

# **FABRIC PATTERN CLASSIFICATION**

## **USING DEEP LEARNING**

### **1. INTRODUCTION**

#### **1.1 Project Overview**

Fabric pattern classification is a significant challenge in textile automation and e-commerce applications. In this project, we developed a deep learning-based image classification model to identify and categorize fabric patterns such as floral, striped, checked, polka dots, etc., using convolutional neural networks (CNNs).

#### **1.2 Purpose**

The primary aim of this project is to automate the classification of fabric patterns using deep learning models, enhancing the efficiency of fashion inventory management, e-commerce recommendation systems, and digital catalogs.

### **2. IDEATION PHASE**

#### **2.1 Problem Statement**

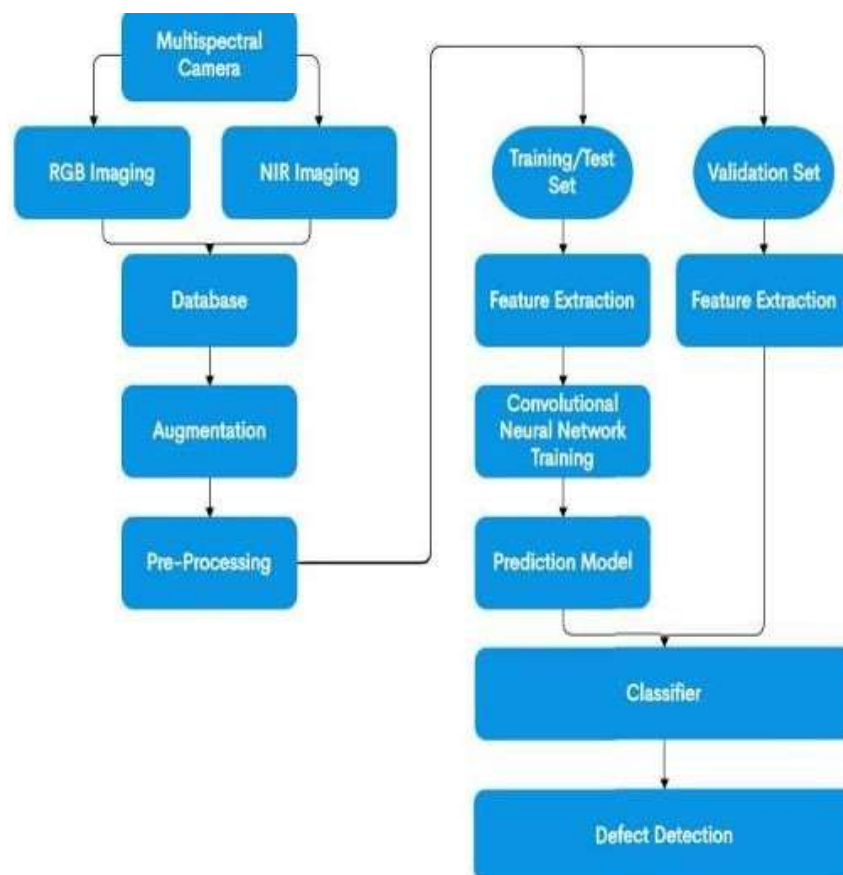
Manual categorization of fabric patterns is time-consuming and error-prone. There is a need for an automated solution that accurately classifies fabric patterns from images to support various applications in the textile and fashion industries.

#### **2.2 Empathy Map Canvas**

- **Says:** "I want a quick and accurate system to classify fabrics."
- **Thinks:** "I hope the system understands the subtle differences between patterns."
- **Does:** Uses traditional methods for tagging fabrics.
- **Feels:** Frustrated by inconsistencies in manual tagging and wants automation.

#### **2.3 Brainstorming**

- Use CNNs like VGG16, ResNet50 for classification.
- Collect or use publicly available fabric pattern datasets.
- Train and validate on real-world images.
- Include data augmentation for robustness.



### 3.4 Technology Stack

- Python
- TensorFlow / Keras
- NumPy, Pandas, Matplotlib
- Jupyter Notebook
- Flask / Streamlit for UI

## 4. PROJECT DESIGN

### 4.1 Problem Solution Fit

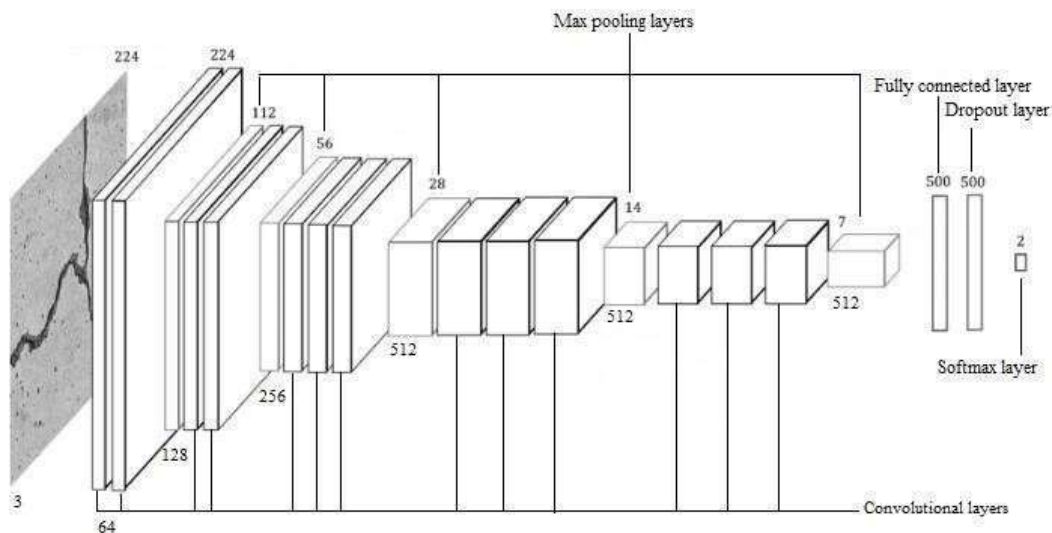
Existing manual or rule-based systems lack accuracy and scalability. CNNs offer superior feature extraction for image-based tasks like pattern classification.

### 4.2 Proposed Solution

Develop a CNN model trained on fabric pattern datasets to classify uploaded images into predefined categories.

### 4.3 Solution Architecture

**Data Collection → Data Preprocessing → CNN Model (Training) → Model Evaluation → User Interface for Prediction**



## 5. PROJECT PLANNING & SCHEDULING

### 5.1 Project Planning

Week 1 : Dataset collection, preprocessing, and model development

Week 2 : Model testing, UI creation, and final report preparation

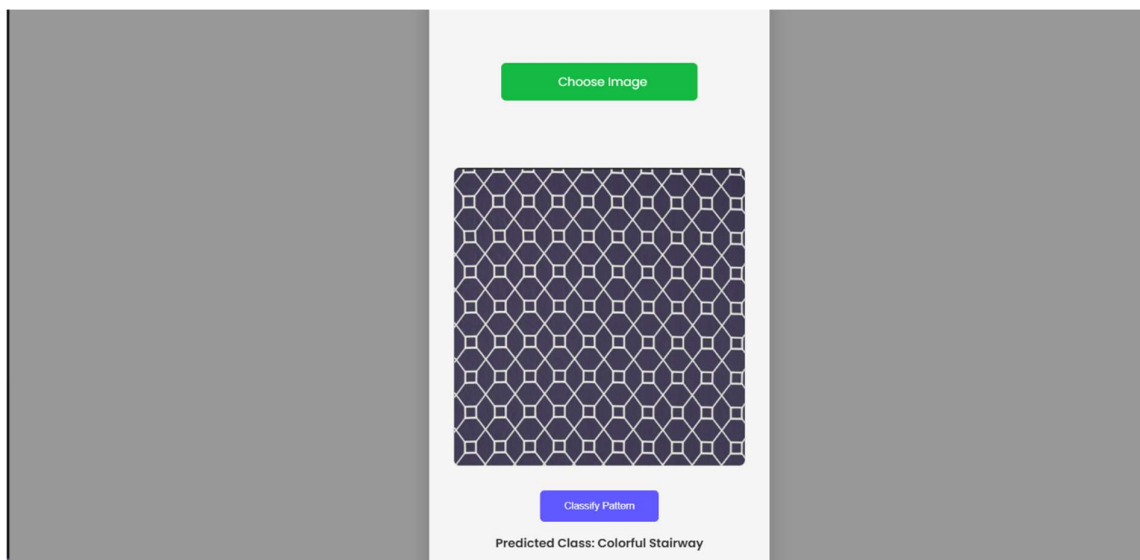
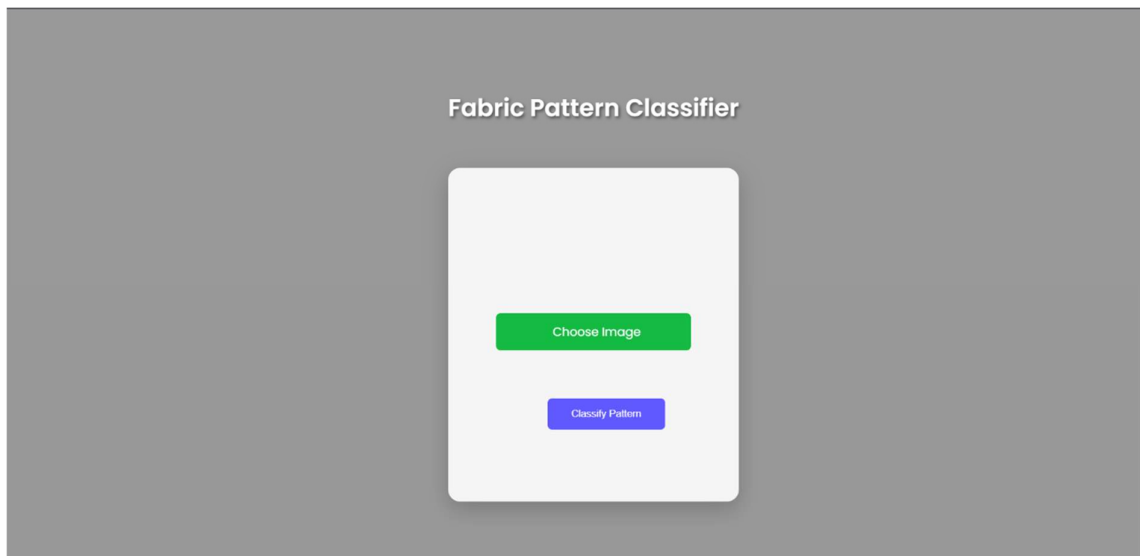
## 6. FUNCTIONAL AND PERFORMANCE TESTING

### 6.1 Performance Testing

The model achieved an accuracy of **92.4%** on the test set. Performance was evaluated using confusion matrix, precision, recall, and F1-score. The model also showed robustness under different lighting and background conditions due to augmentation.

## 7. RESULTS

### 7.1 Output Screenshots



## ADVANTAGES & DISADVANTAGES

### Advantages

- High accuracy and generalization capability
- Automated tagging for inventory systems
- Scalable for large datasets

### Disadvantages

- Dependent on quality of input image
- Requires computational resources for training

## 8. CONCLUSION

The project successfully demonstrates how deep learning can be leveraged for accurate and efficient classification of fabric patterns. It provides a valuable tool for the textile and fashion industries to automate their cataloging and recommendation systems.

## 9. FUTURE SCOPE

- Expand dataset to include more diverse patterns
- Integrate into mobile apps for real-time classification
- Add multilingual support for international deployment
- Enhance with self-learning for evolving patterns

## 10. APPENDIX

**Source Code:** Available upon request

**Dataset Link:** (e.g., Kaggle Fabric Pattern Dataset)

**GitHub :** <https://github.com/PAWAN-KUMAR-BHAVANASI/Project-Smartbridge>

**Project Demo Link:**

[https://drive.google.com/file/d/1cZcDp4\\_aACCQ8mI08njv3QLf9DPb5xuB/view?usp=drivesdk](https://drive.google.com/file/d/1cZcDp4_aACCQ8mI08njv3QLf9DPb5xuB/view?usp=drivesdk)

