Ref	Title	Year	Dataset descrip tion	Metho ds	Resul ts	Pros	Cons	Future Work
1	Classification of types Roasted Coffee Beans using Convolutional Neural Network Method	2024	Total Images: 1600 Dark beans: 400 Green beans: 400 Light beans: 400 Medium beans: 400 URL:htt ps://ww w.kaggle .com/dat asets/gpi osenka/c offee-be an-datas et-resize d-224-x- 224	Mobile NetV2	74.26 % 94.79 %	Mobile NetV2 allows efficient deployment on mobile/edge devices	No mention of data augmentati on or class balance handling.	Use larger and more diverse datasets with realworld conditions. Apply data augmentation and regularization to reduce overfitting.
2	Enhancing coffee bean classification: a comparative analysis of pre-trained deep	2025	Total Images: 1600	AlexNet	95.75 %	Delivers a comprehens ive may comparison across 10+ real world pretrained CNN es, lacking architecture High accuracy may overlook real world complexiti es, lacking tests on	accuracy may	Extend evaluation to include external or in-field
*	learning model		Dark beans: 400	LeNet	60.75		datasets to validate model generalization.	

			Green beans: 400 Light beans: 400 Medium beans: 400 URL:htt ps://ww w.kaggle .com/dat asets/gpi osenka/c offee-be an-datas et-resize d-224-x- 224	Google Net Mobile V2 ResNet (50) VGG Darknet	96.50 % 97.75 % 99.25 % 58.05 % 100% 98.25 %	s, guiding optimal model selection	varied environme nts or unseen data.	
				Net	%			
3	Deep Convolutional Neural Network for Coffee Bean Inspection	2021	Total images: 4626 Good beans: 2150 Bad beans: 2476	ResNet (18) Lightwe ight Student Model without KD	73%	Lightweight , explainable, and energy-effic ient CNN model with 91% accuracy, ideal for embedded systems.	Limited dataset diversity (only 4,626 images) and low generalizati on across varied real-world lighting or backgroun d conditions.	Expand dataset with varied conditions; enhance robustness and deploy system on low-power devices like Raspberry Pi or Jetson Nano.

				3.Light weight Student Model with	91%			
4	Smart agriculture: real time classification of green coffee beans by using a convolution neural network	2020	Total images: 72,000 Good beans: 36,000 Bad beans: 36,000	Custom CNN	94.63	High accuracy system (94.63%) with real-time defect detection using CNN and IP camera integration.	False positive rate (bad bean as good): 4.41%	Add back-side inspection, robotic sorting, and deploy on low-cost edge devices like Jetson Nano.
5	Custom experimental dataset of coffee beans (images in RGB, converted to grayscale)	2025	Total Images: 900 3 Categori es	ResNet- 101	100%	Extremely high accuracy (100%) achieved using transfer learning with ResNet-101 and feature integration	Dataset is experiment al and small; high accuracy may not generalize well to real-world or diverse data.	Combine multiple transfer learning techniques and fine-tune specific CNN layers to enhance generalization and scalability.
6	An Intelligent System for Coffee Grading and Disease Identification	2020	Total images: 562	CNN	89.1%	Effective even with small, real-world datasets using transfer learning and ensemble CNNs.	Performanc e affected by dataset shift due to inconsisten t imaging conditions.	Develop better domain adaptation strategies to handle dataset shift across sources.

7	Coffee Grading with Convolutional Neural Networks using Small Datasets with High Variance	2020	Total Image: 2109 12 Categori es	Cascade d CNN with CS Ensembl e	89.1%	Ensemble methods reduce variance and improve performanc e	High sensitivity to initial training conditions	Treat problem as regression instead of classification
8	Refining Content Based Segmentation for Prediction of Coffee Bean Quality	2023	Total Image: 800	Hybrid Model (Region Growin g + CBIR using GLCM texture features)	85.4%	Improved precision vs previous traditional methods	Moderate dataset size (800 images total)	Compare with deep learning (CNN-based) approaches

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