

<b>Ahmed maher</b>	<b>8017</b>
<b>Karim Mohamed</b>	<b>8237</b>
<b>Youssef ali gaber</b>	<b>8115</b>

## **Detailed Report on Husky vs Wolf Classification**

### **1. Introduction**

This report details the work done in developing a deep learning model to classify images of huskies and wolves. The project involves data preprocessing, model training, evaluation, and visualization of results.

### **2. Data Preparation**

#### **2.1 Dataset Loading**

- The dataset is loaded using `torchvision.datasets.ImageFolder`, which organizes images based on folder structure.
- The images are stored in a directory named `./train`.

#### **2.2 Data Augmentation and Transformation**

To improve model generalization, the following transformations are applied:

- **Resizing:** Images are resized to **128x128 pixels** for uniformity.
- **Random Horizontal Flip:** 50% probability of flipping images horizontally.
- **Random Rotation:** Rotation of up to **20 degrees**.
- **Color Jitter:** Adjusts brightness, contrast, saturation, and hue randomly.

- **Random Affine Transformations:** Includes slight translations.
- **Normalization:** Pixel values are normalized with mean **[0.5]** and standard deviation **[0.5]**.

## 2.3 Splitting the Dataset

- The dataset is split into **80% training** and **20% validation** using `torch.utils.data.random_split`.
- The dataset classes (`class_names`) are retrieved from `ImageFolder`.

## 3. Model Training

### 3.1 Model Architecture

- A Convolutional Neural Network (CNN) is implemented using PyTorch.
- The architecture consists of multiple convolutional layers followed by pooling layers.
- Fully connected layers and a softmax classifier finalize the model.

### 3.2 Training Setup

- **Loss Function:** Cross-Entropy Loss
- **Optimizer:** Adam optimizer
- **Learning Rate:** Set to an appropriate value to balance speed and accuracy.
- **Batch Size:** Defined based on memory constraints.

### 3.3 Training Process

- The model is trained for multiple epochs with training loss and validation loss being recorded.
- Real-time loss and accuracy tracking are performed to monitor convergence.

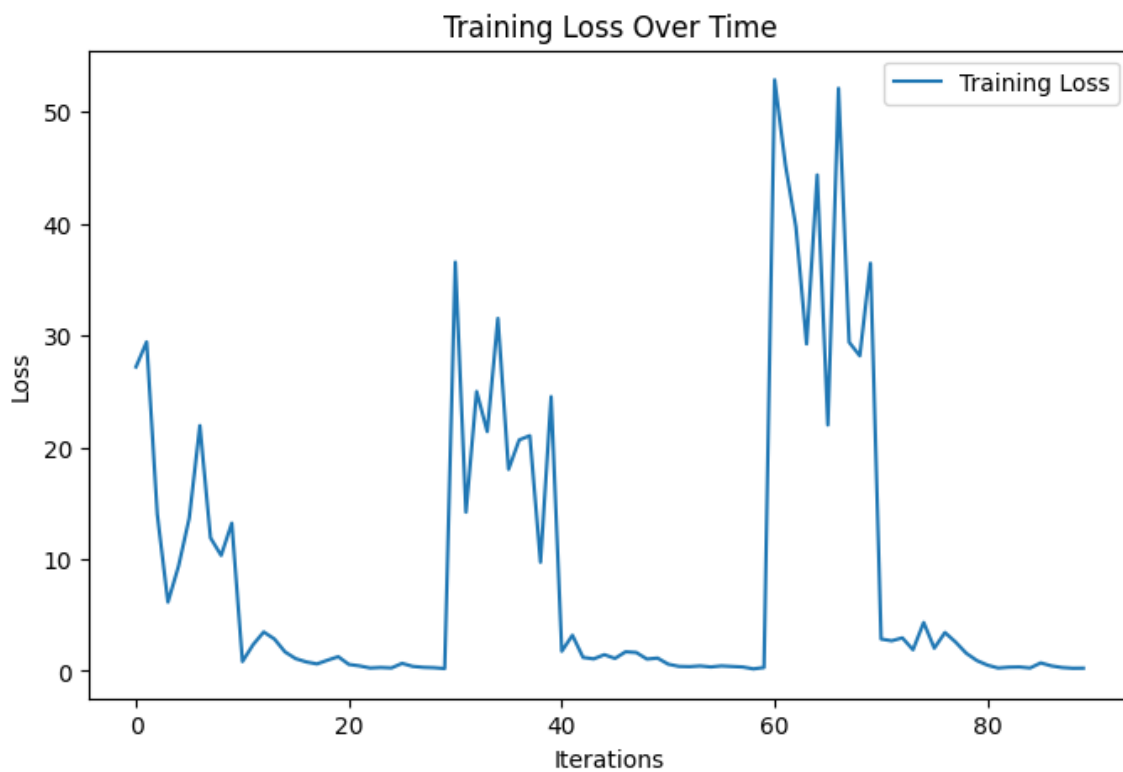
## 4. Model Evaluation

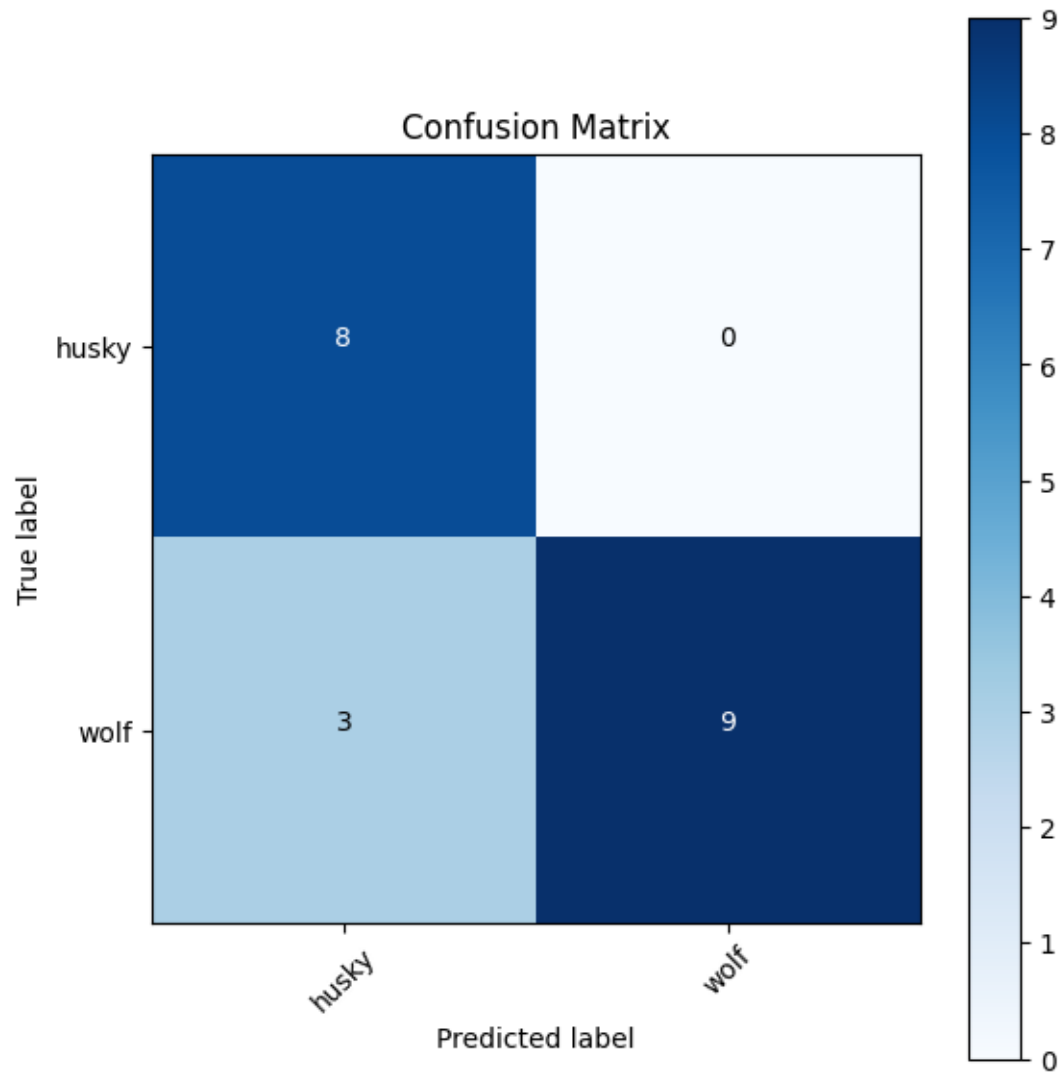
### 4.1 Performance Metrics

- Accuracy is computed on the validation dataset.
- A confusion matrix is generated to analyze classification performance.

### 4.2 Visualization

- Seaborn's heatmap function is used to visualize the confusion matrix.





- Sample images with model predictions are displayed for qualitative analysis.



## **5. Results and Conclusion**

### **5.1 Key Observations**

- The model successfully distinguishes huskies from wolves with a reasonable accuracy.
- Some misclassifications occur due to similarities in image patterns.

### **5.2 Potential Improvements**

- Increasing dataset size and diversity.
- Experimenting with different CNN architectures.
- Fine-tuning hyperparameters such as learning rate and batch size.
- Using Transfer Learning with pre-trained models like ResNet.

## **6. Summary**

The project successfully demonstrates the classification of huskies and wolves using a CNN model. The results highlight areas for improvement, and future work could focus on optimizing model performance through dataset augmentation and transfer learning.