

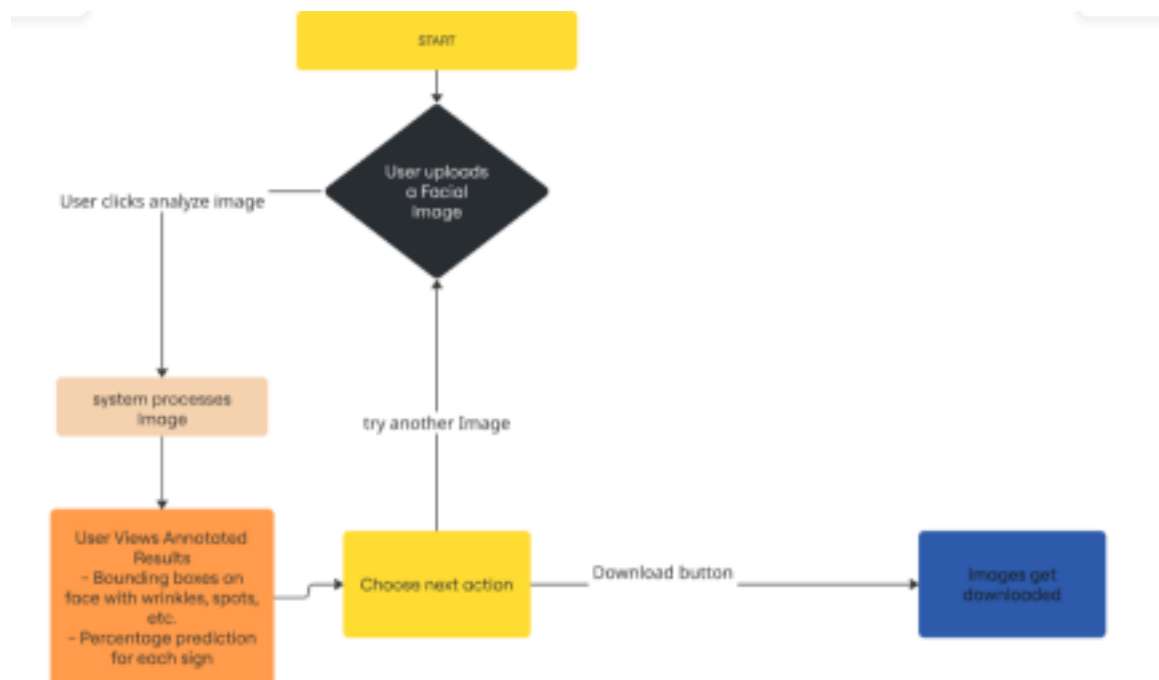
# DermalScan:AI\_Facial Skin Aging Detection App

## Project Statement:

The objective is to develop a deep learning-based system that can detect and classify facial aging signs—such as wrinkles, dark spots, puffy eyes, and clear skin—using a pretrained EfficientNetB0 model. The pipeline includes face detection using Haar Cascades, custom preprocessing and data augmentation, and classification with percentage predictions. A web-based frontend will enable users to upload images and visualize aging signs with annotated bounding boxes and labels.

## Outcomes:

- Detect and localize facial features indicating aging.
- Classify detected features into categories like wrinkles, dark spots, puffy eyes, and clear skin using a trained CNN model.
- Train and evaluate an EfficientNetB0 model for robust classification.
- Build a web-based frontend for uploading facial images and viewing annotated outputs.
- Integrate a backend pipeline that processes input images and returns annotated results.
- Export annotated outputs and logs for documentation or analysis.



## Modules to be implemented:

- Dataset Setup and Image Labeling
- Image preprocessing, augmentation, and one-hot encoding

- EfficientNetB0-based image classification using TensorFlow/Keras
- Frontend interface for image upload and result display
- Backend pipeline for processing and model inference
- Testing, Evaluation & Optimization
- Final Presentation & Documentation

## **Milestone 1: Dataset Preparation and Preprocessing (Weeks 1–2)**

### **Module 1: Dataset Setup and Image Labeling**

Tasks:

- Set up and inspect dataset of facial images or create own dataset which you can categorize or use the dataset:

<https://drive.google.com/drive/folders/1HtKbXujonS0jVNOKg0qzzAdDvCaGDd1A?usp=sharing>

- Label classes: wrinkles, dark spots, puffy eyes, clear skin.
- Ensure balanced distribution and clean samples.

Deliverables:

- Cleaned and labeled dataset
- Class distribution plot

Evaluation:

- Proper class balance
- Accurate labeling and inspection

### **Module 2: Image Preprocessing and Augmentation**

Tasks:

- Resize and normalize images (224x224).
- Apply image augmentation (flip, rotation, zoom).
- Encode class labels using one-hot encoding.

Deliverables:

- Preprocessed and augmented dataset
- Augmentation script with visualization

Evaluation:

- Augmentation quality and dataset readiness
- Class diversity retained post-augmentation

## **Milestone 2: Model Training and Evaluation (Weeks 3–4)**

### **Module 3: Model Training with EfficientNetB0**

Tasks:

- Use pretrained EfficientNetB0 for transfer learning.
- Train with categorical cross-entropy loss and Adam optimizer.
- Validate model and plot training metrics.

Deliverables:

- Trained CNN model (.h5 file)
- Accuracy and loss curves

Evaluation:

- $\geq 90\%$  classification accuracy
- Stable validation accuracy

#### **Module 4: Face Detection and Prediction Pipeline**

##### Tasks:

- Use OpenCV and Haar Cascade for face detection.
- Apply model to cropped face regions.
- Display predictions as percentages along with age.

##### Deliverables:

- Face detection and prediction script
- Test output with bounding boxes and percentages

##### Evaluation:

- Face detection accuracy
- Correct class prediction

#### **Milestone 3: Frontend and Backend Integration (Weeks 5–6)**

#### **Module 5: Web UI for Image Upload and Visualization**

##### Tasks:

- Develop frontend using Streamlit or HTML/CSS.
- Implement image upload field and output preview.
- Display labels and bounding boxes with class probability.

##### Deliverables:

- Frontend app.py or HTML/CSS script
- Responsive web interface

##### Evaluation:

- No UI lag on upload or render
- Clean annotation visualization

#### **Module 6: Backend Pipeline for Model Inference**

##### Tasks:

- Modularize inference and preprocessing code.
- Load EfficientNet model and return results to UI.
- Log predictions and bounding box data.

##### Deliverables:

- Integrated backend script
- End-to-end testing with UI

##### Evaluation:

- Seamless input-to-output flow
- $\leq 5$  seconds per image

#### **Milestone 4: Finalization and Delivery (Weeks 7–8)**

#### **Module 7: Export and Logging**

##### Tasks:

- Allow download of annotated image and CSV predictions.
- Run testing on diverse images and finalize results.

##### Deliverables:

- Export option added to frontend
- Final result logs and annotated outputs

Evaluation:

- Accurate export and log consistency
- Proper CSV formatting

#### **Module 8: Documentation and Final Presentation**

Tasks:

- Create user and developer guides (README).
- Prepare GitHub repo, slides, and walkthrough video.

Deliverables:

- Final documentation and GitHub project
- Presentation slides and video (optional)

Evaluation:

- Clear documentation structure
- Demo-ready output

#### **Evaluation Criteria:**

Milestone	Focus Area	Metric / Evaluation Method	Target / Goal
Milestone 1	Data Preparation & Preprocessing	Dataset quality, augmentation effectiveness	Balanced & clean dataset
Milestone 2	Model Performance	Accuracy & loss metrics	good test accuracy
Milestone 3	UI & Backend	Upload-to-output time & usability	≤ 5s per image
Milestone 4	Final Delivery	Export functionality & documentation	Complete & professional

#### **Tech Stack:**

Area	Tools / Libraries
Image Ops	OpenCV, NumPy, Haarcascade
Model	TensorFlow/Keras, EfficientNetB0
Dataset	Labeled facial images dataset
Frontend	Streamlit or HTML, CSS

Backend	Python, Modularized Inference
Evaluation	Accuracy, Loss, Confusion Matrix
Exporting	CSV, Annotated Image, PDF (opt.)