

# Accessible Large-Scale Plant Pathology Recognition

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Project page : <https://github.com/mv-lab/mlplants>

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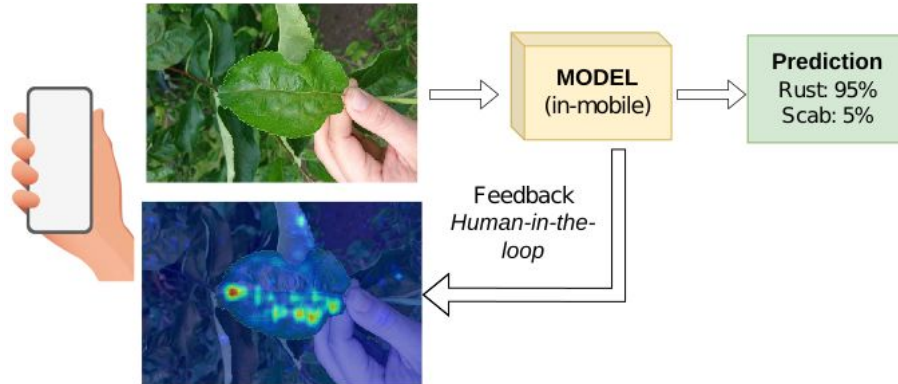


# Introduction

**Plant diseases are costly** and threaten agricultural production and food security worldwide.

Climate change is **increasing the frequency** and severity of plant diseases and pests.

**Early detection** can have a significant impact, especially in **developing countries**.



# Our proposal

We study **efficient computer vision** models that can be deployed in smartphones.

We train state-of-the-art models using the most advanced techniques for image classification and considering **real-world scenarios**: (1) Imbalanced classes, (2) Different backgrounds, light, angles, and noise conditions. (3) Different physiological age of the plants, (4) Co-occurrence of multiple diseases on the same plant, and (5) Different focus of the images.

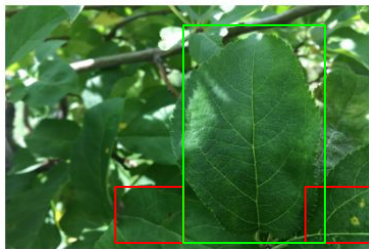


# Results and Challenges

Efficient and compact models can achieve competitive performance if trained using our empirical “tricks”, even using few-shot (FS) data, active learning (AL), or self-supervised (SS) learning.

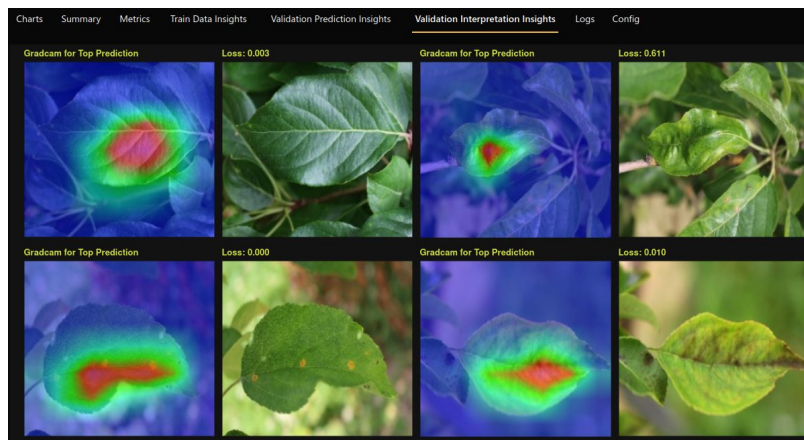
The main **challenge**, considering an in-the-wild realistic scenario is the noise in training data. For instance, **healthy** leaves surrounded by **diseased** leaves, and these with different type of diseases.

Model	img. size	ROC AUC	Mobile
1st place (SEResNeXt [10])	512	0.984	✗
2nd place (ResNeSt101 [26])	545	0.981	✗
3rd place (EffNet B7 [22])	768	0.980	✗
ResNet34 [8]	512	0.969	✗
Ours EfficientNet B0 [21]	512	0.975	✓
Ours GhostNet [7]	512	0.970	✓
Ours MobileNetV2 [21]	512	0.967	✓
Ours FS MobileNetV2 [21]	512	0.910	✓
Ours FS GhostNet [7]	512	0.900	✓
Ours SS MobileNetV2 [21]	512	0.973	✓
Ours AL MobileNetV2 [21]	512	0.950	✓



# H2O Hydrogen Torch

Our **framework** for training and evaluating state-of-the-art image classification models and integrate well-known “tricks” [2] such as custom LR schedulers, advanced augmentations, gradient accumulation, mixed precision, etc.



[1] H2O Hydrogen Torch: <https://docs.h2o.ai/h2o-hydrogen-torch/v1.2.0/>

[2] He et al. “Bag of tricks for image classification with convolutional neural networks”, CVPR 2019