

# Assignment: Animal Image Classification using CNN

## Problem Statement

In this assignment, you will build a Convolutional Neural Network (CNN) to classify animal images using the **Animals10** dataset. The dataset contains 10 classes of animals (e.g. dog, cat, horse, etc.). Students will perform an end-to-end workflow: Exploratory Data Analysis (EDA), preprocessing, baseline CNN model, optimized CNN, and evaluation. The goal is to understand how CNNs work for image classification and how to tune them.

## Dataset Link

[Animals10 Dataset \(Kaggle\)](#)

## Guidelines for Students

### Data Understanding

- Download and inspect the dataset.
- Identify the classes and count the number of images per class.
- Observe image sizes, file types; note any irregularities.

### EDA (Exploratory Data Analysis)

- Display sample images from each class.
- Plot class distribution (how many images per animal category).
- Visualize image size distribution (width & height).
- Display data augmentation examples: take one image and show original plus rotated, flipped, zoomed versions side by side.

## Preprocessing

- Normalize image pixel values (e.g., scale to  $[0, 1]$ ).
- Resize images to a consistent target size (e.g.  $64 \times 64$  or  $128 \times 128$ ).
- Split dataset into training and validation sets (e.g., 80/20).
- Use appropriate data augmentation (rotation, zoom, flip).

## Model Building

### Baseline CNN

- Simple architecture: e.g. Conv2D  $\rightarrow$  MaxPooling  $\rightarrow$  Conv2D  $\rightarrow$  MaxPooling  $\rightarrow$  Flatten  $\rightarrow$  Dense  $\rightarrow$  Output.
- Use categorical crossentropy loss, Adam optimizer.
- Evaluate training & validation accuracy.

### Optimized CNN

- Make a deeper network: more Conv2D layers and/or more filters.
- Add Dropout or BatchNormalization.
- Use callbacks such as EarlyStopping and ReduceLROnPlateau to avoid overfitting.
- Experiment with different hyperparameters: number of layers, number of neurons/filters, learning rate, batch size.

## Evaluation

- Metrics: Accuracy, Confusion Matrix, Classification Report (Precision, Recall, F1-score).
- Plot training vs validation loss & accuracy curves.

- Show several example predictions: image + true label + predicted label.
- Analyze where the model fails (which classes are misclassified and why).

## **Expected Outcomes**

- Students will learn to explore image datasets, perform visualizations, and understand class imbalances.
- Students will gain skills in image preprocessing + data augmentation.
- Students will be able to build CNNs and tune hyperparameters to improve performance.
- Students will learn to interpret model results, including misclassifications and model generalization.