

Model Documentation: Strawberry Leaf Health Classifier

Model Purpose

This Convolutional Neural Network (CNN) is designed to classify images of strawberry leaves as either:

-  Healthy
-  Unhealthy

If the leaf is predicted to be unhealthy, a secondary detection stage (e.g., Roboflow) can classify the specific disease affecting the leaf.

Model Details

Attribute	Value
Model Type	Binary Classification CNN
Input Shape	(height, width, 3) (RGB)
Output	Single sigmoid value in [0,1]
Threshold	0.75 (above = healthy)
File Format	Keras HDF5 .h5
Filename	my_model.h5

Model Layers

```

input_shape = (150, 150, 3)

model = Sequential()

# طبقات Conv2D مع زيادة عدد الفلاتر
model.add(Conv2D(512, (3, 3), padding='same', input_shape=input_shape, activation='relu'))
model.add(Conv2D(512, (3, 3), padding='same', activation='relu'))
model.add(Conv2D(512, (3, 3), padding='same', activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))

# طبقات Conv2D مع عدد أقل من الفلاتر
model.add(Conv2D(256, (3, 3), padding='same', activation='relu'))
model.add(Conv2D(256, (3, 3), padding='same', activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))

# طبقات Conv2D مع عدد أقل من الفلاتر
model.add(Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))

# طبقة Conv2D مع مخرجات أقل
model.add(Conv2D(64, (3, 3), padding='same', activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))

# Flatten
model.add(Flatten())

# طبقة Dense مع Regularizer و Dropout
model.add(Dense(256, activation='relu', kernel_regularizer=regularizers.l2(0.01)))
model.add(Dropout(0.5))

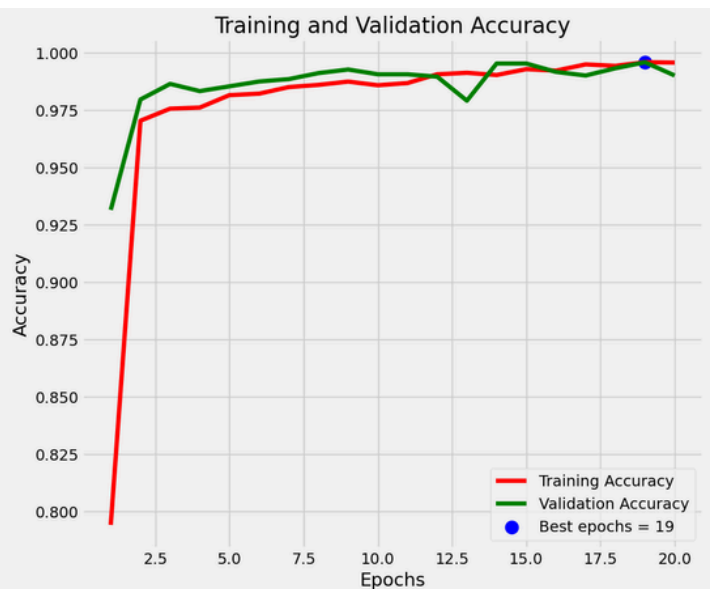
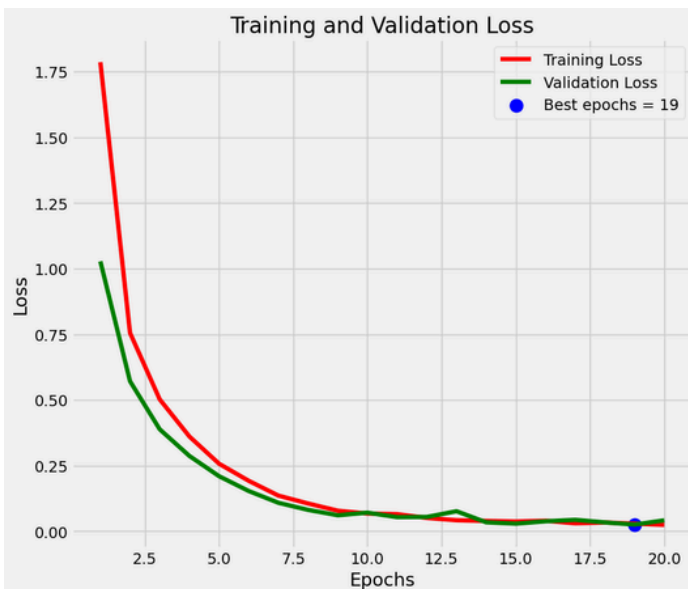
# طبقة التمثيل
model.add(Dense(1, activation='sigmoid'))

# optimizer Adamax
model.compile(optimizer=Adamax(learning_rate=0.001),
              loss='binary_crossentropy',
              metrics=['accuracy'])

```

Training & Validation Performance

- Model shows stable training with minimal overfitting.
- Loss decreases consistently and accuracy plateaus after ~10 epochs.



Input Requirements

- Input Type: RGB image of a strawberry leaf
- Input Size: Must be resized to match the model's expected shape (e.g. (150, 150) or whatever `model.input_shape[1:3]` returns)
- Preprocessing:
 - Resize to model input shape

- **Normalize pixel values: divide by 255.0**
- **Expand dimensions to match shape (1, height, width, 3)**

```
img = img.resize(model.input_shape[1:3])
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array /= 255.0
```

Output

The model returns a single float between 0 and 1:

- **0.0 → 0.74: interpreted as Unhealthy**
- **0.75 → 1.0: interpreted as Healthy**

You can convert the output to a health score:

```
health_score = (1 - prediction) * 100
```

Example Inference

```
prediction = model.predict(img_array)[0][0]
if prediction ≥ 0.75:
    print("Healthy Leaf")
else:
    print("Unhealthy Leaf")
```

Extended Diagnosis

If the leaf is unhealthy, the system integrates with Roboflow's Object Detection API to identify the exact disease class. Example response classes include:

- *Angular Leafspot*
- *Anthracnose Fruit Rot*
- *Gray Mold*
- etc.

Each class is linked to metadata (cause, treatment, prevention) stored in a Python dictionary for real-time feedback.

Deployment Notes

- This model is built using TensorFlow/Keras, easily portable to backends using:
 - Flask / FastAPI (via `load_model` and `predict`)
 - TensorFlow.js (after conversion using `tensorflowjs_converter`)
- Avoid Streamlit in production; instead:
 - Load and use the model inside a Python API
 - Accept image files from React frontend
 - Send predictions + disease analysis JSON back to the client

Example Response (to frontend)

```
{
  "health_score": 62.4,
  "is_healthy": false,
  "detected_disease": "Gray Mold",
  "confidence": 87.5,
  "disease_info": {
    "cause": "Fungal infection (Botrytis cinerea)",
    "treatment": "Apply fungicides like fenhexamid, remove and destroy infected parts",
    "prevention": "Improve ventilation, avoid excess moisture, remove plant debris"
  }
}
```

Dependencies

- tensorflow>=2.x
- PIL
- numpy
- requests (for Roboflow API)
- inference_sdk (optional)

Notes

- Adjust the health threshold based on model performance if needed.
- Ensure consistent image quality in production (e.g., daylight, clear background).
- Roboflow detection is optional but adds high value for disease-specific interventions.

References & Resources

-  Kaggle Notebook: [Strawberry Health CNN by abdomogahed](#)
-  Roboflow Docs: <https://docs.roboflow.com>
-  TensorFlow Documentation: https://www.tensorflow.org/api_docs
-  Pillow Docs: <https://pillow.readthedocs.io>
-  NumPy Docs: <https://numpy.org/doc>
-  Requests Library: <https://docs.python-requests.org>
-  Keras API: <https://keras.io/api>