

# Proyek Akhir : Klasifikasi Gambar

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```
In [1]: # from google.colab import drive  
# drive.mount('/content/drive')
```

```
In [2]: # local_zip = '/content/drive/MyDrive/rockpaperscissors.zip'  
# zip_ref = zipfile.ZipFile(local_zip, 'r')  
# zip_ref.extractall('/content/drive/MyDrive/extractedData')  
# zip_ref.close()
```

```
In [3]: # base_dir = '/content/drive/MyDrive/extractedData/rockpaperscissors/rps-cv-im'
```

## Import semua library yang dibutuhkan

```
In [4]: import tensorflow as tf  
import numpy as np  
import matplotlib.pyplot as plt  
import matplotlib.image as mpimg  
import zipfile, os  
from google.colab import files  
from tensorflow.keras.preprocessing import image  
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

## Import dataset

```
In [5]: !wget --no-check-certificate \  
        https://github.com/dicodingacademy/assets/releases/download/release/rockpape  
        -O /tmp/rockpaperscissors.zip #import dari fungsi wget agar dataset tidak ad
```

```
--2023-12-05 02:13:26-- https://github.com/dicodingacademy/assets/releases/download/release/rockpaperscissors.zip (https://github.com/dicodingacademy/assets/releases/download/release/rockpaperscissors.zip)
Resolving github.com (github.com)... 140.82.113.3
Connecting to github.com (github.com)|140.82.113.3|:443... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://objects.githubusercontent.com/github-production-release-asset-2e65be/391417272/7eb836f2-695b-4a46-9c78-b65867166957?X-Amz-Algorithm=AWS4-HMAC-SHA256&X-Amz-Credential=AKIAIWNJYAX4CSVEH53A%2F20231205%2Fus-east-1%2Fus3%2Faws4_request&X-Amz-Date=20231205T021326Z&X-Amz-Expires=300&X-Amz-Signature=10c2a7b1865c4ce1b1f6f887d718a8acef07e32d0e223e1f2702b7e424c4c5ad&X-Amz-SignedHeaders=host&actor_id=0&key_id=0&repo_id=391417272&response-content-disposition=attachment%3B%20filename%3Drockpaperscissors.zip&response-content-type=application%2Foctet-stream (https://objects.githubusercontent.com/github-production-release-asset-2e65be/391417272/7eb836f2-695b-4a46-9c78-b65867166957?X-Amz-Algorithm=AWS4-HMAC-SHA256&X-Amz-Credential=AKIAIWNJYAX4CSVEH53A%2F20231205%2Fus-east-1%2Fus3%2Faws4_request&X-Amz-Date=20231205T021326Z&X-Amz-Expires=300&X-Amz-Signature=10c2a7b1865c4ce1b1f6f887d718a8acef07e32d0e223e1f2702b7e424c4c5ad&X-Amz-SignedHeaders=host&actor_id=0&key_id=0&repo_id=391417272&response-content-disposition=attachment%3B%20filename%3Drockpaperscissors.zip&response-content-type=application%2Foctet-stream) [following]
--2023-12-05 02:13:26-- https://objects.githubusercontent.com/github-production-release-asset-2e65be/391417272/7eb836f2-695b-4a46-9c78-b65867166957?X-Amz-Algorithm=AWS4-HMAC-SHA256&X-Amz-Credential=AKIAIWNJYAX4CSVEH53A%2F20231205%2Fus-east-1%2Fus3%2Faws4_request&X-Amz-Date=20231205T021326Z&X-Amz-Expires=300&X-Amz-Signature=10c2a7b1865c4ce1b1f6f887d718a8acef07e32d0e223e1f2702b7e424c4c5ad&X-Amz-SignedHeaders=host&actor_id=0&key_id=0&repo_id=391417272&response-content-disposition=attachment%3B%20filename%3Drockpaperscissors.zip&response-content-type=application%2Foctet-stream (https://objects.githubusercontent.com/github-production-release-asset-2e65be/391417272/7eb836f2-695b-4a46-9c78-b65867166957?X-Amz-Algorithm=AWS4-HMAC-SHA256&X-Amz-Credential=AKIAIWNJYAX4CSVEH53A%2F20231205%2Fus-east-1%2Fus3%2Faws4_request&X-Amz-Date=20231205T021326Z&X-Amz-Expires=300&X-Amz-Signature=10c2a7b1865c4ce1b1f6f887d718a8acef07e32d0e223e1f2702b7e424c4c5ad&X-Amz-SignedHeaders=host&actor_id=0&key_id=0&repo_id=391417272&response-content-disposition=attachment%3B%20filename%3Drockpaperscissors.zip&response-content-type=application%2Foctet-stream)
Resolving objects.githubusercontent.com (objects.githubusercontent.com)... 185.199.108.133, 185.199.109.133, 185.199.110.133, ...
Connecting to objects.githubusercontent.com (objects.githubusercontent.com)|185.199.108.133|:443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 322873683 (308M) [application/octet-stream]
Saving to: '/tmp/rockpaperscissors.zip'

/tmp/rockpapersciss 100%[=====>] 307.92M  185MB/s  in 1.7s

2023-12-05 02:13:28 (185 MB/s) - '/tmp/rockpaperscissors.zip' saved [322873683/322873683]
```

## Unzip dan Assign dataset base directory

```
In [6]: local_zip = '/tmp/rockpaperscissors.zip' #select zipped file
zip_ref = zipfile.ZipFile(local_zip, 'r') #read zipped file
zip_ref.extractall('/tmp') #extract to location
zip_ref.close() #close zipped file

base_dir = '/tmp/rockpaperscissors/rps-cv-images' #directory yang memuat file
```

```
In [7]: os.listdir(base_dir)
```

```
Out[7]: ['scissors', 'paper', 'README_rpc-cv-images.txt', 'rock']
```

## Image Data Generator

```
In [8]: train_datagen = ImageDataGenerator(
        rescale=1./255, # rescaling factor. If None or 0, no resca
        rotation_range=20, # Int. Degree range for random rotation
        horizontal_flip=True, # Boolean. Randomly flip inputs hori
        vertical_flip=True, # Boolean. Randomly flip inputs vertic
        shear_range = 0.2, # Float. Shear Intensity (Shear angle i
        fill_mode = 'nearest', # Points outside the boundaries of
        validation_split=0.4,) # split data menjadi 40% validation

# test_datagen = ImageDataGenerator( #no actual need for this i guess because
#     rescale=1./255,
#     validation_split=0.4,)

# source for ImageDataGenerator: https://www.tensorflow.org/api_docs/python/tf,
```

## Pembagian Dataset Train dan Test

```
In [9]: train_generator = train_datagen.flow_from_directory(
        base_dir, # folder/size yang digunakan
        target_size=(200, 300), # ubah resolusi gambar menjadi 200 x 300 (dic
        batch_size=4,
        class_mode='categorical', #categorical karena pilihan output lebih dar
        subset="training") # karena di split dari train_datagen menggunakan va

validation_generator = train_datagen.flow_from_directory(
        base_dir,
        target_size=(200, 300),
        batch_size=4,
        class_mode='categorical',
        subset="validation") # karena di split dari train_datagen menggunakan

# source for flow_from_directory: https://www.tensorflow.org/api_docs/python/tf
```

Found 1314 images belonging to 3 classes.  
Found 874 images belonging to 3 classes.

## Sequential Model

```
In [10]: # ini dari contoh
model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(32, (3,3), activation='relu', input_shape=(200, 300
    tf.keras.layers.MaxPooling2D(2, 2),
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Conv2D(128, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Conv2D(256, (3,3), activation='relu'), #254 was
    # tf.keras.layers.MaxPooling2D(2,2),
    # tf.keras.layers.Conv2D(512, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Flatten(),
    # tf.keras.layers.Dense(256, activation='relu'),
    tf.keras.layers.Dense(512, activation='relu'),
    tf.keras.layers.Dense(3, activation='softmax') #diganti 1->3 , sigmoid->so
])

# contoh penggunaan dan penjelasan dari softmax https://youtu.be/7HPwo4wnJeA?t
```

```
In [11]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 198, 298, 32)	896
max_pooling2d (MaxPooling2D)	(None, 99, 149, 32)	0
conv2d_1 (Conv2D)	(None, 97, 147, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 48, 73, 64)	0
conv2d_2 (Conv2D)	(None, 46, 71, 128)	73856
max_pooling2d_2 (MaxPooling2D)	(None, 23, 35, 128)	0
conv2d_3 (Conv2D)	(None, 21, 33, 256)	295168
max_pooling2d_3 (MaxPooling2D)	(None, 10, 16, 256)	0
flatten (Flatten)	(None, 40960)	0
dense (Dense)	(None, 512)	20972032
dense_1 (Dense)	(None, 3)	1539
=====		
Total params: 21361987 (81.49 MB)		
Trainable params: 21361987 (81.49 MB)		
Non-trainable params: 0 (0.00 Byte)		
=====		

## Model Compile and Fit

```
In [12]: model.compile(loss='categorical_crossentropy', #menggunakan 'categorical_crossentropy',
                      optimizer=tf.optimizers.Adam(),
                      metrics=['accuracy'])
```

```
In [13]: history = model.fit( #di set ke variabel 'history' untuk plotting data epochs  
    train_generator, # train data  
    steps_per_epoch=25, # Total number of steps (batches of samples) before  
    epochs=30, # 15 epochs baru bisa membedakan gunting dari lainnya dengan  
    validation_data=validation_generator, # validation data  
    validation_steps=5, # Total number of steps (batches of samples) to draw  
    verbose=2) #verbosity mode for information per epoch  
  
# source for .fit: https://www.tensorflow.org/api_docs/python/tf/keras/Model#f
```

Epoch 1/30  
25/25 - 46s - loss: 1.2809 - accuracy: 0.3900 - val\_loss: 1.0853 - val\_accuracy: 0.5500 - 46s/epoch - 2s/step

Epoch 2/30  
25/25 - 31s - loss: 1.0926 - accuracy: 0.3000 - val\_loss: 1.0496 - val\_accuracy: 0.4000 - 31s/epoch - 1s/step

Epoch 3/30  
25/25 - 30s - loss: 1.0790 - accuracy: 0.4200 - val\_loss: 1.0446 - val\_accuracy: 0.7500 - 30s/epoch - 1s/step

Epoch 4/30  
25/25 - 30s - loss: 1.0157 - accuracy: 0.5800 - val\_loss: 0.8356 - val\_accuracy: 0.4500 - 30s/epoch - 1s/step

Epoch 5/30  
25/25 - 29s - loss: 0.8262 - accuracy: 0.6600 - val\_loss: 0.7247 - val\_accuracy: 0.6000 - 29s/epoch - 1s/step

Epoch 6/30  
25/25 - 29s - loss: 0.7116 - accuracy: 0.6800 - val\_loss: 0.8921 - val\_accuracy: 0.6000 - 29s/epoch - 1s/step

Epoch 7/30  
25/25 - 29s - loss: 0.7524 - accuracy: 0.6800 - val\_loss: 1.3949 - val\_accuracy: 0.4500 - 29s/epoch - 1s/step

Epoch 8/30  
25/25 - 29s - loss: 0.7882 - accuracy: 0.6900 - val\_loss: 1.0005 - val\_accuracy: 0.4500 - 29s/epoch - 1s/step

Epoch 9/30  
25/25 - 29s - loss: 0.5827 - accuracy: 0.7800 - val\_loss: 0.5351 - val\_accuracy: 0.8000 - 29s/epoch - 1s/step

Epoch 10/30  
25/25 - 29s - loss: 0.5775 - accuracy: 0.7800 - val\_loss: 0.3099 - val\_accuracy: 0.9000 - 29s/epoch - 1s/step

Epoch 11/30  
25/25 - 28s - loss: 0.4315 - accuracy: 0.8300 - val\_loss: 0.2978 - val\_accuracy: 0.9000 - 28s/epoch - 1s/step

Epoch 12/30  
25/25 - 28s - loss: 0.4583 - accuracy: 0.8200 - val\_loss: 0.6025 - val\_accuracy: 0.7500 - 28s/epoch - 1s/step

Epoch 13/30  
25/25 - 30s - loss: 0.4389 - accuracy: 0.8400 - val\_loss: 0.4192 - val\_accuracy: 0.8000 - 30s/epoch - 1s/step

Epoch 14/30  
25/25 - 29s - loss: 0.3556 - accuracy: 0.9100 - val\_loss: 0.4957 - val\_accuracy: 0.8000 - 29s/epoch - 1s/step

Epoch 15/30  
25/25 - 28s - loss: 0.2538 - accuracy: 0.9200 - val\_loss: 0.5475 - val\_accuracy: 0.6500 - 28s/epoch - 1s/step

Epoch 16/30  
25/25 - 29s - loss: 0.4068 - accuracy: 0.8200 - val\_loss: 0.2209 - val\_accuracy: 0.9500 - 29s/epoch - 1s/step

Epoch 17/30  
25/25 - 28s - loss: 0.3132 - accuracy: 0.9000 - val\_loss: 0.1206 - val\_accuracy: 0.9500 - 28s/epoch - 1s/step

Epoch 18/30  
25/25 - 36s - loss: 0.4131 - accuracy: 0.8500 - val\_loss: 0.2887 - val\_accuracy: 0.9000 - 36s/epoch - 1s/step

Epoch 19/30  
25/25 - 29s - loss: 0.3400 - accuracy: 0.8900 - val\_loss: 0.1998 - val\_accuracy: 0.9500 - 29s/epoch - 1s/step



Epoch 20/30  
25/25 - 29s - loss: 0.3189 - accuracy: 0.8400 - val\_loss: 0.3492 - val\_accuracy: 0.8500 - 29s/epoch - 1s/step

Epoch 21/30  
25/25 - 29s - loss: 0.2756 - accuracy: 0.9000 - val\_loss: 0.1386 - val\_accuracy: 0.9500 - 29s/epoch - 1s/step

Epoch 22/30  
25/25 - 31s - loss: 0.3967 - accuracy: 0.8800 - val\_loss: 0.2107 - val\_accuracy: 0.9000 - 31s/epoch - 1s/step

Epoch 23/30  
25/25 - 29s - loss: 0.3455 - accuracy: 0.8800 - val\_loss: 0.2707 - val\_accuracy: 0.9500 - 29s/epoch - 1s/step

Epoch 24/30  
25/25 - 29s - loss: 0.2540 - accuracy: 0.9000 - val\_loss: 0.0947 - val\_accuracy: 1.0000 - 29s/epoch - 1s/step

Epoch 25/30  
25/25 - 28s - loss: 0.1981 - accuracy: 0.9592 - val\_loss: 0.1381 - val\_accuracy: 0.9000 - 28s/epoch - 1s/step

Epoch 26/30  
25/25 - 29s - loss: 0.2831 - accuracy: 0.9100 - val\_loss: 0.3818 - val\_accuracy: 0.9000 - 29s/epoch - 1s/step

Epoch 27/30  
25/25 - 29s - loss: 0.4136 - accuracy: 0.8400 - val\_loss: 0.2885 - val\_accuracy: 0.9500 - 29s/epoch - 1s/step

Epoch 28/30  
25/25 - 28s - loss: 0.2548 - accuracy: 0.9000 - val\_loss: 0.2019 - val\_accuracy: 0.9000 - 28s/epoch - 1s/step

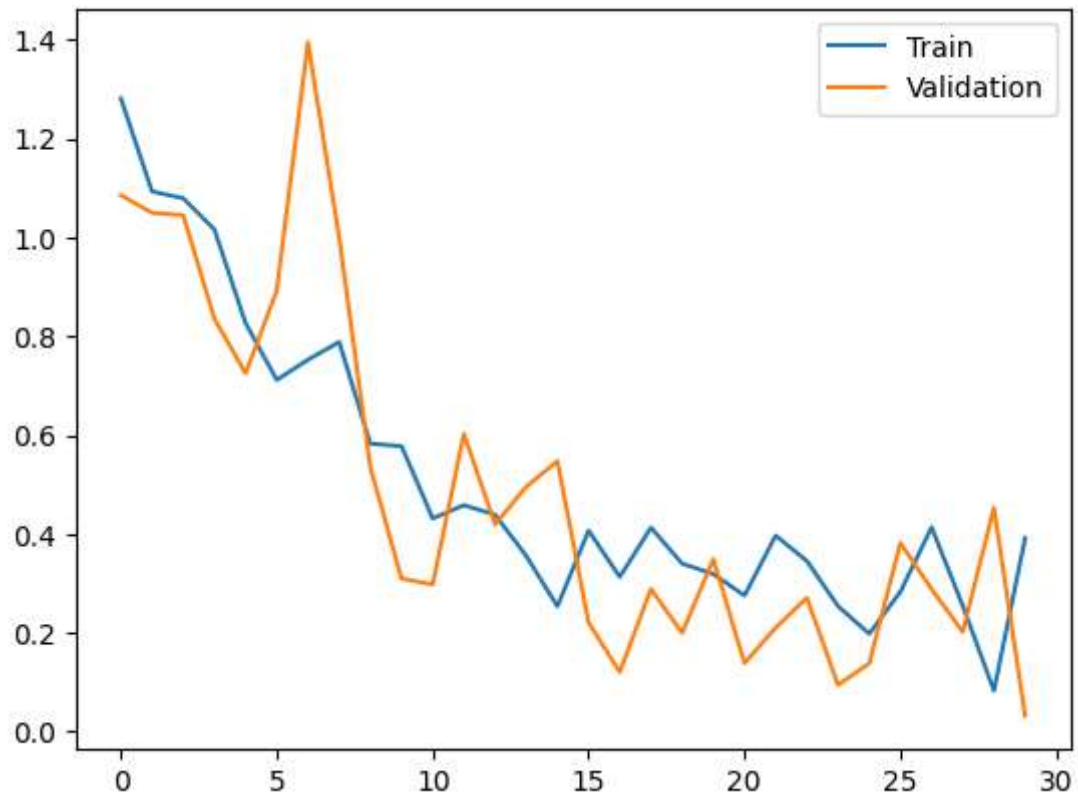
Epoch 29/30  
25/25 - 28s - loss: 0.0832 - accuracy: 0.9700 - val\_loss: 0.4528 - val\_accuracy: 0.8000 - 28s/epoch - 1s/step

Epoch 30/30  
25/25 - 29s - loss: 0.3915 - accuracy: 0.8776 - val\_loss: 0.0325 - val\_accuracy: 1.0000 - 29s/epoch - 1s/step

## Model Epoch Plotting

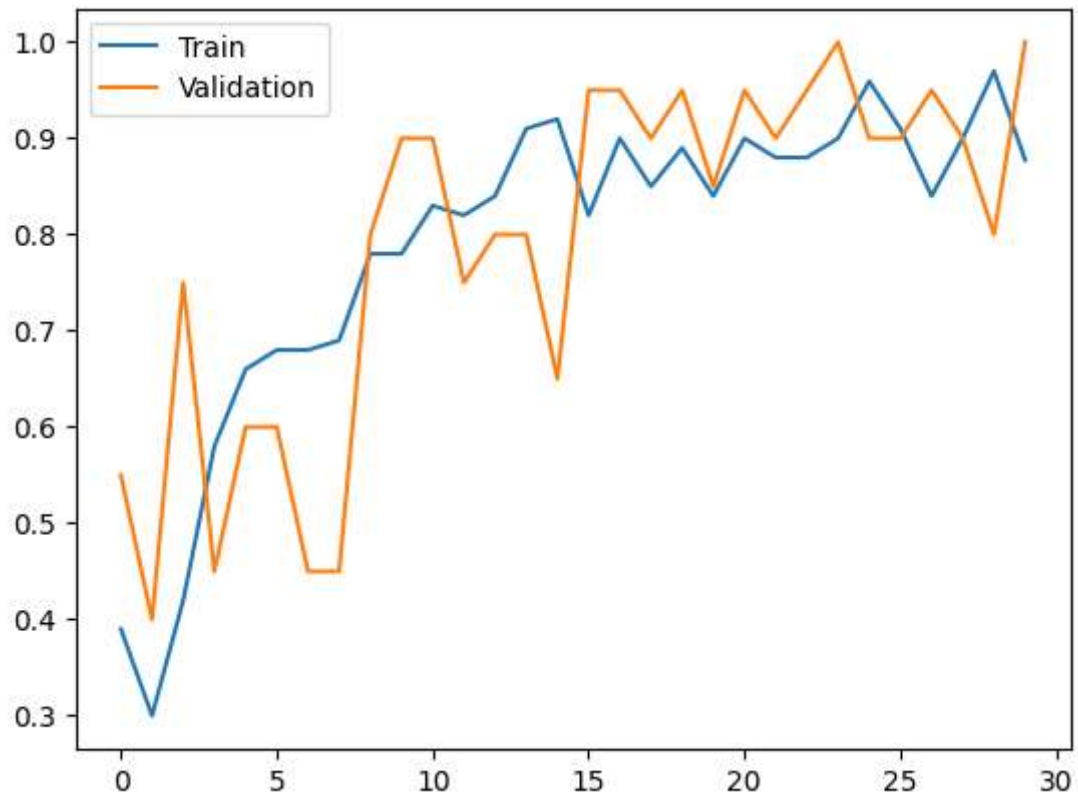
```
In [14]: plt.plot(history.history['loss'],label="Train")  
plt.plot(history.history['val_loss'],label="Validation")  
plt.legend()
```

Out[14]: <matplotlib.legend.Legend at 0x7b1a7c2deda0>



```
In [15]: plt.plot(history.history['accuracy'],label="Train")  
plt.plot(history.history['val_accuracy'],label="Validation")  
plt.legend()
```

Out[15]: <matplotlib.legend.Legend at 0x7b1a6ff26860>



## Predicting image from user input

```
In [16]: %matplotlib inline

uploaded = files.upload()

for fn in uploaded.keys():

    path = fn
    img = image.load_img(path, target_size=(200,300))

    imgplot = plt.imshow(img)
    x = image.img_to_array(img)
    x = np.expand_dims(x, axis=0)

    images = np.vstack([x])
    classes = model.predict(images, batch_size=10)
    classed = np.argmax(classes)

    print(classes)
    print(classed)
    print("Object is:")

    if classes[0,0] == 1:
        print('paper')
    elif classes[0,1] == 1:
        print('rock')
    elif classes[0,2] == 1:
        print('scissors')
    else:
        print('Unknown...')
```

No file chosen

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving paper.png to paper.png

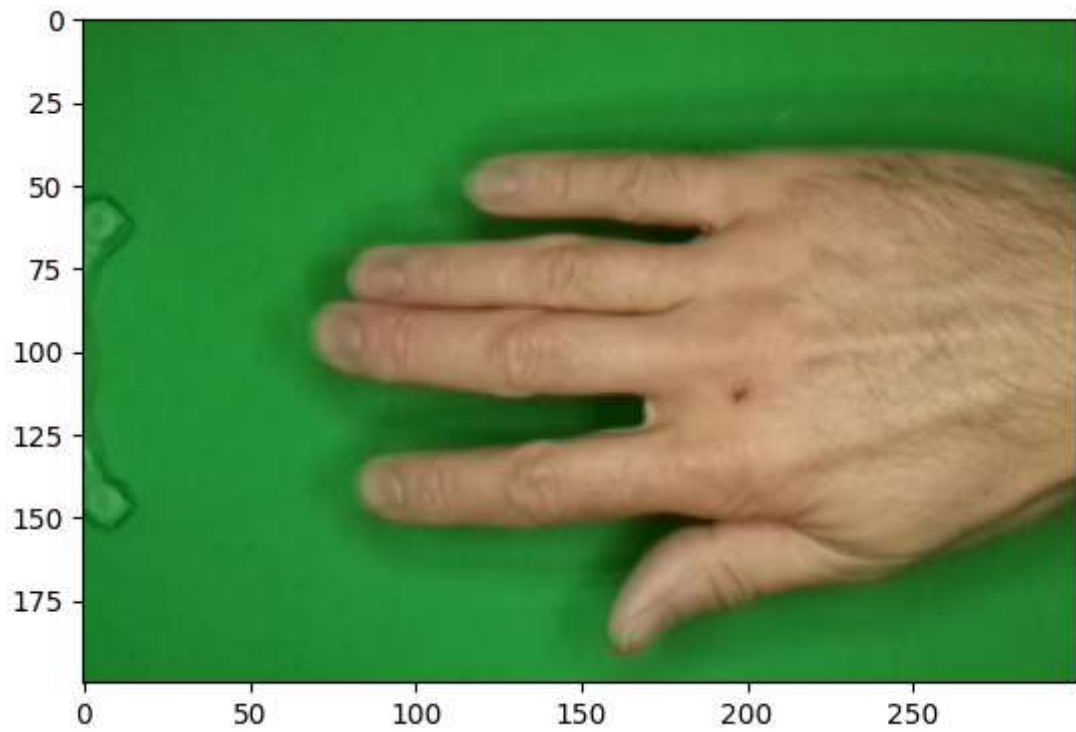
1/1 [=====] - 0s 238ms/step

[[1. 0. 0.]]

0

Object is:

paper



kekurangan dari dataset ini adalah model yang dibuat menggunakan dataset ini hanya bisa memprediksi bentuk tangan dengan baik jika backgroundnya seperti yang di dataset (hijau, dan tidak banyak objek dibelakangnya)