

TUGAS BESAR KULIAH SISTEM PENGUKURAN BERBASIS CITRA

Nama : Ronggur Mahendra Widya Putra

NIM : 13519008

In [48]:

```
# Import Library
import tensorflow as tf

from tensorflow.keras import datasets, layers, models
import tensorflow_datasets as tfds
import matplotlib.pyplot as plt
import cv2
import os
import numpy as np
import scipy
from skimage import color, data, restoration
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from random import uniform
from tensorflow.keras.layers import BatchNormalization
```

Loading Data dari folder ./Data/ClassifiedData

In [21]:

```

ClassifiedDataDir = "./Data/ClassifiedData/Train"
BATCH_SIZE = 32
IMG_SIZE = (128, 128)
train_dataset = tf.keras.utils.image_dataset_from_directory(
    directory = ClassifiedDataDir,
    labels='inferred',
    label_mode='int',
    class_names=None,
    color_mode='rgb',
    batch_size=BATCH_SIZE,
    image_size=IMG_SIZE,
    shuffle=True,
    seed=1337,
    subset="training",
    validation_split=0.1,
    interpolation='bilinear',
    follow_links=False,
    crop_to_aspect_ratio=False,
    # rescale = 1./255,
)
validation_dataset = tf.keras.utils.image_dataset_from_directory(
    directory = ClassifiedDataDir,
    labels='inferred',
    label_mode='int',
    class_names=None,
    color_mode='rgb',
    batch_size=BATCH_SIZE,
    image_size=IMG_SIZE,
    shuffle=True,
    seed=1337,
    subset="validation",
    validation_split=0.1,
    interpolation='bilinear',
    follow_links=False,
    crop_to_aspect_ratio=False,
    # rescale = 1./255,
)

test_dataset = tf.keras.utils.image_dataset_from_directory(
    directory = "./Data/ClassifiedData/Test",
    labels='inferred',
    label_mode='int',
    class_names=None,
    color_mode='rgb',
    batch_size=BATCH_SIZE,
    image_size=IMG_SIZE,
    shuffle=True,
    seed=None,
    interpolation='bilinear',
    follow_links=False,
    crop_to_aspect_ratio=False,
    # rescale = 1./255,
)

```

Found 3525 files belonging to 24 classes.
 Using 3173 files for training.

Found 3525 files belonging to 24 classes.
Using 352 files for validation.
Found 312 files belonging to 24 classes.

List Class

In [22]:

```
# for element in dataset.as_numpy_iterator():  
#     print(element)  
class_names = train_dataset.class_names  
print(class_names)  
print(len(class_names))
```

```
['A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'K', 'L', 'M', 'N', 'O', 'P',  
'Q', 'R', 'S', 'T', 'U', 'V', 'W', 'X', 'Y']  
24
```

In [23]:

```
# print('Number of training batches: %d' % tf.data.experimental.cardinality(train_dataset).  
# print('Number of validation batches: %d' % tf.data.experimental.cardinality(validation_da
```

Optimization buffer untuk Dataset

In [24]:

```
# DataSet  
AUTOTUNE = tf.data.AUTOTUNE  
train_dataset = train_dataset.prefetch(buffer_size=AUTOTUNE)  
validation_dataset = validation_dataset.prefetch(buffer_size=AUTOTUNE)  
test_dataset = test_dataset.prefetch(buffer_size=AUTOTUNE)
```

Model 1

In [47]:

```

# # # Model 1 Initialize
# input_size = 128
# filter_size = 14
# num_filter = 8
# maxpool_size = 2
# batch_size = BATCH_SIZE
# epochs = 30

# model1 = tf.keras.models.Sequential([
#     tf.keras.layers.InputLayer(input_shape=(input_size, input_size, 3)),
#     tf.keras.layers.Reshape((256, 256 * 3)),
#     tf.keras.layers.Bidirectional(tf.keras.layers.LSTM( 256, return_sequences=True, return
#     tf.keras.layers.Bidirectional(tf.keras.layers.LSTM( 256 )),
#     tf.keras.layers.Dense( 256 ),
#     tf.keras.layers.Dropout(.2),
#     tf.keras.layers.Dense( 256 ),
# ])

# # model = models.Sequential()
# # model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(input_size, input_s
# # model.add(layers.MaxPooling2D((2, 2)))
# # model.add(layers.Conv2D(64, (3, 3), activation='relu'))
# # model.add(layers.MaxPooling2D((2, 2)))
# # model.add(layers.Conv2D(64, (3, 3), activation='relu'))

# model1.add(tf.keras.layers.Flatten())
# # model.add(tf.keras.layers.Dense(24))
# model1.add(tf.keras.layers.Dense(24, activation = 'softmax'))

# # model.add(layers.Flatten())
# # model.add(layers.Dense(64, activation='relu'))
# # model.add(layers.Dense(24))

# model1.summary()

```

In [26]:

```

# # Optimizer
# optimizer = tf.keras.optimizers.Nadam(
#     learning_rate=0.00001, beta_1=0.9, beta_2=0.999, epsilon=1e-07,
#     name='Nadam'
# ) # 0.00001

# # Loss Fn
# lossfn = tf.keras.losses.SparseCategoricalCrossentropy( from_logits=False, reduction=tf.k

# # Model Summary
# model.compile(
#     optimizer="rmsprop",
#     loss="sparse_categorical_crossentropy",
#     metrics=["sparse_categorical_accuracy"],
# )

```

Creating CNN Model

In [27]:

```
# Model 2
```

```
input_size = 128
filter_size = 3
num_filter = 8
maxpool_size = 2
batch_size = BATCH_SIZE
epochs = 30

steps_per_epoch = 24720/batch_size

model = tf.keras.models.Sequential()
model.add(tf.keras.layers.Conv2D(16, (filter_size,filter_size),
    input_shape= (input_size,input_size,3),
    activation = 'relu',
    padding='same'))
model.add(tf.keras.layers.Conv2D(16, (filter_size,filter_size),
    input_shape= (input_size,input_size,3),
    activation = 'relu',
    padding='same'))
model.add(tf.keras.layers.MaxPooling2D(pool_size=(maxpool_size, maxpool_size),strides=1))
model.add(tf.keras.layers.Dropout(uniform(0, 1)))

model.add(tf.keras.layers.Conv2D(32, (filter_size,filter_size),
    activation='relu',
    padding='valid'))
model.add(tf.keras.layers.Conv2D(32, (filter_size,filter_size),
    activation='relu',
    padding='valid'))
model.add(tf.keras.layers.BatchNormalization())
model.add(tf.keras.layers.MaxPooling2D(pool_size=(maxpool_size, maxpool_size),strides=2))
model.add(tf.keras.layers.Dropout(uniform(0, 1)))

model.add(tf.keras.layers.Conv2D(32, (filter_size,filter_size),
    activation='relu',
    padding='valid'))
model.add(tf.keras.layers.Conv2D(32, (filter_size,filter_size),
    activation='relu',
    padding='valid'))
model.add(tf.keras.layers.BatchNormalization())
model.add(tf.keras.layers.MaxPooling2D(pool_size=(maxpool_size, maxpool_size),strides=2))
model.add(tf.keras.layers.Dropout(uniform(0, 1)))

model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Dense(120, activation='relu'))
model.add(tf.keras.layers.Dense(24,activation='softmax'))
```

In [28]:

```
METRICS = [ 'accuracy']#, 'precision','recall']
model.compile( optimizer= tf.keras.optimizers.Adam(lr=0.001),loss='sparse_categorical_crossentropy')
model.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d_6 (Conv2D)	(None, 128, 128, 16)	448
conv2d_7 (Conv2D)	(None, 128, 128, 16)	2320
max_pooling2d_3 (MaxPooling2D)	(None, 127, 127, 16)	0
dropout_3 (Dropout)	(None, 127, 127, 16)	0
conv2d_8 (Conv2D)	(None, 125, 125, 32)	4640
conv2d_9 (Conv2D)	(None, 123, 123, 32)	9248
batch_normalization_2 (Batch Normalization)	(None, 123, 123, 32)	128
max_pooling2d_4 (MaxPooling2D)	(None, 61, 61, 32)	0
dropout_4 (Dropout)	(None, 61, 61, 32)	0
conv2d_10 (Conv2D)	(None, 59, 59, 32)	9248
conv2d_11 (Conv2D)	(None, 57, 57, 32)	9248
batch_normalization_3 (Batch Normalization)	(None, 57, 57, 32)	128
max_pooling2d_5 (MaxPooling2D)	(None, 28, 28, 32)	0
dropout_5 (Dropout)	(None, 28, 28, 32)	0
flatten_1 (Flatten)	(None, 25088)	0
dense_2 (Dense)	(None, 120)	3010680
dense_3 (Dense)	(None, 24)	2904
Total params: 3,048,992		
Trainable params: 3,048,864		
Non-trainable params: 128		

Makingsure Training using GPU

In [29]:

```
print(tf.test.is_built_with_cuda())  
print(tf.config.list_physical_devices('GPU'))  
print("Num GPUs Available: ", len(tf.config.list_physical_devices('GPU')))
```

True

[PhysicalDevice(name='/physical_device:GPU:0', device_type='GPU')]

Num GPUs Available: 1

Training

In [30]:

```
# Training
history = model.fit(
    train_dataset,
    epochs=30,
    validation_data = (validation_dataset)
)
```

Epoch 1/30

100/100 [=====] - 8s 69ms/step - loss: 3.7018 - accuracy: 0.1210 - val_loss: 19.0571 - val_accuracy: 0.0511

Epoch 2/30

100/100 [=====] - 7s 70ms/step - loss: 2.1337 - accuracy: 0.3325 - val_loss: 14.2802 - val_accuracy: 0.1477

Epoch 3/30

100/100 [=====] - 8s 72ms/step - loss: 1.4926 - accuracy: 0.5317 - val_loss: 2.2757 - val_accuracy: 0.4830

Epoch 4/30

100/100 [=====] - 8s 73ms/step - loss: 0.8629 - accuracy: 0.7182 - val_loss: 3.0339 - val_accuracy: 0.5227

Epoch 5/30

100/100 [=====] - 8s 73ms/step - loss: 0.5989 - accuracy: 0.7945 - val_loss: 12.5601 - val_accuracy: 0.1733

Epoch 6/30

100/100 [=====] - 8s 75ms/step - loss: 0.4056 - accuracy: 0.8544 - val_loss: 9.9583 - val_accuracy: 0.2670

Epoch 7/30

100/100 [=====] - 8s 77ms/step - loss: 0.3230 - accuracy: 0.8951 - val_loss: 8.1203 - val_accuracy: 0.3011

Epoch 8/30

100/100 [=====] - 8s 80ms/step - loss: 0.2675 - accuracy: 0.9171 - val_loss: 0.8034 - val_accuracy: 0.7500

Epoch 9/30

100/100 [=====] - 8s 78ms/step - loss: 0.2010 - accuracy: 0.9354 - val_loss: 0.7667 - val_accuracy: 0.7983

Epoch 10/30

100/100 [=====] - 8s 75ms/step - loss: 0.1977 - accuracy: 0.9392 - val_loss: 1.3063 - val_accuracy: 0.6705

Epoch 11/30

100/100 [=====] - 8s 75ms/step - loss: 0.2042 - accuracy: 0.9382 - val_loss: 1.2486 - val_accuracy: 0.7273

Epoch 12/30

100/100 [=====] - 8s 75ms/step - loss: 0.1601 - accuracy: 0.9499 - val_loss: 0.5626 - val_accuracy: 0.8324

Epoch 13/30

100/100 [=====] - 8s 75ms/step - loss: 0.1331 - accuracy: 0.9581 - val_loss: 0.9256 - val_accuracy: 0.7358

Epoch 14/30

100/100 [=====] - 8s 75ms/step - loss: 0.1202 - accuracy: 0.9634 - val_loss: 1.4013 - val_accuracy: 0.6420

Epoch 15/30

100/100 [=====] - 8s 76ms/step - loss: 0.1233 - accuracy: 0.9638 - val_loss: 0.7337 - val_accuracy: 0.8295

Epoch 16/30

100/100 [=====] - 8s 76ms/step - loss: 0.1327 - accuracy: 0.9587 - val_loss: 1.1404 - val_accuracy: 0.7784

Epoch 17/30

100/100 [=====] - 8s 77ms/step - loss: 0.1318 - accuracy: 0.9628 - val_loss: 0.8612 - val_accuracy: 0.8040

Epoch 18/30

100/100 [=====] - 8s 76ms/step - loss: 0.1076 - accuracy: 0.9660 - val_loss: 2.7944 - val_accuracy: 0.6449

Epoch 19/30

100/100 [=====] - 8s 76ms/step - loss: 0.0864 - accuracy: 0.9723 - val_loss: 2.8514 - val_accuracy: 0.6449

Epoch 20/30

100/100 [=====] - 8s 76ms/step - loss: 0.0746 - accuracy: 0.9729 - val_loss: 0.1848 - val_accuracy: 0.9574

Epoch 21/30

100/100 [=====] - 8s 78ms/step - loss: 0.1082 - accuracy: 0.9675 - val_loss: 4.1118 - val_accuracy: 0.4517

Epoch 22/30

100/100 [=====] - 8s 78ms/step - loss: 0.0945 - accuracy: 0.9697 - val_loss: 2.9749 - val_accuracy: 0.5597

Epoch 23/30

100/100 [=====] - 8s 75ms/step - loss: 0.0646 - accuracy: 0.9801 - val_loss: 6.3521 - val_accuracy: 0.3523

Epoch 24/30

100/100 [=====] - 8s 77ms/step - loss: 0.0713 - accuracy: 0.9748 - val_loss: 11.6890 - val_accuracy: 0.2358

Epoch 25/30

100/100 [=====] - 8s 81ms/step - loss: 0.0651 - accuracy: 0.9776 - val_loss: 6.2285 - val_accuracy: 0.3494

Epoch 26/30

100/100 [=====] - 8s 79ms/step - loss: 0.0851 - accuracy: 0.9760 - val_loss: 1.5400 - val_accuracy: 0.7585

Epoch 27/30

100/100 [=====] - 8s 79ms/step - loss: 0.0496 - accuracy: 0.9849 - val_loss: 1.5441 - val_accuracy: 0.7614

Epoch 28/30

100/100 [=====] - 8s 79ms/step - loss: 0.0703 - accuracy: 0.9798 - val_loss: 8.9021 - val_accuracy: 0.2841

Epoch 29/30

100/100 [=====] - 8s 80ms/step - loss: 0.0542 - accuracy: 0.9833 - val_loss: 7.8091 - val_accuracy: 0.3068

Epoch 30/30

100/100 [=====] - 8s 77ms/step - loss: 0.0571 - accuracy: 0.9817 - val_loss: 1.2181 - val_accuracy: 0.8011

In [31]:

```
loss, acc = model.evaluate(test_dataset)
print(loss, acc)
```

```
10/10 [=====] - 1s 15ms/step - loss: 0.6346 - accuracy: 0.8910
0.6345966458320618 0.8910256624221802
```

Saving model to load later

In [49]:

```
model.save('saved_model/FinalModel')
```

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op while saving (showing 5 of 6). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: saved_model/FinalModel/assets

INFO:tensorflow:Assets written to: saved_model/FinalModel/assets

Model's Performance

In [45]:

```
TrainingAccuracy = history.history['accuracy'][-1]
print("Train accuracy: {:.5.2f}%".format(100 * TrainingAccuracy))
print("Test accuracy: {:.5.2f}%".format(100 * acc))
```

Train accuracy: 98.17%

Test accuracy: 89.10%

Training History

In [35]:

```
plt.plot(history.history['accuracy'], label='accuracy')
plt.plot(history.history['val_accuracy'], label = 'val_accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.ylim([0.5, 1])
plt.legend(loc='lower right')

# test_loss, test_acc = model.evaluate(test_images, test_labels, verbose=2)
```

Out[35]:

<matplotlib.legend.Legend at 0x224ea73adc0>



