

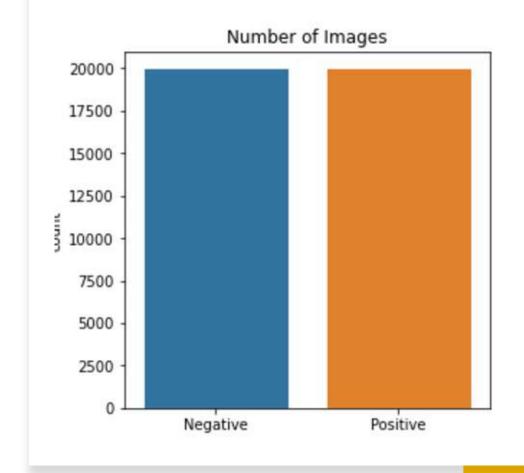
Goal of Study

- The goal of the study is to develop an efficient system for detecting surface cracks in images using convolutional neural networks (CNNs).
- This system aims to address critical issues in industries such as manufacturing and construction by preventing structural failures caused by undetected cracks.
- Additionally, it focuses on leveraging PySpark for large-scale data processing across multiple virtual machines, ensuring scalability and efficiency.



Data Structure

- We collected the data from 'Kaggle'
- Total Number of Images: 40,000 images
- Classes: The dataset is divided into 2 classes:
- Positive Crack Images: 20,000 images
- Negative Crack Images: 20,000 images



Data Splitting

Training Set: 70% (28,000 images)

Validation Set: 15% (6,000 images)

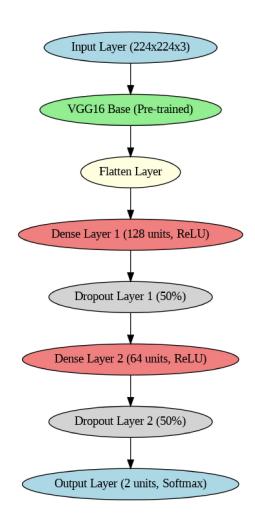
• **Test Set**: 15% (6,000 images)



Models

- VGG16
- CNN
- RestNet50
- LeNet

VGG 16



| odel: "functional_1" | | | |
|---------------------------------------|---------------------|-----------|--|
| Layer (type) | Output Shape | Param # | |
| <pre>input_layer_1 (InputLayer)</pre> | (None, 34, 34, 3) | 0 | |
| block1_conv1 (Conv2D) | (None, 34, 34, 64) | 1,792 | |
| block1_conv2 (Conv2D) | (None, 34, 34, 64) | 36,928 | |
| block1_pool (MaxPooling2D) | (None, 17, 17, 64) | 0 | |
| block2_conv1 (Conv2D) | (None, 17, 17, 128) | 73,856 | |
| block2_conv2 (Conv2D) | (None, 17, 17, 128) | 147,584 | |
| block2_pool (MaxPooling2D) | (None, 8, 8, 128) | 0 | |
| block3_conv1 (Conv2D) | (None, 8, 8, 256) | 295,168 | |
| block3_conv2 (Conv2D) | (None, 8, 8, 256) | 590,080 | |
| block3_conv3 (Conv2D) | (None, 8, 8, 256) | 590,080 | |
| block3_pool (MaxPooling2D) | (None, 4, 4, 256) | 0 | |
| block4_conv1 (Conv2D) | (None, 4, 4, 512) | 1,180,160 | |
| block4_conv2 (Conv2D) | (None, 4, 4, 512) | 2,359,808 | |
| block4_conv3 (Conv2D) | (None, 4, 4, 512) | 2,359,808 | |

| block3_pool (MaxPooling2D) | (None, 4, 4, 256) | .0 |
|----------------------------|-------------------|-----------|
| block4_conv1 (Conv2D) | (None, 4, 4, 512) | 1,180,160 |
| block4_conv2 (Conv2D) | (None, 4, 4, 512) | 2,359,808 |
| block4_conv3 (Conv2D) | (None, 4, 4, 512) | 2,359,808 |
| block4_pool (MaxPooling2D) | (None, 2, 2, 512) | 0 |
| block5_conv1 (Conv2D) | (None, 2, 2, 512) | 2,359,808 |
| block5_conv2 (Conv2D) | (None, 2, 2, 512) | 2,359,808 |
| block5_conv3 (Conv2D) | (None, 2, 2, 512) | 2,359,808 |
| block5_pool (MaxPooling2D) | (None, 1, 1, 512) | 0 |
| flatten_1 (Flatten) | (None, 512) | 0 |
| dense_3 (Dense) | (None, 128) | 65,664 |
| dropout_2 (Dropout) | (None, 128) | 0 |
| dense_4 (Dense) | (None, 64) | 8,256 |
| dropout_3 (Dropout) | (None, 64) | 0 |
| dense_5 (Dense) | (None, 2) | 130 |

Total params: 14,788,738 (56.41 MB)
Trainable params: 74,050 (289.26 KB)
Non-trainable params: 14,714,688 (56.13 MB)



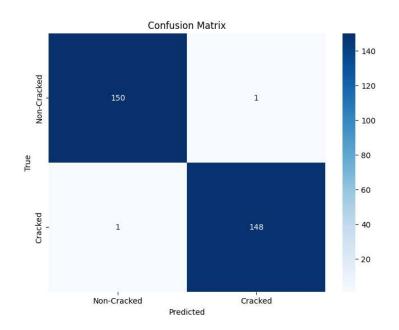
Hyper Parameters

- Learning Rate: Start at 0.0001 and use scheduling to decrease it over time for smoother convergence.
- **Grid Search**: Explore different combinations of learning rate, dropout rate, and batch size to find the best configuration.
- Batch Size: A batch size of 32 will be used to balance training time and memory usage.
- Optimizer: Apply Adam for adaptive learning rates, ensuring faster and more stable convergence.
- Dropout Rate: Apply a dropout rate of 0.5 to prevent overfitting during training

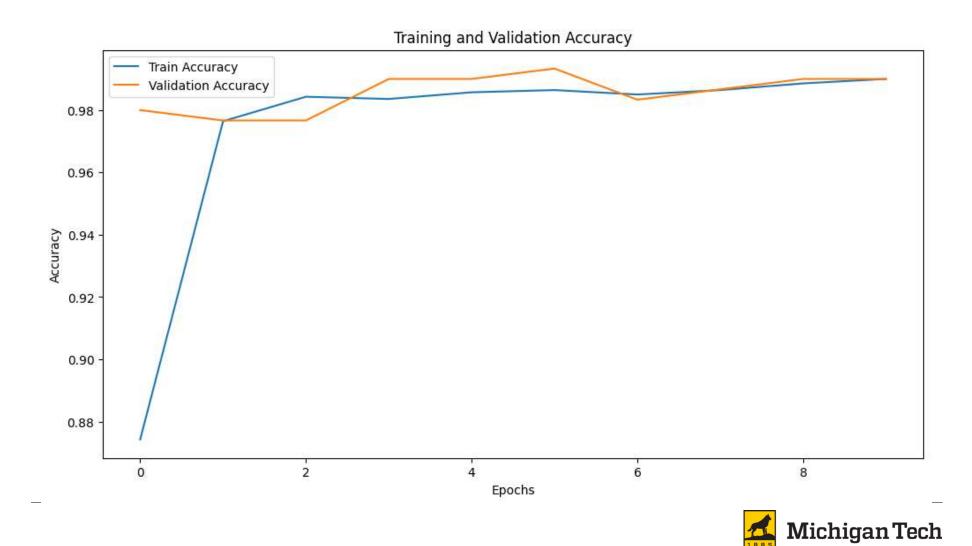


Metrics

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| Non-Cracked | 0.99 | 0.99 | 0.99 | 151 |
| Cracked | 0.99 | 0.99 | 0.99 | 149 |
| accuracy | | | 0.99 | 300 |
| macro avg | 0.99 | 0.99 | 0.99 | 300 |
| weighted avg | 0.99 | 0.99 | 0.99 | 300 |

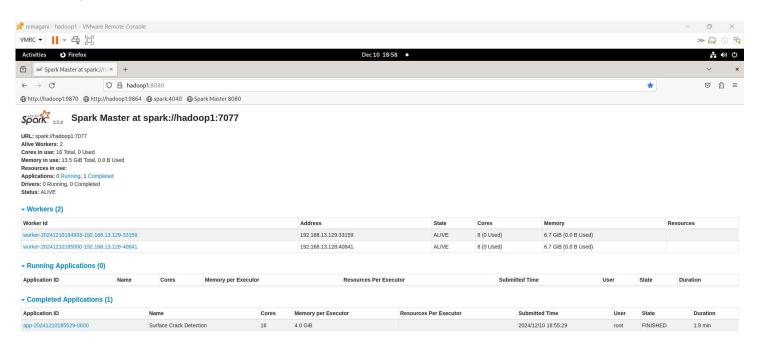






VGG 16

Computational Time





CNN

Input Layer (64x64x3)

First Conv2D (32 filters, 3x3, ReLU, Same)

Batch Normalization

MaxPooling2D (2x2, Stride 2)

Second Conv2D (64 filters, 3x3, ReLU, Same)

Batch Normalization

Flatten Layer

| Layer (type) | Output Shape | Param # |
|-----------------------------------------------|--------------------|---------|
| input_layer (InputLayer) | (None, 64, 64, 3) | 0 |
| conv2d (Conv2D) | (None, 64, 64, 32) | 896 |
| batch_normalization (BatchNormalization) | (None, 64, 64, 32) | 128 |
| max_pooling2d (MaxPooling2D) | (None, 32, 32, 32) | 0 |
| conv2d_1 (Conv2D) | (None, 32, 32, 64) | 18,496 |
| batch_normalization_1 (BatchNormalization) | (None, 32, 32, 64) | 256 |

| conv2d_1 (Conv2D) | (None, 32, 32, 64) | 18,496 |
|--------------------------------------------|--------------------|-----------|
| batch_normalization_1 (BatchNormalization) | (None, 32, 32, 64) | 256 |
| max_pooling2d_1 (MaxPooling2D) | (None, 16, 16, 64) | 0 |
| flatten (Flatten) | (None, 16384) | 0 |
| dense (Dense) | (None, 128) | 2,097,280 |
| dropout (Dropout) | (None, 128) | 0 |
| dense_1 (Dense) | (None, 64) | 8,256 |
| dropout_1 (Dropout) | (None, 64) | 0 |
| dense_2 (Dense) | (None, 2) | 130 |

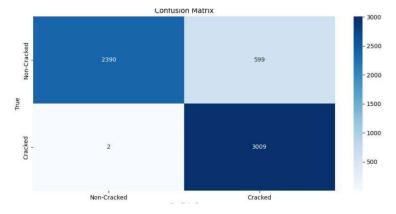
Total params: 2,125,442 (8.11 MB)
Trainable params: 2,125,250 (8.11 MB)
Non-trainable params: 12 (768.08 MB)
Usr/local/lib/python3.11/site-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: User
arning: Your "PyDataset: class should call 'super().__init__(*-kwargs)' in its constructor.' *-kwargs'
can include 'workers', 'use_multiprocessing', 'max_queue_size'. Do not pass these arguments to 'fit()',
s they will be ignored.
self._warn_if_super_not_called()

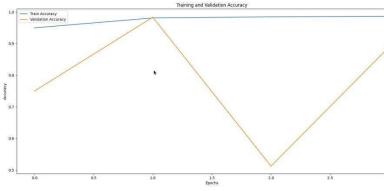
Hyperparameters

- Learning Rate: Default value of the Adam optimizer: **0.001** (not explicitly set in the code).
- Batch Size: 16 (used in datagen.flow for training).
- **Epochs**: 4 (number of complete passes through the dataset).
- **Dropout Rate**: **0.5** (used after each dense layer to reduce overfitting).
- This rate specifies the fraction of neurons to drop during training.
- Optimizer: Adam (used for optimization).

Model Performance:

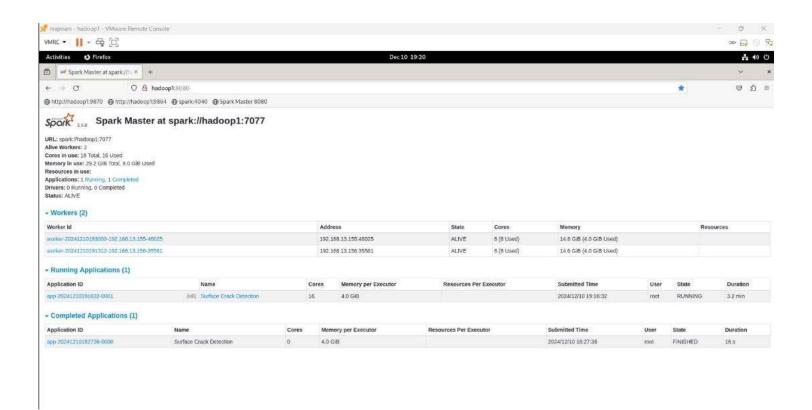
- The model achieved following results:
- Accuracy: 0.98
- Test accuracy:0.98
- Precision:0.96
- Recall:1.00
- F1 Score:0.98

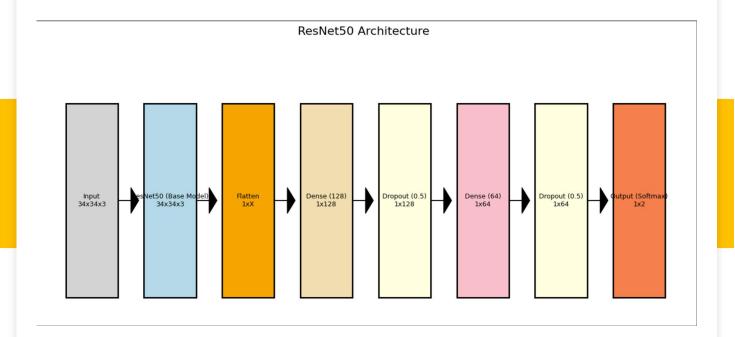




| lassificatio | precision | recall | f1-score | sup |
|--------------|-----------|--------|----------|-----|
| Non-Cracked | 1.00 | 0.23 | 0.38 | |
| Cracked | 0.57 | 1.00 | 0.72 | |
| accuracy | | | 0.62 | |
| macro avg | 0.78 | 0.62 | 0.55 | |
| eighted avg | 0.78 | 0.62 | 0.55 | |

Computational Time





RestNet50

| odel: "functional" | | | | |
|-------------------------------------------|--------------------|---------|-----------------------|--|
| Layer (type) | Output Shape | Param # | Connected to | |
| input_layer (InputLayer) | (None, 34, 34, 3) | 0 | - | |
| conv1_pad (ZeroPadding2D) | (None, 40, 40, 3) | 0 | input_layer[0][0] | |
| conv1_conv (Conv2D) | (None, 17, 17, 64) | 9,472 | conv1_pad[0][0] | |
| conv1_bn (BatchNormalization) | (None, 17, 17, 64) | 256 | conv1_conv[⊘][∅] | |
| conv1_relu (Activation) | (None, 17, 17, 64) | 0 | conv1_bn[0][0] | |
| pool1_pad (ZeroPadding2D) | (None, 19, 19, 64) | 0 | conv1_relu[0][0] | |
| pool1_pool (MaxPooling2D) | (None, 9, 9, 64) | 0 | pool1_pad[0][0] | |
| conv2_block1_1_conv (Conv2D) | (None, 9, 9, 64) | 4,160 | pool1_pool[0][0] | |
| conv2_block1_1_bn (BatchNormalization) | (None, 9, 9, 64) | 256 | conv2_block1_1_conv[8 | |
| conv2_block1_1_relu (Activation) | (None, 9, 9, 64) | 0 | conv2_block1_1_bn[0][| |
| conv2_block1_2_conv | (None, 9, 9, 64) | 36,928 | conv2_block1_1_relu[0 | |

| conv5_block3_2_relu (Activation) | (None, 2, 2, 512) | 0 | conv5_block3_2_bn[0][|
|-------------------------------------------|--------------------|-----------|------------------------------------------------|
| conv5_block3_3_conv (Conv2D) | (None, 2, 2, 2048) | 1,050,624 | conv5_block3_2_relu[0 |
| conv5_block3_3_bn (BatchNormalization) | (None, 2, 2, 2048) | 8,192 | conv5_block3_3_conv[0 |
| conv5_block3_add (Add) | (None, 2, 2, 2048) | Ø | conv5_block2_out[0][0 conv5_block3_3_bn[0][|
| conv5_block3_out (Activation) | (None, 2, 2, 2048) | 0 | conv5_block3_add[0][0] |
| flatten (Flatten) | (None, 8192) | Ø | conv5_block3_out[0][0] |
| dense (Dense) | (None, 128) | 1,048,704 | flatten[0][0] |
| dropout (Dropout) | (None, 128) | Ø | dense[0][0] |
| dense_1 (Dense) | (None, 64) | 8,256 | dropout[0][0] |
| dropout_1 (Dropout) | (None, 64) | Ø | dense_1[0][0] |
| dense_2 (Dense) | (None, 2) | 130 | dropout_1[0][0] |
| | | | |

Total params: 24,644,802 (94.01 MB)
Trainable params: 1,057,090 (4.03 MB)
Non-trainable params: 23,587,712 (89.98 MB)



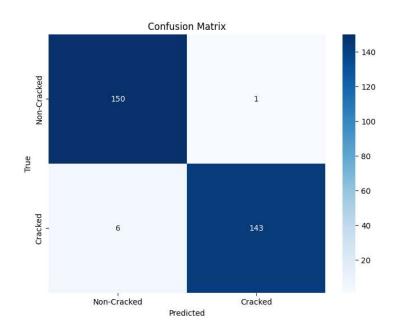
Hyper Parameters

- Learning Rate: Default value of the Adam optimizer:
 0.001
- Batch Size: 16 (a balance between training speed and stability).
- Epochs: 10 (To balance the computational time).
- Dropout Rate: 0.5 (used after each dense layer to reduce overfitting).
- Optimizer: Adam (used for optimization).

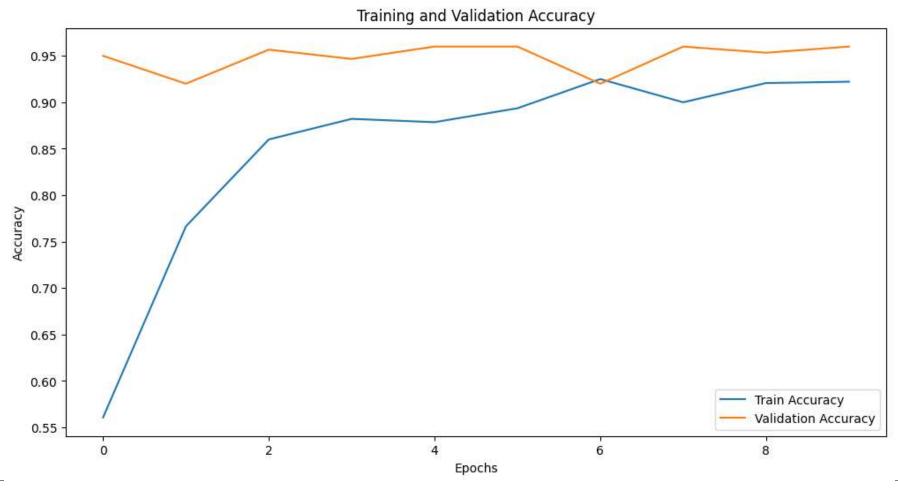


Metrics

| Classificatio | precision | recall | f1-score | support |
|---------------|-----------|--------|----------|---------|
| Non-Cracked | 0.96 | 0.99 | 0.98 | 151 |
| Cracked | 0.99 | 0.96 | 0.98 | 149 |
| accuracy | | | 0.98 | 300 |
| macro avg | 0.98 | 0.98 | 0.98 | 300 |
| weighted avg | 0.98 | 0.98 | 0.98 | 300 |



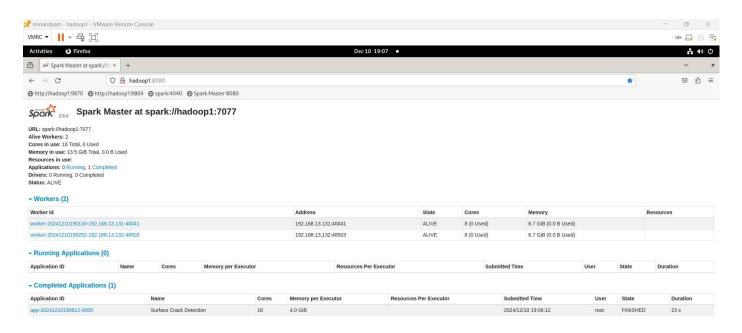






RestNet50

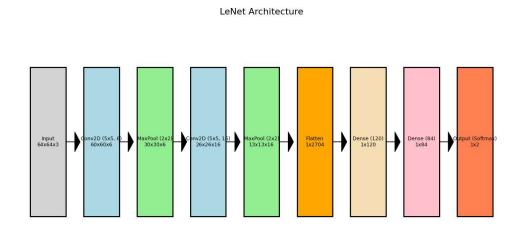
Computational Time





LeNet Model

| Layer (type) | Output Shape | Param # |
|--------------------------------|--------------------|---------|
| conv2d (Conv2D) | (None, 60, 60, 6) | 456 |
| max_pooling2d (MaxPooling2D) | (None, 30, 38, 5) | 0 |
| conv2d_1 (Conv20) | (None, 26, 26, 16) | 2,416 |
| max_pooling2d_1 (MaxPooling2D) | (None, 13, 13, 16) | 9 |
| flatten (Flatten) | (None, 2704) | |
| dense (Dense) | (None, 128) | 324,600 |
| dense_1 (Dense) | (None, S4) | 10,164 |
| dense_2 (Dense) | (None, ②) | 1.76 |



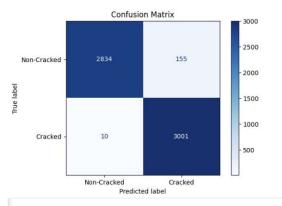
Hyper Parameters

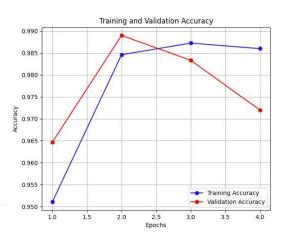
- **Batch Size:** A batch size of 64 is chosen, balancing training time and memory usage for effective learning.
- Learning Rate: Default learning rate of the Adam optimizer is used, allowing adaptive learning. Fine-tuning or scheduling could further improve convergence.
- **Epochs:** The model is trained for 4 epochs to balance computational time with performance, though this can be extended for improved results.
- Optimizer: The Adam optimizer is employed to leverage adaptive learning rates, ensuring faster and stable convergence during training.



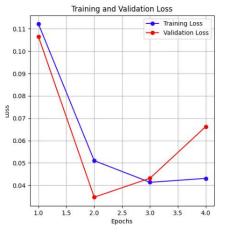
Model Performance:

- The model achieved following results:
- Accuracy: 0.96
- Test accuracy:0.97
- Precision:0.95
- Recall:0.97
- F1 Score:0.97

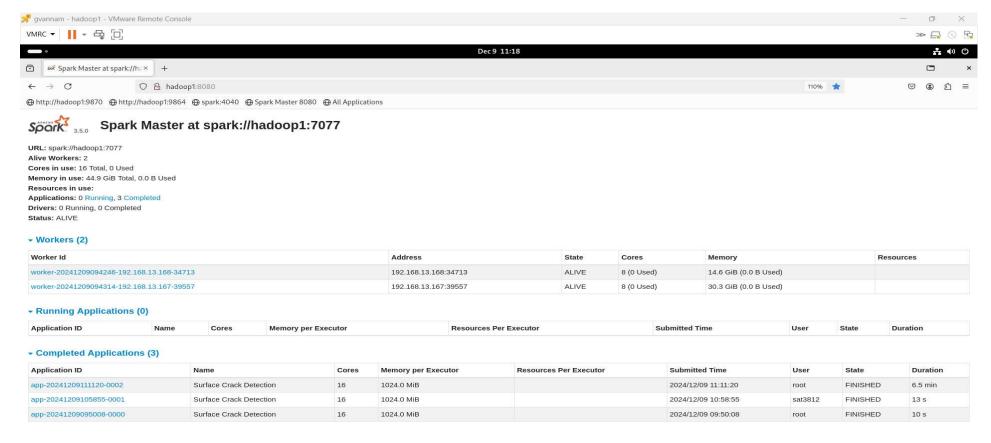




| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| Non-Cracked | 1.00 | 0.96 | 0.98 | 2989 |
| Cracked | 0.96 | 1.00 | 0.98 | 3011 |
| accuracy | | | 0.98 | 6000 |
| macro avg | 0.98 | 0.98 | 0.98 | 6000 |
| weighted avg | 0.98 | 0.98 | 0.98 | 6000 |



Computational Time



Best Model

- Best Model: VGG16
- Reason:
 - Fatal Errors: Only 1 cracked surface was misclassified as noncracked.
 - Metrics:
 - Recall for cracked surfaces is **0.99**, meaning it identified almost all cracks.
 - Precision is also 0.99, indicating high reliability.
 - Confusion Matrix: The misclassification rate is minimal, with just 1 error for cracked surfaces.
- **Recall for "Cracked"**: 0.95 (missed 7 cracked surfaces).
 - Precision for "Cracked": 0.95.
 - Overall accuracy: **0.97**.

