# **Project Documentation**

# **Project Title**

Pattern Sense: Classifying Fabric Patterns Using Deep Learning

### **Team Members**

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# **Phase 1: Brainstorming & Ideation**

## **Objective:**

Generate a viable and innovative idea to classify fabric patterns using deep learning.

## **Key Points:**

- **Problem Statement:** Manual sorting of fabric patterns is inefficient, subjective, and unscalable.
- **Idea:** Build an AI-powered image classification system that can detect and label fabric patterns like floral, striped, geometric, Ethnic style etc.
- Inspiration: Challenges in textile/fashion industries, high demand for automation.
- **Technologies Considered:** CNN, TensorFlow, OpenCV, Flask/Streamlit, Transfer Learning (ResNet50/MobileNet).
- **Final Direction:** Train a CNN model and deploy it via a simple web interface for easy use.

# **Phase 2: Requirement Analysis**

## **Objective:**

Identify functional and technical requirements of the solution.

#### **Functional Requirements:**

- Upload an image of fabric through a web interface.
- Predict and display the pattern class (e.g., Ethnic style, geometric).
- Display a confidence score.

### **Non-Functional Requirements:**

- Real-time prediction (under 2 seconds).
- Accuracy above 85%.
- Responsive and secure deployment.

#### **Technology Stack:**

- Python, Flask, TensorFlow/Keras, OpenCV.
- **Dataset:** Kaggle clothing pattern dataset + custom augmentation.

#### **Data Flow:**

 $User \rightarrow Web \; UI \rightarrow Flask \; Backend \rightarrow CNN \; Model \rightarrow Predicted \; Output$ 

# Phase 3: Project Design

## **Objective:**

Plan the system architecture and UI components.

## **Key Points:**

- Architecture:

User uploads image via frontend.

Flask backend handles preprocessing and inference.

CNN predicts the pattern and returns output to user.

## - Model Design:

Transfer learning with MobileNet/ResNet50.

Image preprocessing: resize, normalize.

Metrics: Accuracy, Precision, Recall, F1-score.

- UI/UX Design:

Simple upload button.

Clear output with label and confidence score.

# **Phase 4: Project Planning (Agile)**

## **Objective:**

Divide project into iterative sprints for timely execution.

## **Sprints:**

- Sprint 1: Dataset cleaning and augmentation.
- Sprint 2: Model training and evaluation.
- Sprint 3: Web UI development (Flask).
- Sprint 4: Integration, testing, documentation.

Project Tools: GitHub, Jupyter Notebook, VS Code

# **Phase 5: Project Development**

## **Objective:**

Develop the model and integrate it with the interface.

## **Technology Stack:**

- Frontend: HTML/CSS (minimal Flask templating)
- Backend: Flask (Python), TensorFlow/Keras

- Model: Transfer Learning (ResNet/MobileNet), trained on augmented dataset
- **Deployment:** Local server / Streamlit / Docker (optional)

# **Phase 6: Functional & Performance Testing:**

## **Objective:**

Ensure system accuracy and speed meet expectations.

#### **Performance Results:**

- Accuracy: 87.2%

- Precision: 86.5%

- Recall: 88.0%

- F1-Score: 87.2%

- Inference Time: ~1.89 sec per image

- UI Responsiveness: Under 2 sec for full prediction cycle

#### **Test Types:**

- Unit Testing (model input/output)
- Integration Testing (UI + backend)
- User Testing (for usability)

# **Final Submission:**

### **Deliverables:**

- Trained Model (.h5/.pb)
- Flask Application Source Code
- Documentation (this file)
- Dataset Source Reference
- Demo video

# **Github Link:**

https://github.com/nsh536/Pattern-Sense-Classifying-Fabric-Pattern-using-Deep-Learning.git

## **Conclusion:**

This project demonstrates the power of deep learning in textile pattern recognition. It achieves reliable performance, reduces manual effort, and provides a scalable solution for fabric classification in fashion-tech environments.

# **Future Scope:**

- Mobile app integration
- Multi-label classification (e.g., texture + pattern)
- Real-time camera-based predictions
- REST API for third-party integration
- Edge deployment using TensorFlow Lite