

IDEATIONPHASE:BRAINSTORM&IDEA PRIORITIZATION

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ProjectName:AI-PoweredDogBreedIdentificationusingTransfer Learning

STEP-1:TEAMGATHERINGANDPROBLEMSELECTION

Brainstorming provides an open environment for creative thinking to lead to effective problem-solving.Forthisproject,theteamfocusedonidentifying120differentdogbreedswithhigh accuracy.

- **Identified Problem Area:** Manual dog breed identification is time-consuming, requires expert knowledge,andispronetohumanerrorduetofine-grainedvisualsimilaritiesbetweenbreeds.
- **Finalized Problem Statement:** Current manual methods are inefficient for large-scale classification.ThereisaneedforanautomatedAIsystemcapableofidentifying120breeds from single images (JPG/PNG) using deep learning techniques.

STEP-2:BRAINSTORMING,IDEALISTINGAND GROUPING

Duringthisphase,weprioritizedvolumeovervaluetogeneratediverseideasfortheclassification system.

GeneratedIdeas	IdeaGrouping
• ManualFeatureExtraction (SIFT/ORB)	• TraditionalComputerVision: Usinghand-crafted features and classical algorithms.
• SupportVectorMachines (SVM)	• MachineLearning: ClassificationusingtraditionalML algorithms.
• BuildingaCustomCNNfrom scratch	• DeepLearning: End-to-endCNN-basedclassification.
• MobileNetV2Transfer Learning	• TransferLearning: Leveragingpre-trainedCNN architectures.

STEP-3:IDEAPRIORITIZATION

To determine the best path forward, the team evaluated the ideas based on specific criteria to ensure the solution was both important and feasible.

Evaluation Criteria:

- **Accuracy Potential:** Ability to reach >60% validation accuracy.
- **Computational Efficiency:** Fast inference speed (<100ms) on standard hardware.
- **Implementation Complexity:** Ease of building a production-ready system.
- **Scalability:** Capability to handle 120 distinct classes.

Selected Approach (Chosen Solution):

The team selected **Transfer Learning using MobileNetV2**.

- **Why:** It provides higher classification accuracy, faster convergence (training in ~30-60 minutes), and efficient feature extraction while reducing overall training complexity compared to building a model from scratch.
- **Outcome:** Achieved a validation accuracy of **67.26%** with a lightweight model size suitable for web deployment.