# **PlantDx**

### The Ultimate Mobile Plant Doctor

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### **Overview**

PlantDx is a modern plant doctor, using artificial intelligence to diagnose plant health

#### **Upload picture**



#### Model determines PlantDx diagnosis



#### Solving two key problems:

- Frustration with the amount of time spent to research and diagnose plant care issues
- 2) **Cost** to replace dead plants and/or care for dying plants

## **Problem Space**

**Opportunity** is large and growing, without clear market leader

#### \$25bn +

US indoor plant market size, 2029

- Average household spends approx. \$600 on garden supplies per year
- Target market is a subset of plant owners that are digitally-enabled "plant parents"

#### **67%**

Adults growing or plan to grow edible plants

- House plant interest surged during pandemic
- We are betting on this **trend** continuing

#### **47%**

People reporting apprehension about purchasing plants due to care concerns (\$, time, and uncertainty)

• We strive to solve this problem, **freeing up time and saving money** 

### **Project Website: plantdx.net**





#### Plant Care: Are You Doing it Right?

7 out of 10 plants die in the care of average plant parents! We understand that finding reliable plant care advice can be challenging.

Due to the number and variety of plant species now available for residential ownership, consumer plant owners often struggle with properly caring for their plants and responding appropriately to the physical signs of potential plant stress.

#### Demo



# **Dataset / EDA: PlantVillage**

Dataset / EDA:

- ~55K w/o augmentation, ~61K w/ augmentation
- ~42K images of healthy and unhealthy labeled plant leaves
- 38 unique classes, ~25 plant types with both healthy and unhealthy labels
- Image size: 256x256

Key Takeaways

- Augmentation requirements for training and generalization (lighting, brightness, etc.)
- Nuanced health characteristics
- Model best suited for image classification is CNN

class names = [ "Apple\_\_\_Apple\_scab", "Apple\_\_\_Black\_rot", "Apple\_\_\_Cedar\_apple\_rust", "Apple healthy", "Blueberry healthy", "Cherry\_(including\_sour)\_\_\_Powdery\_mildew", "Cherry\_(including\_sour)\_\_\_healthy", "Corn\_(maize)\_\_\_Cercospora\_leaf\_spot Gray\_leaf\_spot" "Corn\_(maize)\_\_\_Common\_rust\_", "Corn (maize) Northern Leaf Blight", "Corn\_(maize)\_\_\_healthy", "Grape\_\_\_Black\_rot", "Grape\_\_\_Esca\_(Black\_Measles)", "Grape Leaf blight (Isariopsis Leaf Spot)",



#### "Healthy"

# **Modeling approach**

- Model selection
  - Focused on Convolutional Neural Networks (CNNs)
  - CNNs have become the standard model for computer vision tasks
- Experimented with 3 different CNN architectures pre-trained on ImageNet data
  - EfficientNet, MobileNet and ShuffleNet
  - Best results achieved with EfficientNet
- Transfer learning: fine-tuned all layers
- Used data augmentation to reduce model overfitting
  - Background transformation
  - The Albumentation library for several image transformations

### Final Model: EfficientNet-B0

- ➢ Number of parameters: ~5.3M
- ➢ Hyperparameters
  - Optimizer: SGD
  - Learning rate: 0.001
  - Batch size: 32
  - Epochs: 10-30

### Best performance achieved with moderate data augmentation

Model Training and Evaluation

- 1. Training on original data
- 2. Training on moderately augmented data
- 3. Training on significantly augmented data



Model Performance Trained on Moderately Augmented Data

# Model Architecture and Output



- 1. Plant Species & Health Status
- 2. Output Confidence (0-100%)
- 3. Explanation of Diagnosis
- 4. Actionable Treatment Plan and Recovery Procedure
- 5. Alternative Diagnoses (not pictured)



We believe this is an unhealthy tomato plant with early blight, with 96.9% confidence. This is a fungal disease that can cause dark spots on leaves and stems. Consider removing infected plant parts and treating with a fungicide.

# How did we get here?

### Challenges

- Data Availability
  - Solution: Data Augmentation
- Mobile Availability
  - Solution: Mobile App
- Model Deployment
  - $\circ$   $\,$  Solution: Web App  $\,$



## What's Next?

### **Future Work**

- 1. Find deployment solution with greater scalability
- 2. Optimize model performance
- 3. DATA!
- 4. More classes (more plants and diseases)

### **Mission**

### Use machine learning to improve plant and consumer happiness outcomes

- Save users time and money
- Increase satisfaction from owning plants



