

Potato Disease Classification

Leveraging Convolutional Neural Networks to identify and categorize potato diseases from images of potato tubers

Artificial Intelligence

Hendrik Lüdemann, Linus von Maltzan



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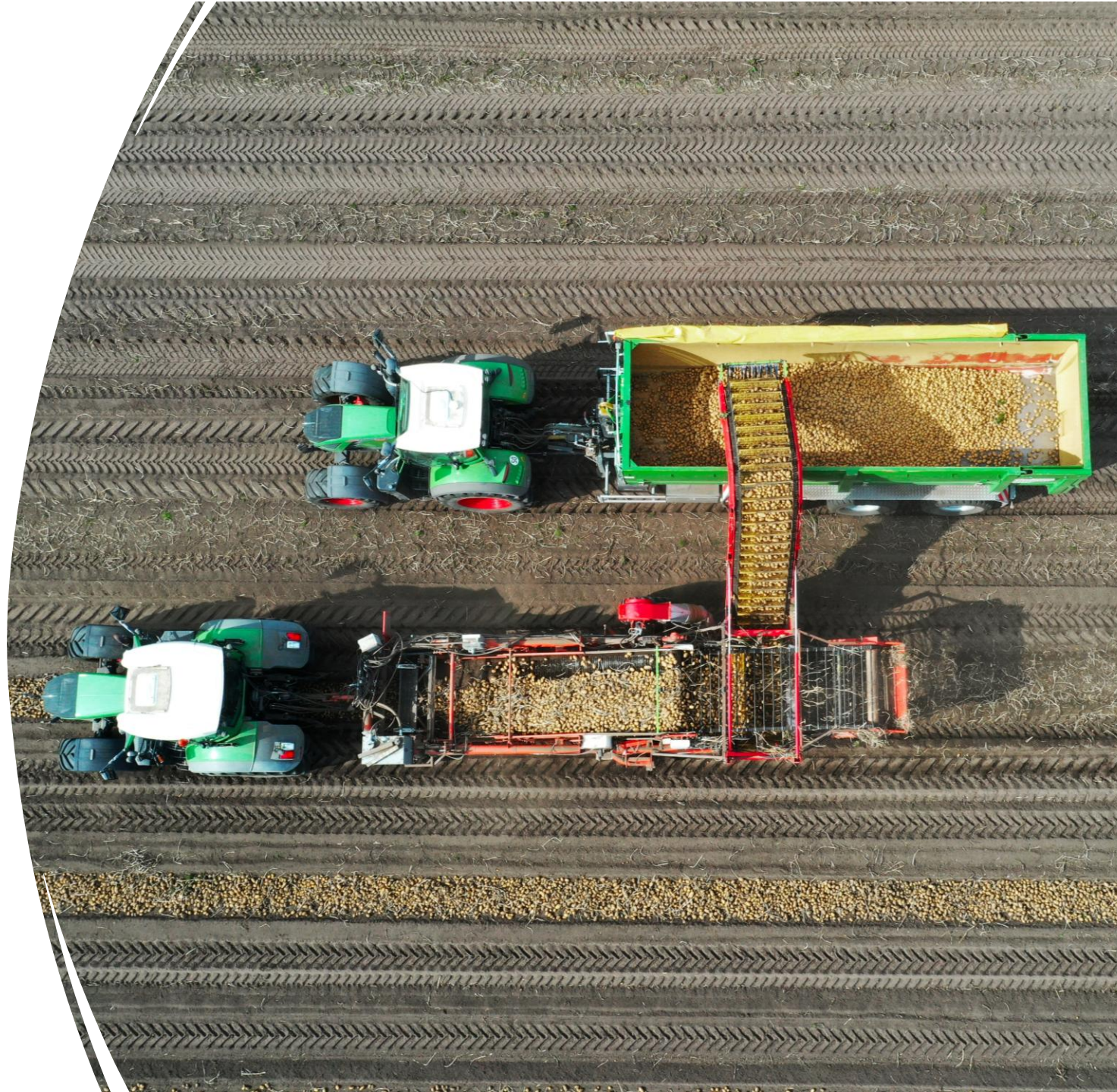
Use Case

- Potato farmers sort multiple million tubers manually
 - Shortage of workers raises wages and reduces capacities
- Optical potato sorter

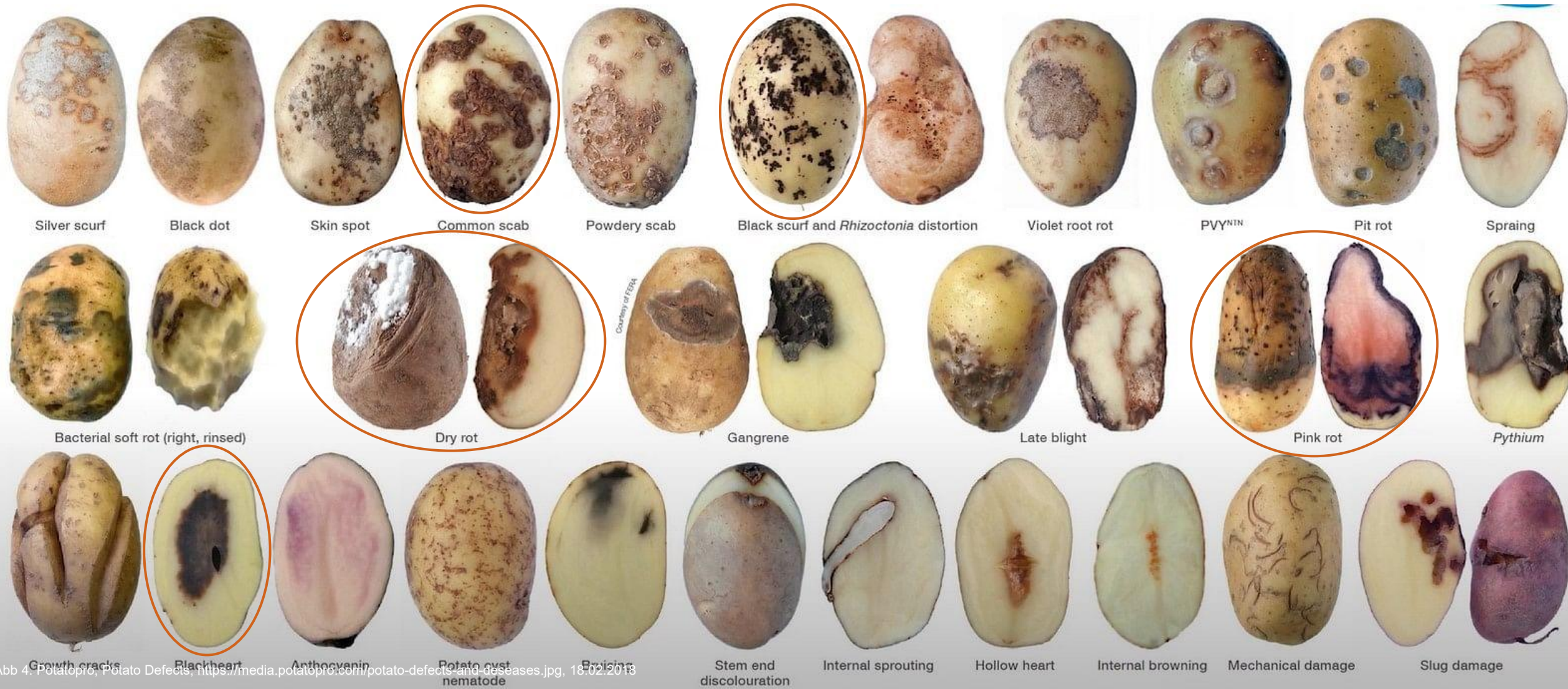


Agenda

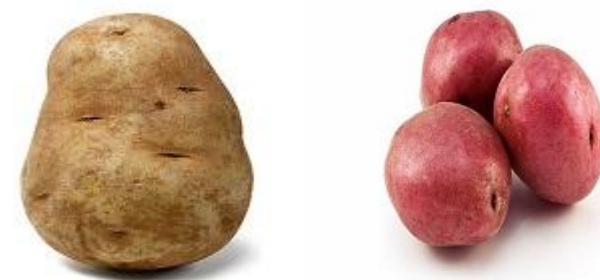
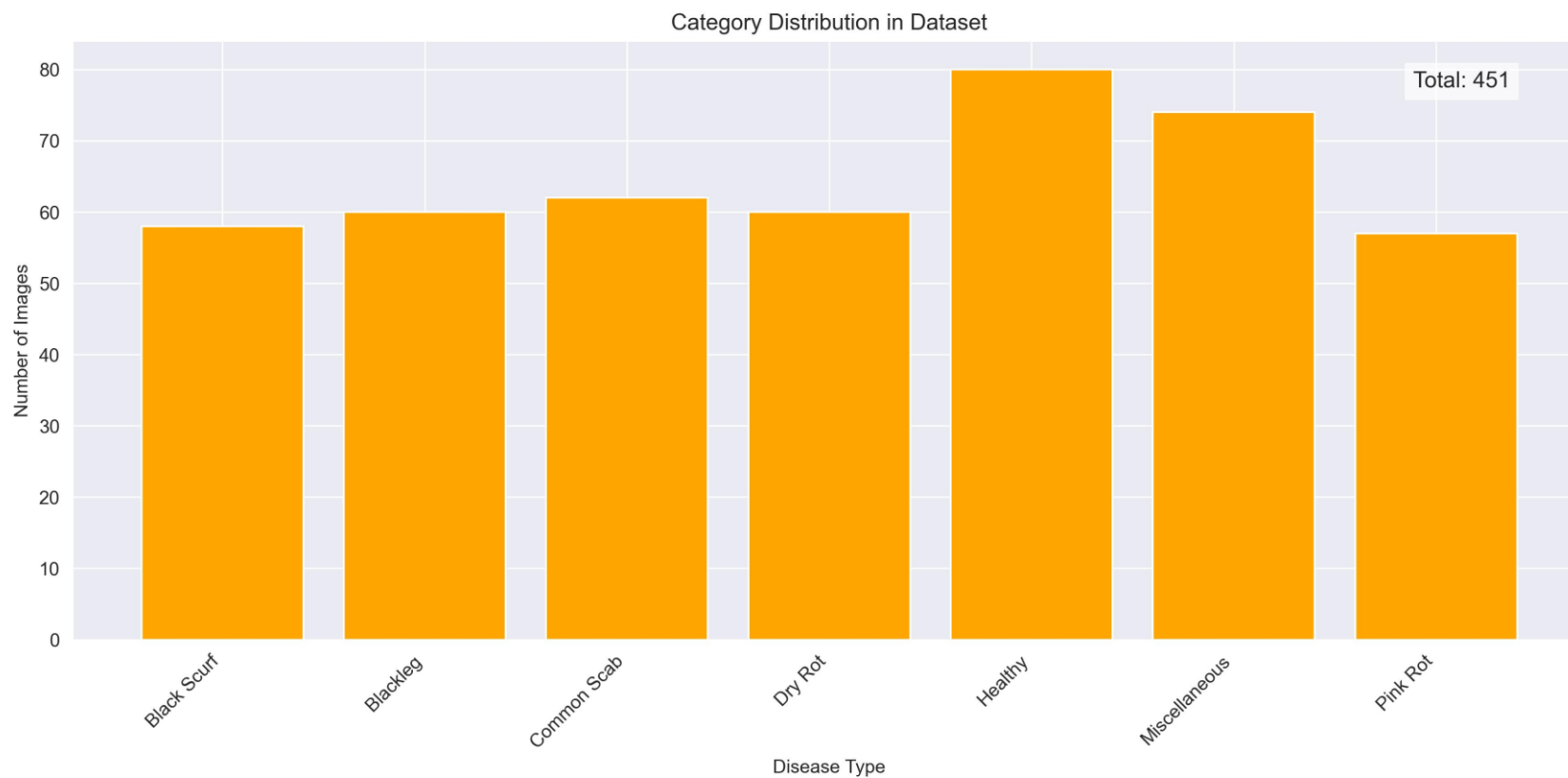
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Problem Statement



Dataset Description



Healthy



Black Scurf

Resolution Analysis

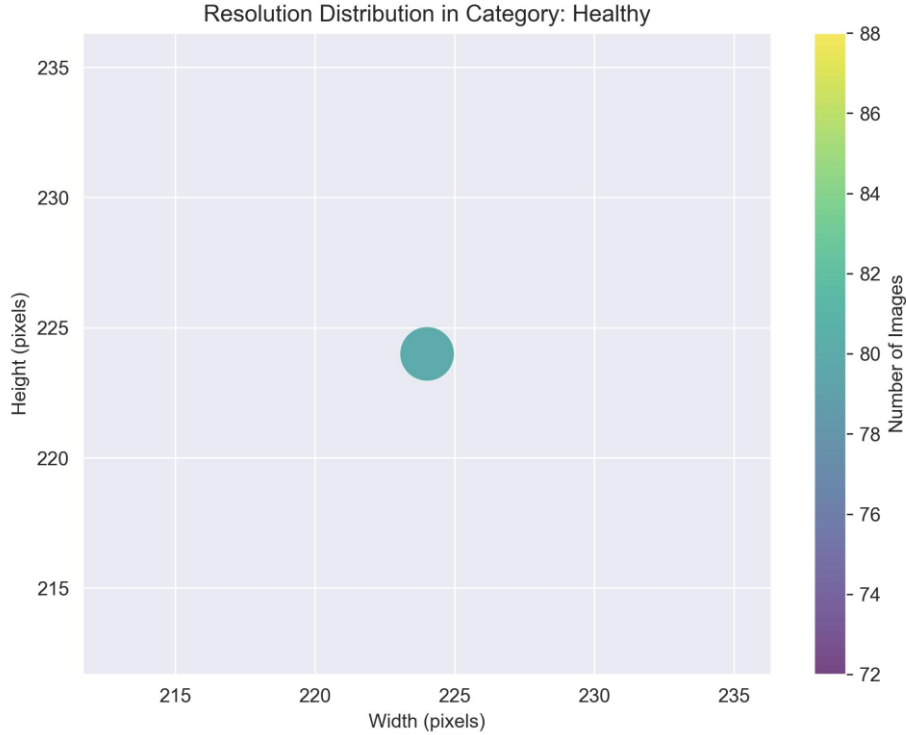
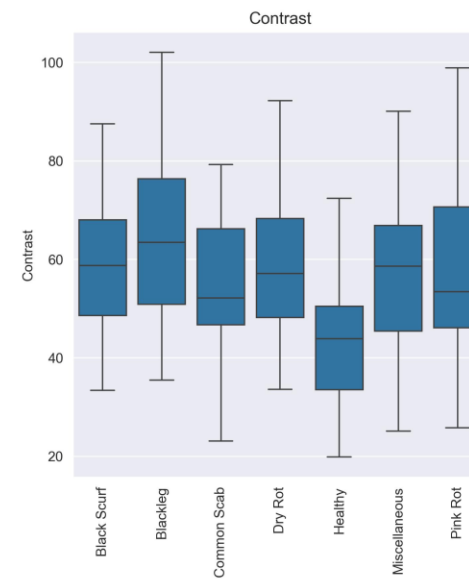
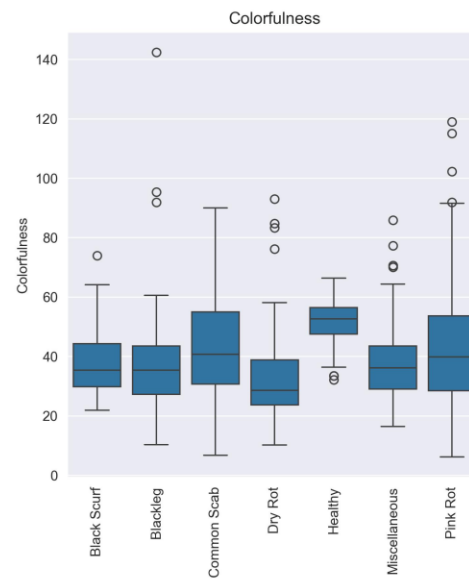
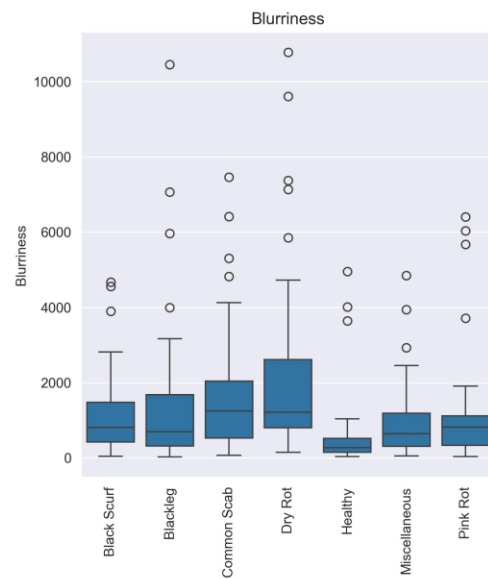
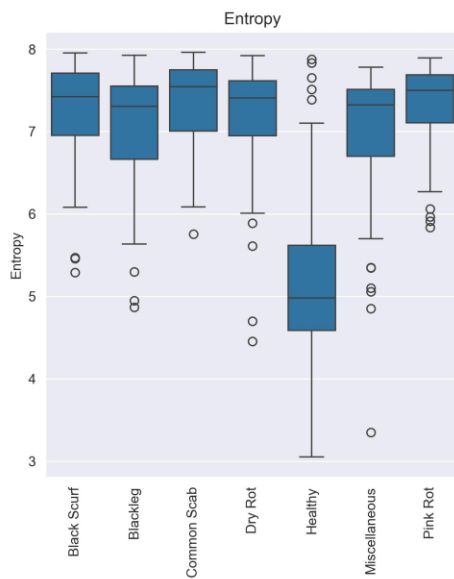
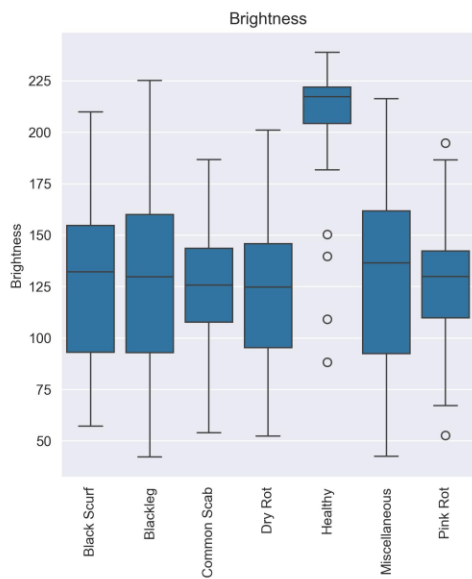
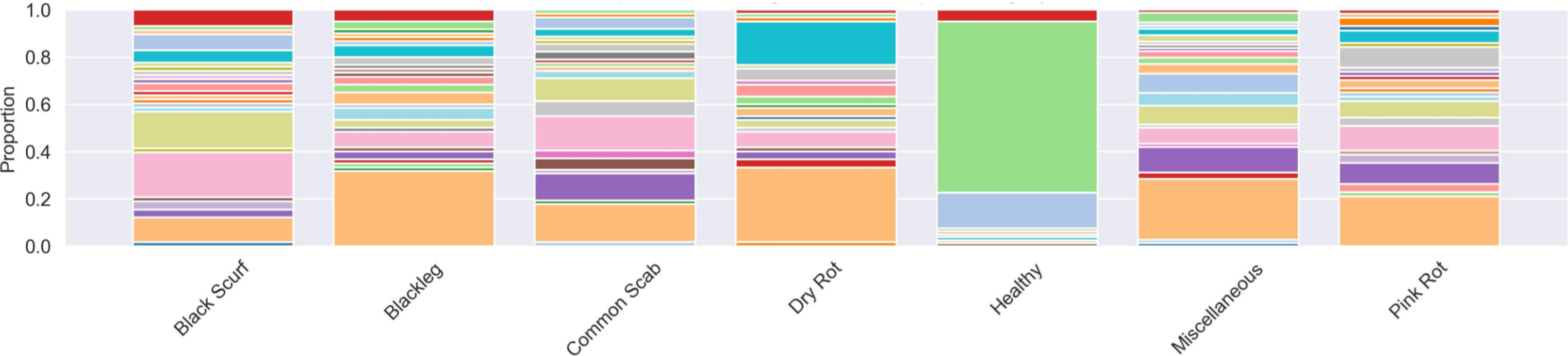


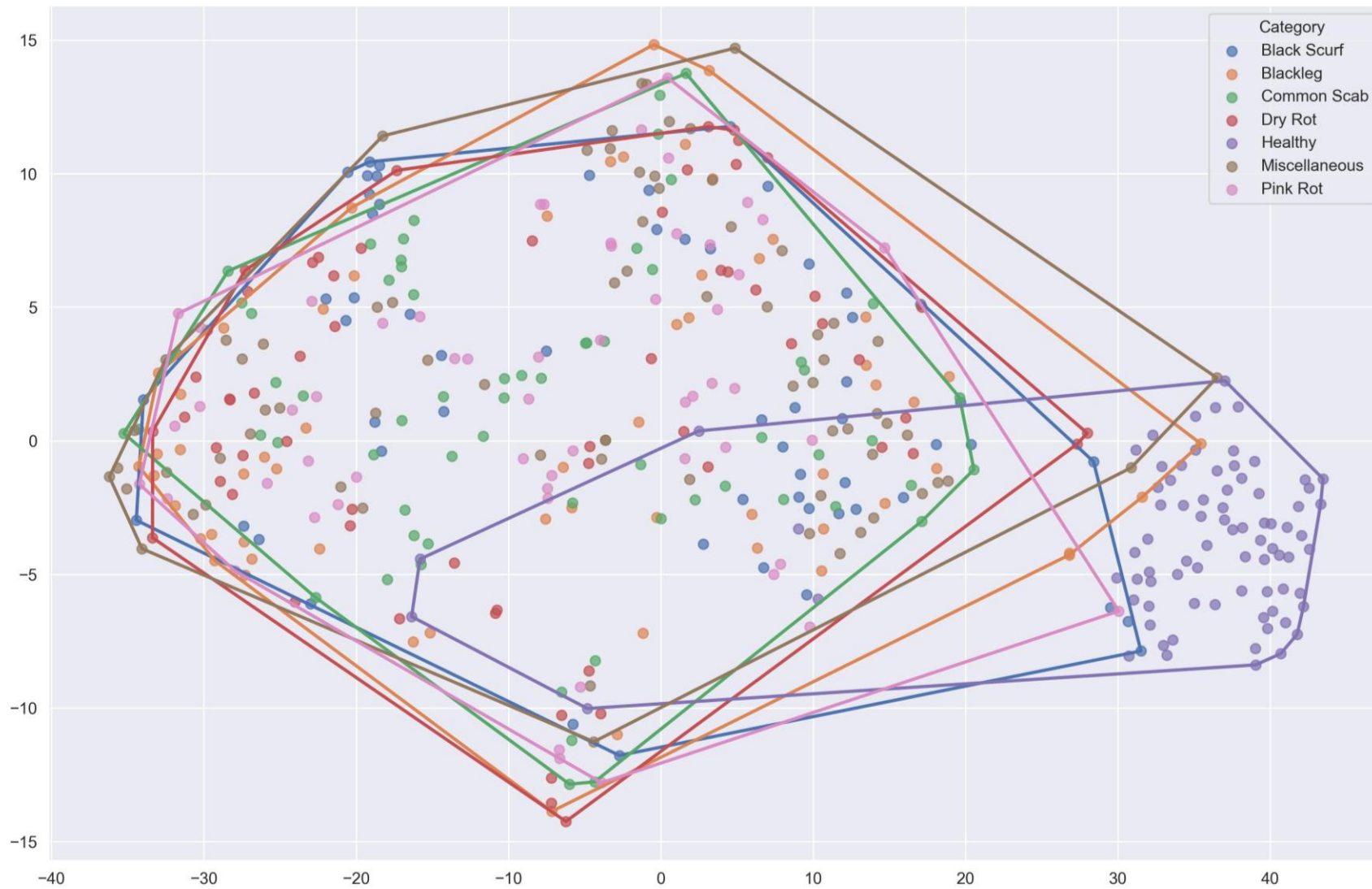
Image Property Analysis



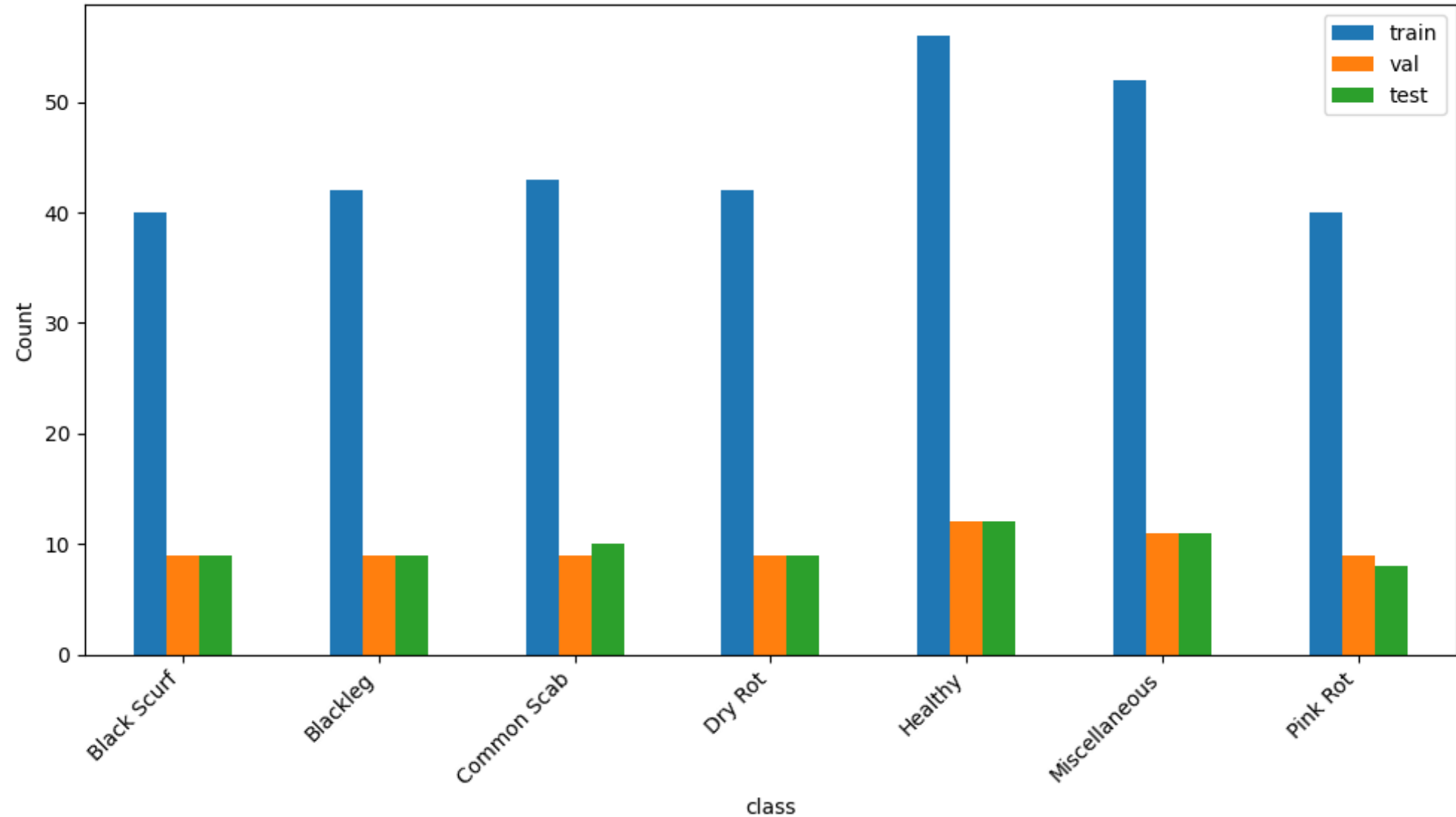
Background Color Analysis



t-SNE Projection



Dataset Splitting using Stratified Sampling



Preprocessing

Original



Variant 1



Variant 2



Original



Variant 1



Variant 2



Machine Learning Approach

Model Architecture

Input Image

Takes an RGB image (224×224) as input to the network.



Pre-processing

Applies Efficient-Net specific normalization to prepare pixel values.



Efficient-NetV2

Extracts rich visual features using a pretrained convolutional backbone.



Global Avg. Pooling

Compresses each feature map to a single value, reducing spatial dimensions.



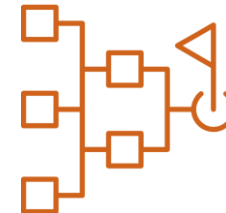
Dropout

Randomly drops 30% of features to reduce overfitting and improve generalization.



Dense Softmax

Maps features to class probabilities using a fully connected softmax layer.



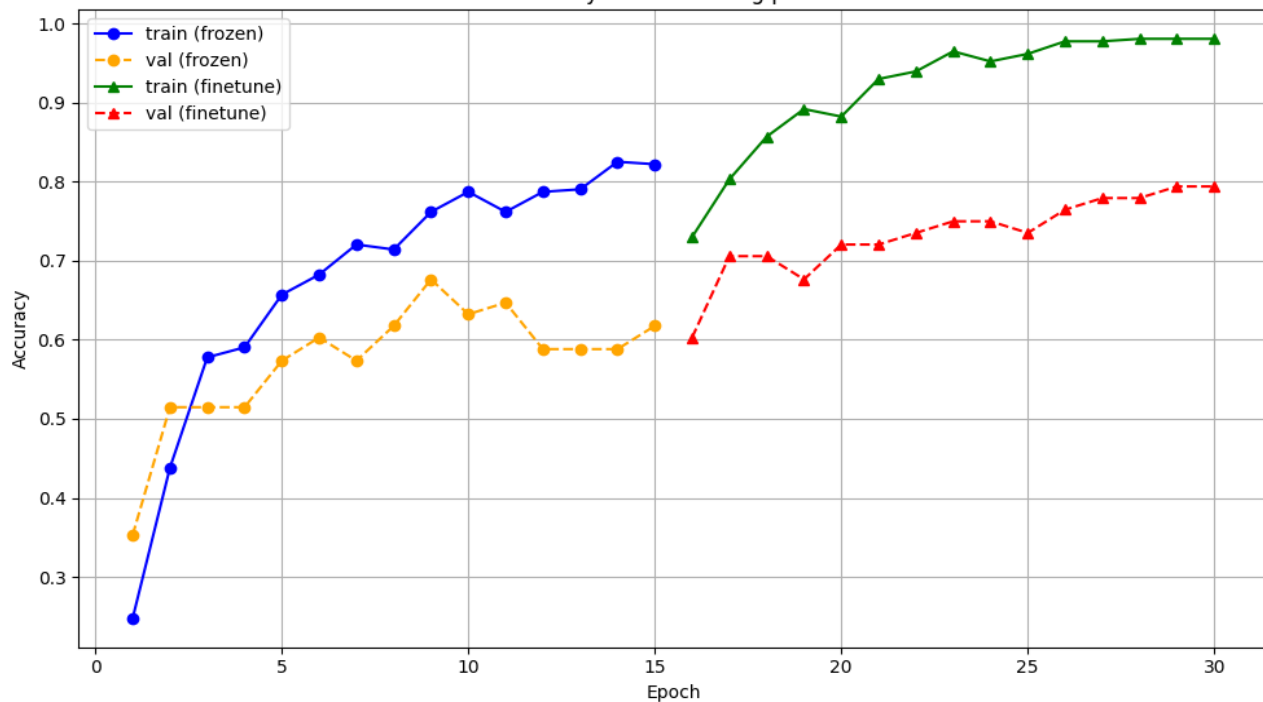
Output

Outputs a probability distribution over target classes for final prediction.

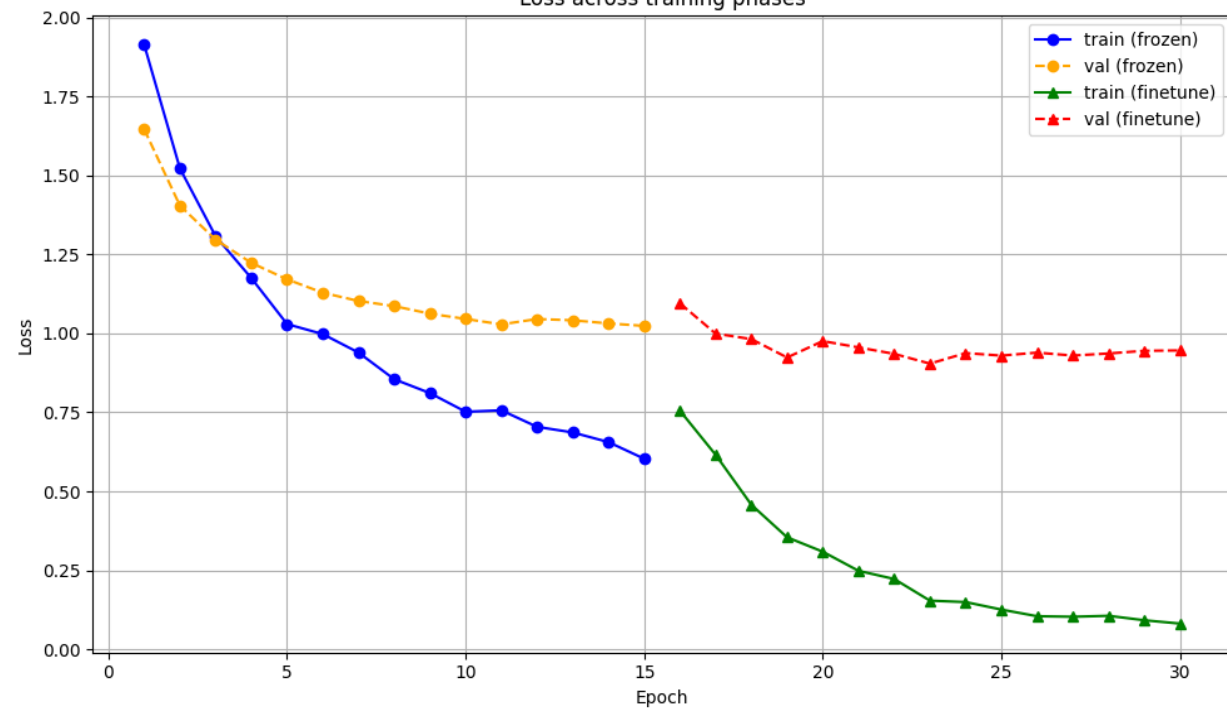


Evaluation Metrics

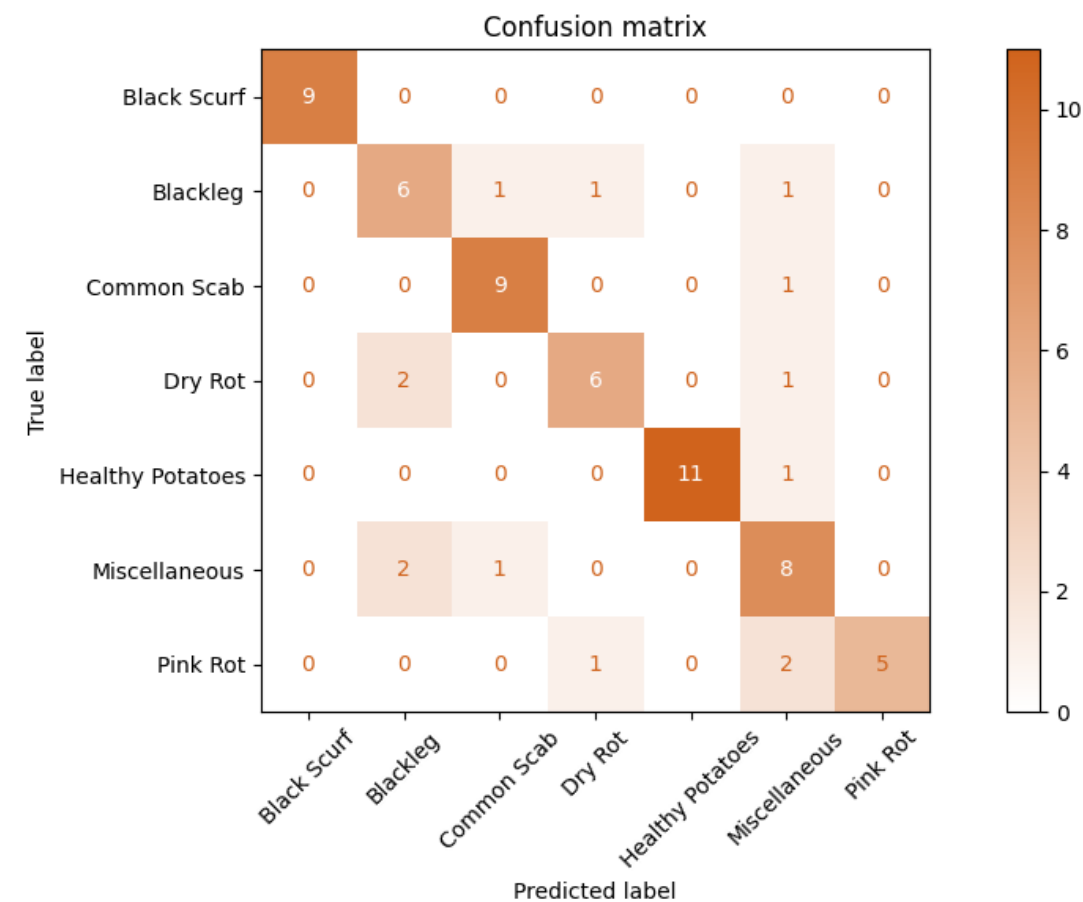
Accuracy across training phases



Loss across training phases

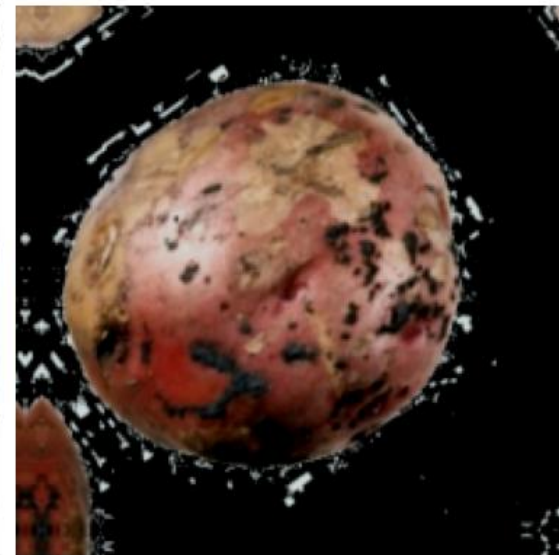
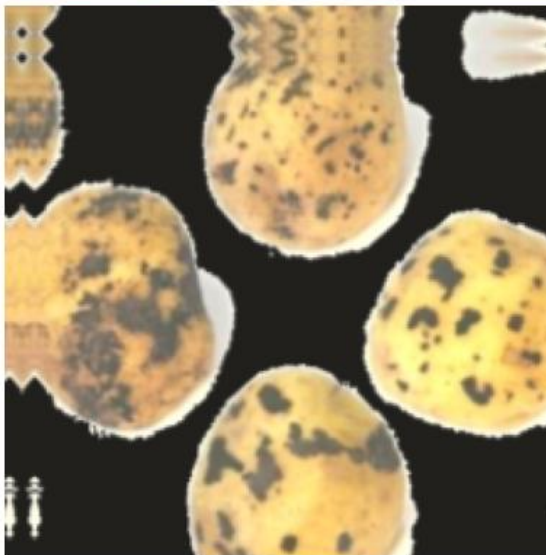
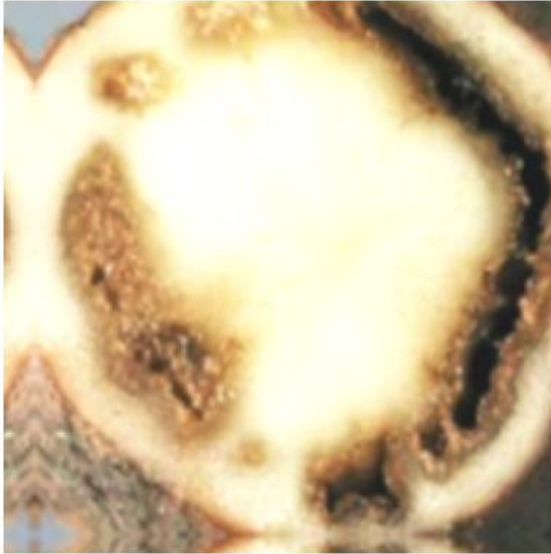


Evaluation Metrics

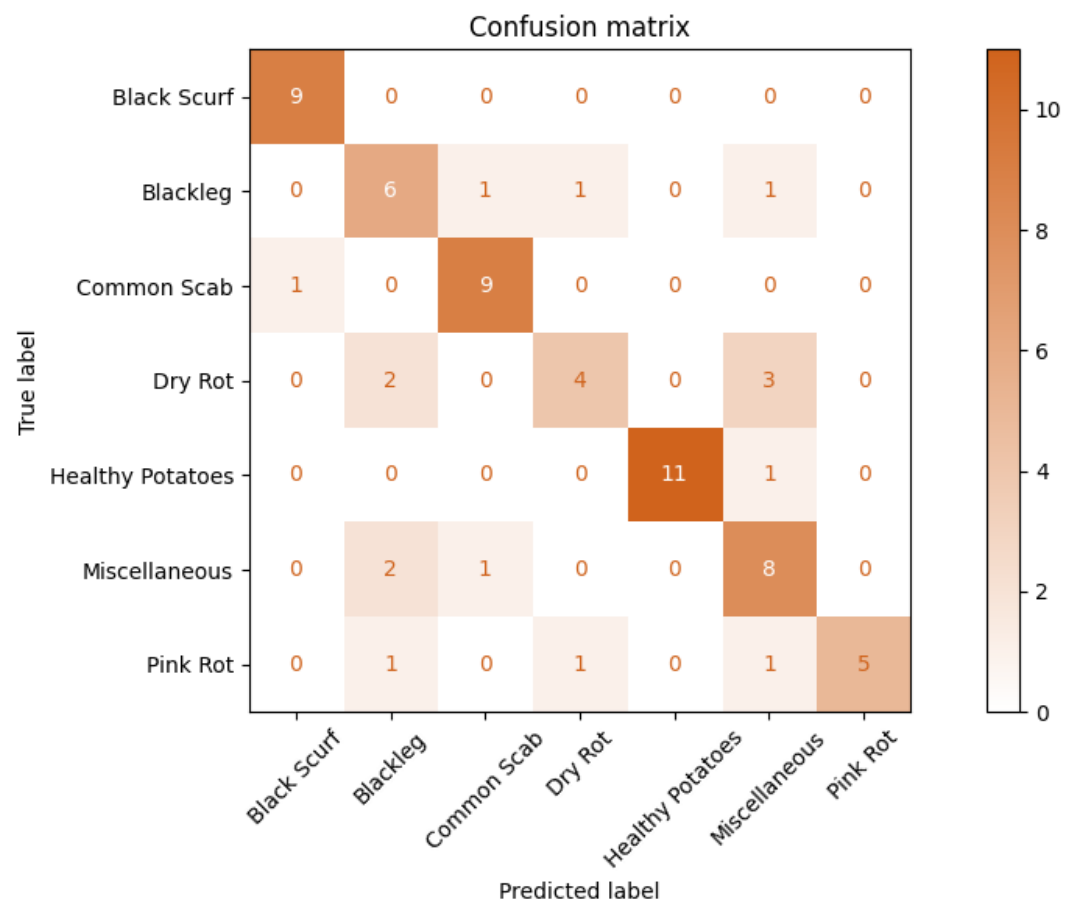


Class	Precision	Recall	F1-Score	Support
Black Scurf	1.00	1.00	1.00	9
Blackleg	0.60	0.67	0.63	9
Common Scab	0.82	0.90	0.86	10
Dry Rot	0.75	0.67	0.71	9
Healthy Potatoes	1.00	0.92	0.96	12
Miscellaneous	0.57	0.73	0.64	11
Pink Rot	1.00	0.62	0.77	8
Overall Accuracy			0.79	68

Hyperparameter-Tuning

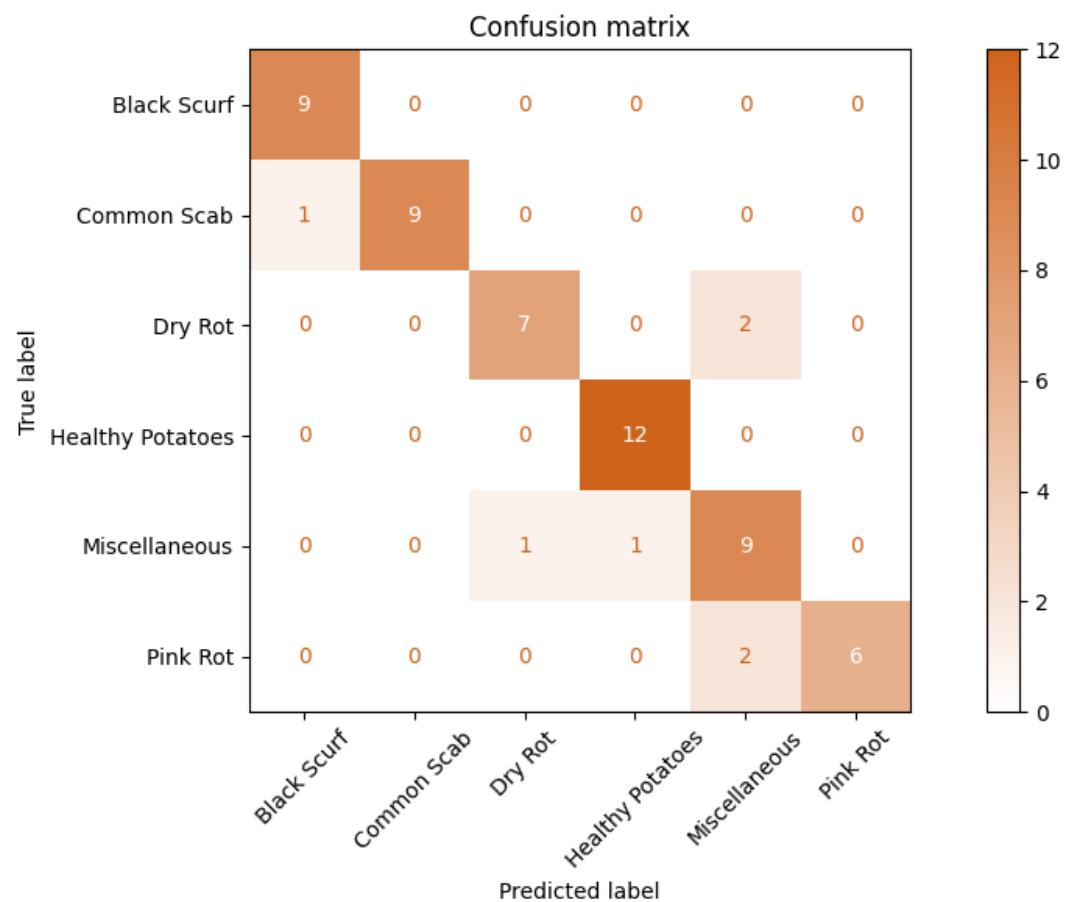


Training with blackleg



Class	Precision	Recall	F1-Score	Support
Black Scurf	0.90	1.00	0.95	9
Blackleg	0.55	0.67	0.60	9
Common Scab	0.82	0.90	0.86	10
Dry Rot	0.67	0.44	0.53	9
Healthy Potatoes	1.00	0.92	0.96	12
Miscellaneous	0.57	0.73	0.64	11
Pink Rot	1.00	0.62	0.77	8
Overall Accuracy			0.76	68

Training without blackleg



Class	Precision	Recall	F1-Score	Support
Black Scurf	0.90	1.00	0.95	9
Common Scab	0.82	0.90	0.95	10
Dry Rot	0.88	0.78	0.82	9
Healthy Potatoes	0.92	1.00	0.96	12
Miscellaneous	0.69	0.82	0.75	11
Pink Rot	1.00	0.75	0.86	8
Overall Accuracy			0.88	59

Conclusion & Future Work



Expand Dataset

Increase the volume and diversity of training images by sourcing potatoes from multiple regions, varieties and growth stages to boost model robustness.



Real-World Imagery

Gather real-world photos directly from the production line's conveyor belt, reflecting true lighting, motion blur and occlusions for reliable in-line classification.



Broaden Disease Coverage

Add images of additional potato diseases to ensure the algorithm detects and differentiates every relevant disease category.

Integrate diverse imagery and expanded disease categories to achieve robust, potato disease classification in real-world production environments.

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