

# Capstone Project 1: Skin Disease Classification using CNN

## 1. Problem Statement

Develop a deep learning-based **Convolutional Neural Network (CNN)** capable of classifying dermatology images into **20 predefined skin disease categories** using the *20 Skin Diseases Dataset* from Kaggle.

The system supports dermatologists and telemedicine workflows by providing reliable automated predictions.

## 2. Objective

- Build a complete image classification workflow using CNN.
- Develop robust preprocessing pipelines for medical images.
- Train, validate, and test a multi-class classifier.
- Achieve strong performance using proper evaluation metrics.
- Deliver a clean, reproducible codebase suitable for model deployment (backend only).

## 3. Dataset Details

**Dataset:** *20 Skin Diseases Dataset*

**Source:** [Skin Diseases CNN](#)

### Dataset Download Code

```
import kagglehub
```

```
# Download the latest version of the dataset  
path = kagglehub.dataset_download("haroonalam16/20-skin-diseases-dataset")  
print("Path to dataset files:", path)
```

## Dataset Characteristics

- 20 distinct skin disease classes
- Real-world dermatological images
- Variation in lighting, texture, and quality
- No two images are identical → strong generalization required

# 4. Project Goals

## A. Data Preparation

- Load raw images, ensure proper directory structure
- Resize (e.g., 224×224) and normalize images
- Apply augmentations such as:
  - Rotation
  - Flip
  - Zoom
  - Brightness shifts
- Create train/validation/test splits

## B. Model Development

Build CNN models using either:

1. **Custom CNN**
  - a. Convolution → Activation → Pooling → Dropout → Dense
2. **Transfer Learning - optional**
  - a. MobileNetV2

- b. EfficientNetB0
- c. ResNet50

Both approaches acceptable depending on desired accuracy and compute resources.

## C. Model Training

- Train for 20–50 epochs
- Techniques to implement:
  - EarlyStopping
  - ReduceLROnPlateau
  - ModelCheckpoint
- Plot training vs. validation accuracy & loss
- Track overfitting and adjust hyperparameters

## D. Model Evaluation

Evaluate using:

- Accuracy
- Precision (per class)
- Recall (per class)
- F1-score (macro & weighted)
- Confusion matrix heatmap
- Classification report
- Error analysis:
  - Misclassified samples
  - Low-confidence predictions

## E. Model Saving

- Save trained model in .h5 (Keras) or .pt (PyTorch) format
- Save preprocessing steps

- Save class-label mapping file (`labels.json`)

## F. Inference Script (Coding Only, No UI)

A Python script that:

- Loads the trained model
- Preprocesses a given test image
- Outputs predicted class + probability

## 5. Expected Outcome

A fully-trained CNN model capable of classifying **20 skin disease categories** with high reliability.

This model can be integrated into:

- Teledermatology platforms
- Research workflows
- Backend inference services (no UI needed)
- Clinical decision-support pipelines

## 6. Final Deliverables Summary

### Mandatory

- Data preprocessing pipeline
- Custom or transfer-learning CNN model
- Training script
- Evaluation results + confusion matrix
- Saved trained model
- Documentation (README)