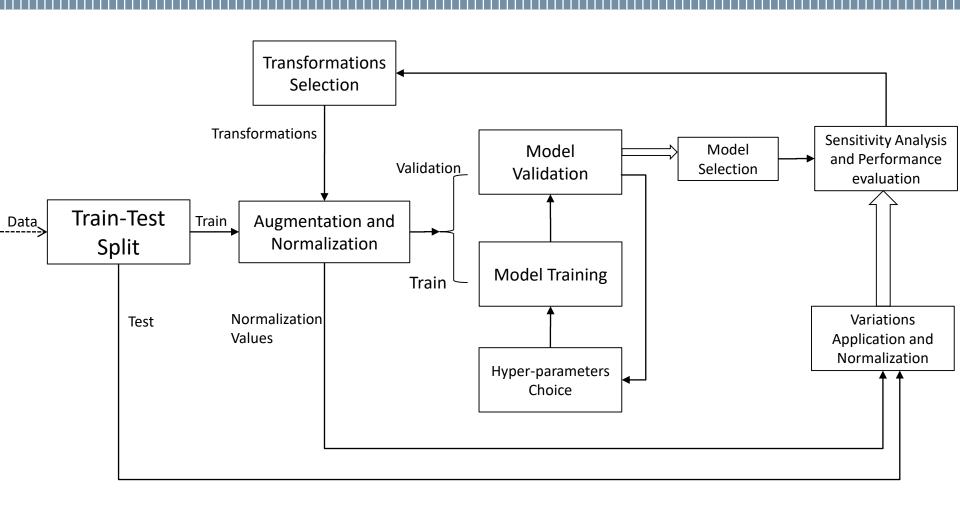


# Deep Learning approach to Tomato Plant Diseases Recognition and related Sensitivity Analysis

Deep Learning: Theory, Techniques and Applications A.Y. 2017/2018

Alessandro Erba, Mirco Manzoni, Giuseppe Mascellaro

### **Proposed Workflow Approach**



### **Dataset**



Bacterial Spot 2.127 samples



Early Blight 1.000 samples



Healthy
1.590 samples



Late Blight 1.909 samples



Leaf Mold 952 samples



Septoria leaf spot 1.771 samples



Two Spotted Spider mite 1.676 samples



Target Spot 1.404 samples



Tomato Mosaic Virus 373 samples



Tomato Yellow Leaf Curl virus 5.356 samples

### **Data Augmentation and Normalization**

Before the Training and Validation pahses data are augmented and Normalized.

Each sample is transformed taking two transformations from this pool:

- Flip Top Bottom,
- Flip Left Right,
- Rotate 90°,
- Rotate 180°,
- Rotate 270°,
- Flip Top Bottom and Rotate 90°,
- Flip Top Bottom and Rotate 270°.

Data are normalized in two different ways:

- Dataset Normalization
- Per-leaf Normalization

### **Binary Classification - Setup**

- Healthy vs Infected
- Architecture: LeNet
- Xavier initialization
- Rectified linear unit (ReLU) activation function
- Binary cross-entropy Loss
- Adam Optimizer
- 10-Fold cross Validation

## **Binary Classification - Results**

**Accuracy 99.92%** 

Precision 99.34%

Recall 99.67%

F1 - measure 99.5%

#### **Confusion Matrix**

TP = 300	FN = 1
FP = 2	TN = 3375

### **Multi-class Classification**

- AlexNet architecture
  - Train from scratch
  - Transfer Learning
- Xavier initialization
- Rectified linear unit (ReLU) function
- Cross-entropy Loss
- Adam Optimizer
- Cross-validation

### **Multi-class Classification - Results**

Accuracy: 99.85%

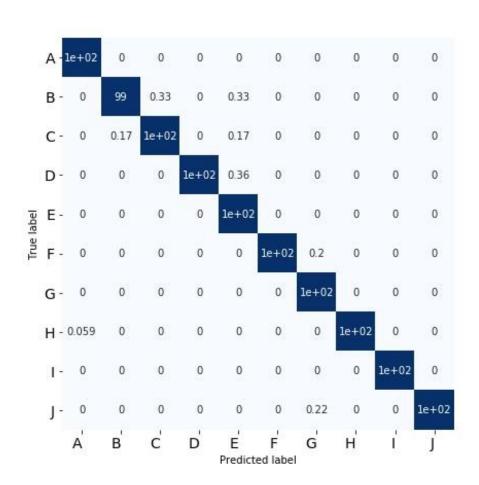
Macro precision: 99.83%

Macro recall (Trues Rate): 99.82%

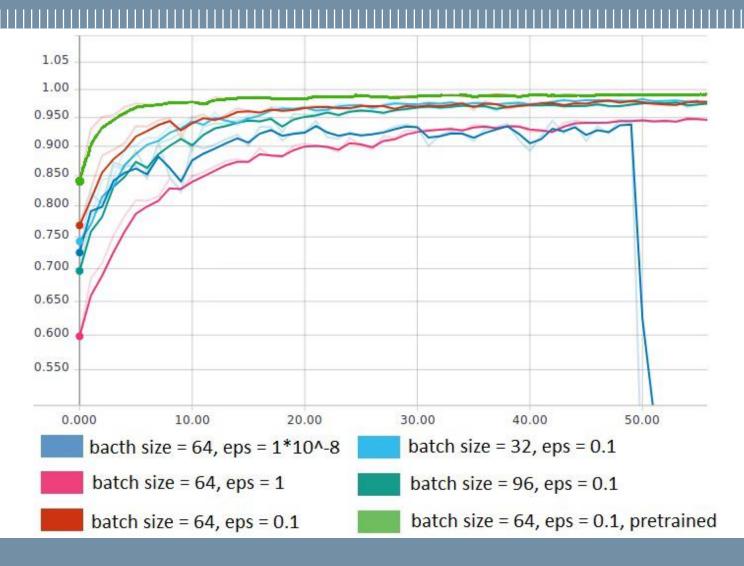
Falses Rate: 0.18%

Macro F1: 99.83%

- (A) Bacterial Spot
- (B) Early Blight
- (C) Late Blight
- (D) Leaf Mold
- (E) Septoria Leaf Spot
- (F) Two-spotted Spider Mites
- (G) Target Spot
- (H) Tomato Yellow Leaf Curl Virus
- (I) Tomato Mosaic Virus
- (J) Healthy



### Multiclass Classification - Training accuracy with different setup



### **Sensitivity Analysis Datasets**



Original dataset



Segmented dataset



Random Background Crop dataset



Random Noise Background dataset



Random Noise Background dataset

# **Sensitivity Analysis Normal Train vs Segmented Test**

Train



Original dataset

Test



Segmented dataset

Accuracy: 39.01%

Macro precision: NanN

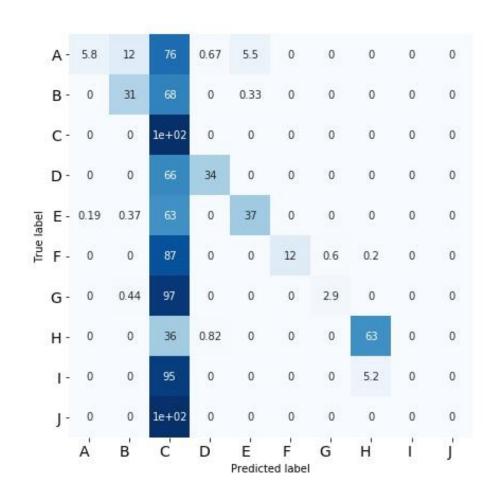
Macro recall (Trues Rate): 28.58%

Falses Rate: 71.41%

Macro F1: NaN

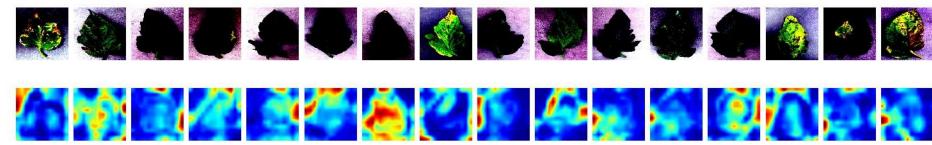
# **Sensitivity Analysis Normal Train vs Segmented Test**

- (A) Bacterial Spot
- (B) Early Blight
- (C) Late Blight
- (D) Leaf Mold
- (E) Septoria Leaf Spot
- (F) Two-spotted Spider Mites
- (G) Target Spot
- (H) Tomato Yellow Leaf Curl Virus
- (I) Tomato Mosaic Virus
- (J) Healthy

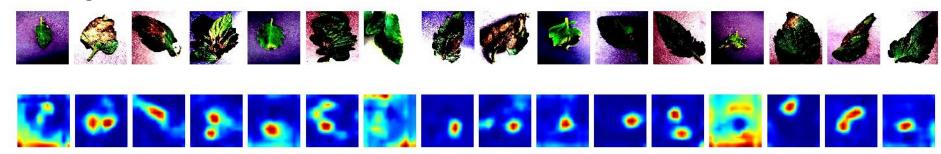


# Sensitivity Analysis Normal Train vs Normal Test Activation Visualization

#### **Bacterial Spot Class**



#### Late Blight Class



## Sensitivity Analysis - Segmented Test vs Normal Train

Train



Segmented dataset

Test



Original dataset

Accuracy: 61.79%

Macro precision: 73.02%

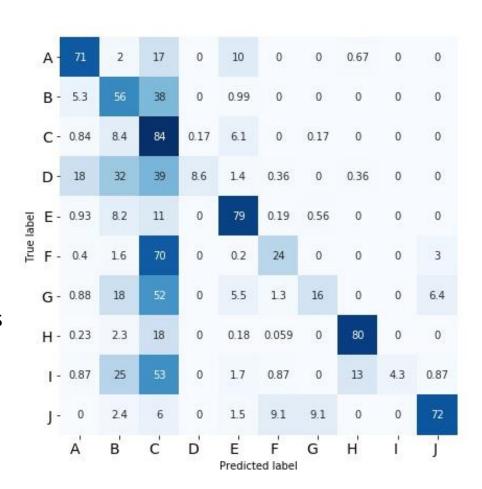
Macro recall (Trues Rate): 49.41%

Falses Rate: 50.59%

Macro F1: 48.69%

# **Sensitivity Analysis Segmented Train vs Normal Test**

- (A) Bacterial Spot
- (B) Early Blight
- (C) Late Blight
- (D) Leaf Mold
- (E) Septoria Leaf Spot
- (F) Two-spotted Spider Mites
- (G) Target Spot
- (H) Tomato Yellow Leaf Curl Virus
- (I) Tomato Mosaic Virus
- (J) Healthy



# **Sensitivity Analysis Random Background Crop Train vs Normal Test**

Train



Random Background Crop dataset

Test



Original dataset

Accuracy: 76.92%

Macro precision: 76.92%

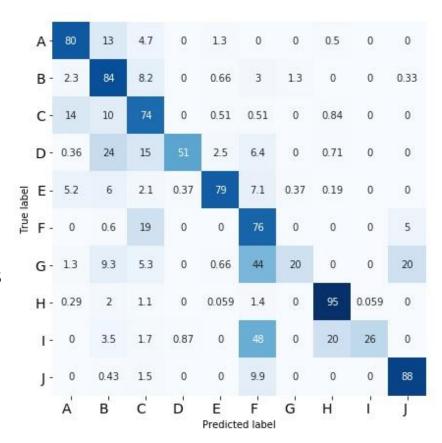
Macro recall (Trues Rate): 79.43%

Falses Rate: 32.70%

Macro F1: 67.15%

# Sensitivity Analysis Random Background Crop Train vs Normal Test

- (A) Bacterial Spot
- (B) Early Blight
- (C) Late Blight
- (D) Leaf Mold
- (E) Septoria Leaf Spot
- (F) Two-spotted Spider Mites
- (G) Target Spot
- (H) Tomato Yellow Leaf Curl Virus
- (I) Tomato Mosaic Virus
- (J) Healthy



# Sensitivity Analysis Random Noise Background Train vs Normal Test

Train



Random Noise Background dataset

Test



Original dataset

Accuracy: 57.89 %

Macro precision: 66.68%

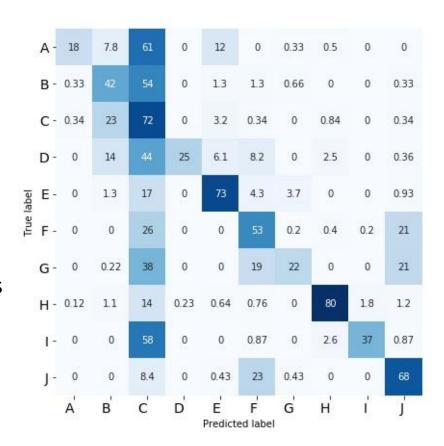
Macro recall (Trues Rate): 49.01%

Falses Rate: 50.99%

Macro F1: 49.93%

# Sensitivity Analysis Random Noise Background Train vs Normal Test

- (A) Bacterial Spot
- (B) Early Blight
- (C) Late Blight
- (D) Leaf Mold
- (E) Septoria Leaf Spot
- (F) Two-spotted Spider Mites
- (G) Target Spot
- (H) Tomato Yellow Leaf Curl Virus
- (I) Tomato Mosaic Virus
- (J) Healthy



# Sensitivity Analysis – Per Leaf Normalization Random Background Image Train vs Normal Test

Train



Random Noise Background dataset

**Per Leaf Normalization** 

Test



Original dataset

Accuracy: 97.77%

Macro precision: 97.52%

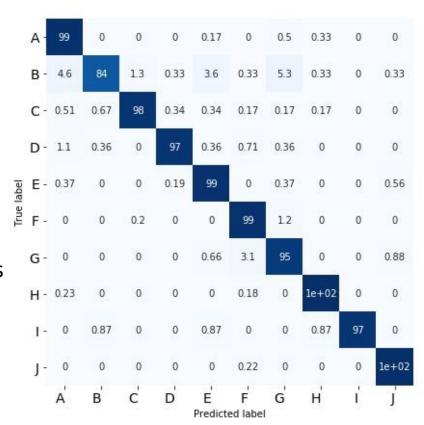
Macro recall (Trues Rate): 96.69%

Falses Rate: 3.31%

Macro F1: 97.04%

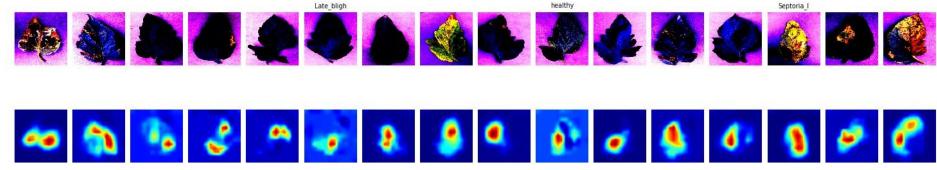
# Sensitivity Analysis Random Background Image Train vs Normal Test

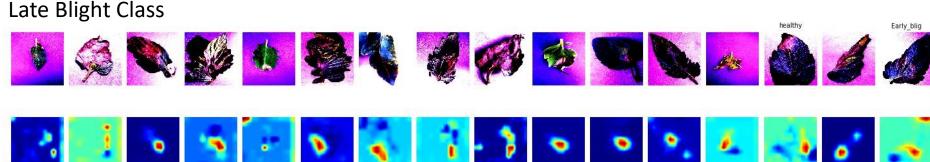
- (A) Bacterial Spot
- (B) Early Blight
- (C) Late Blight
- (D) Leaf Mold
- (E) Septoria Leaf Spot
- (F) Two-spotted Spider Mites
- (G) Target Spot
- (H) Tomato Yellow Leaf Curl Virus
- (I) Tomato Mosaic Virus
- (J) Healthy



### Random Background Image Train vs Normal Test **Activation Visualization**

#### **Bacterial Spot Class**





# **Extension: Test against Real Plantation Images**

We built a set of 256 found on internet.



Bacterial Spot 26 samples



Early Blight 41 samples



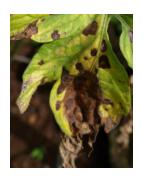
Healthy 27 samples



Late Blight 31 samples



Leaf Mold 33 samples



Septoria leaf spot 36 samples



Two Spotted Spider mite 12 samples



Target Spot 12 samples



Tomato Mosaic Virus 18 samples



Tomato Yellow Leaf Curl virus 29samples

# **Test against Real Plantation Images Results**

AlexNet

Accuracy: 29.81%

Macro precision: 30.36%

Macro recall (Trues Rate): 25.86%

Falses Rate: 74.13%

VGG11

Accuracy: 35.85%

Main Limitation: the set is not validated by a domain expert.