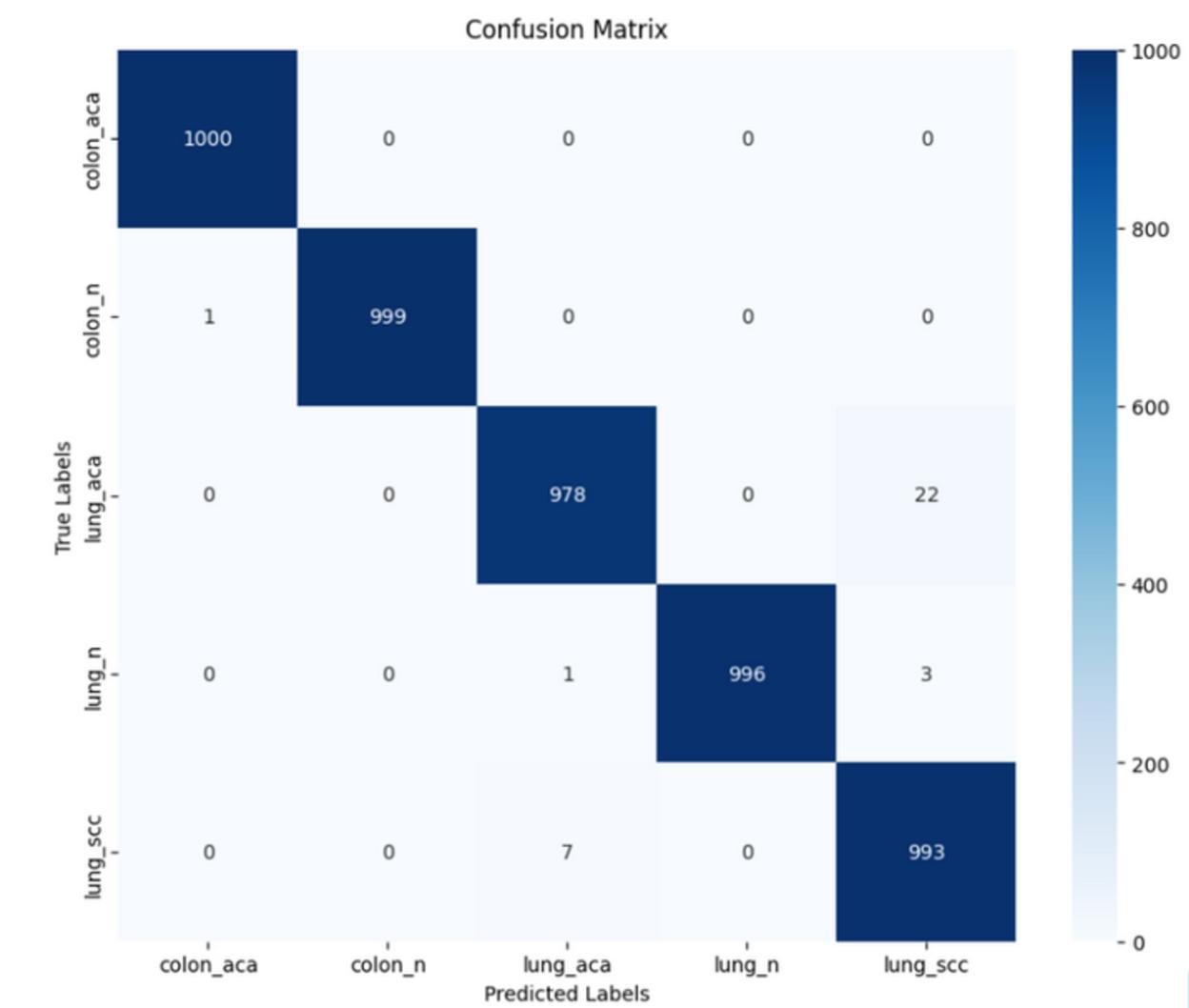
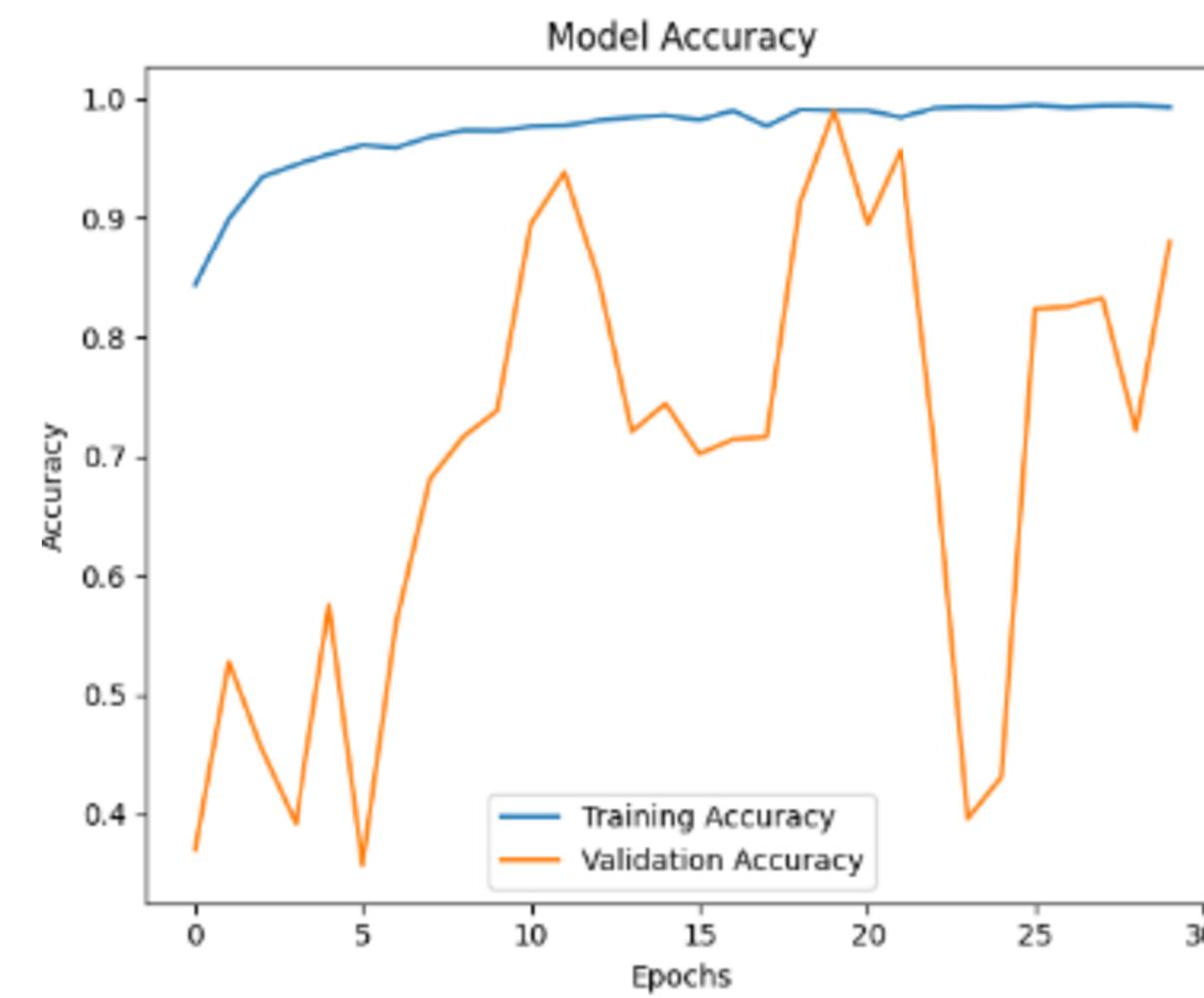
# DenseNet Model on Grayscale Images

- **Architecture:**
  - DenseNet pre-trained model trained from scratch on grayscale data.
- **Preprocessing:**
  - Image resized to pixels
  - Rescaling:  $1./255$
- **Optimizer:** Adamax (Learning rate = 0.001)
- **Epochs:** 100 (Early Stopping applied)

# DenseNet Model on Grayscale Images

## ■ Performance Metrics:

- Accuracy: 99% on test data.



# Model Comparison

- **Metric for Comparison:**
  - Accuracy, precision, recall

Metric	CNN (gray)	CNN (colored)	DenseNet (gray)	DenseNet (colored)
Accuracy	96%	98%	99%	98%
Precision	90%	97%	98%	96%
Recall	99%	98%	99%	97%

- **Best Model:** DenseNet on grayscale images.
  - Accuracy: 99% on test data

# Deployment

- **Tools Used:**
  - Flask for deployment
  - **MLOps Process:** Tested sample cancer images on each model to determine the best performance.
- **Result:**
  - DenseNet (grayscale) selected for deployment due to highest accuracy 99%

# Challenged & Solutions

- Challenges:
  - Computational resources: Managing model training time.
  - Preprocessing different image sizes: Handled with rescaling for each model.
- Solutions:
  - Utilized pre-trained model for efficiency.
  - Optimized models using Early Stopping.

# Future Work

- Expand dataset to include more cancer types.
- Investigate additional transfer learning models (e.g., EfficientNet).
- Implement cloud-based deployment for better scalability.

# Conclusion

- **Summary:** successfully built a model for lung and colon cancer detection. DenseNet (Grayscale) provided the best accuracy at .
- **Takeaway:** Pre-trained models (DenseNet) are highly effective for cancer detection, especially when paired with grayscale images.

*Thank You*