

# COMPARISON

IMAGE  
CLASSIFICATION  
MODELS

CLOUD  
IMAGES  
DATASET



\* CLASSIFICATION \*

# CLOUD IMAGE DATASET

**Data Explorer**

94.53 MB

- └ test
  - └ Cloud
  - └ RainCloud
- └ train
  - └ Cloud
  - └ RainCloud

**Summary**

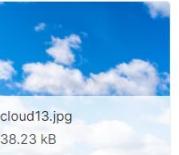
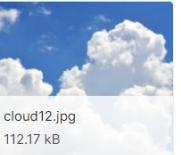
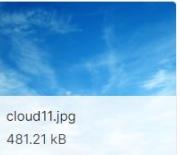
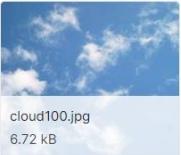
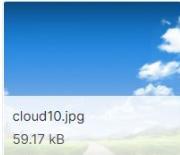
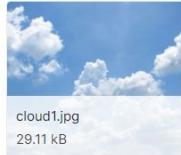
└ 1200 files



◀ Cloud (100 files)

About this directory

This file does not have a description yet.



cloud1.jpg  
29.11 kB

cloud10.jpg  
59.17 kB

cloud100.jpg  
6.72 kB

cloud11.jpg  
481.21 kB

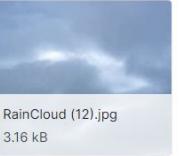
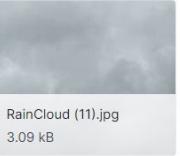
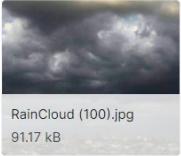
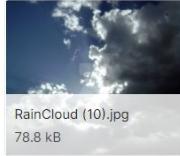
cloud12.jpg  
112.17 kB

cloud13.jpg  
38.23 kB

◀ RainCloud (100 files)

About this directory

This file does not have a description yet.



RainCloud (1).jpg  
4.85 kB

RainCloud (10).jpg  
78.8 kB

RainCloud (100).jpg  
91.17 kB

RainCloud (11).jpg  
3.09 kB

RainCloud (12).jpg  
3.16 kB

RainCloud (13).jpg  
9.38 kB

CLASSIFICATION \* IMAGE \* CLASSIFICATION \* IMAGE \* CLASSIFICATION

# PREPROCESS

## ติดตั้ง Kaggle Library

```
[1] 1 !pip install kaggle

Requirement already satisfied: kaggle in /usr/local/lib/python3.7/dist-packages (1.5.12)
Requirement already satisfied: urllib3 in /usr/local/lib/python3.7/dist-packages (from kaggle) (1.24.3)
Requirement already satisfied: certifi in /usr/local/lib/python3.7/dist-packages (from kaggle) (2021.5.30)
Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from kaggle) (2.23.0)
Requirement already satisfied: tqdm in /usr/local/lib/python3.7/dist-packages (from kaggle) (4.62.0)
Requirement already satisfied: python-dateutil in /usr/local/lib/python3.7/dist-packages (from kaggle) (2.8.2)
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.7/dist-packages (from kaggle) (1.15.0)
Requirement already satisfied: python-slugify in /usr/local/lib/python3.7/dist-packages (from kaggle) (5.0.2)
Requirement already satisfied: text-unidecode>=1.3 in /usr/local/lib/python3.7/dist-packages (from python-slugify->kaggle) (1.3)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->kaggle) (2.10)
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests->kaggle) (3.0.4)
```

## Upload kaggle.json

```
[2] 1 from google.colab import files
2
3 files.upload()

Choose files | kaggle.json
• kaggle.json(application/json) - 75 bytes, last modified: 05/09/2021 - 100% done
Saving kaggle.json to kaggle.json
{'kaggle.json': b'{"username":"nuttidalapthanachai","key":"ee0a24585b49f8ef1fd3f6eab971ee01"}'}
```

# PREPROCESS

สร้างโฟลเดอร์ kaggle เพื่อเก็บไฟล์ kaggle.json และเปลี่ยน Permission ของไฟล์ kaggle.json

```
[3] 1 !mkdir kaggle  
[4] 1 !mv kaggle.json kaggle  
[5] 1 !chmod 600 /content/kaggle/kaggle.json
```

Config Kaggle Environment

```
[6] 1 import os  
2 os.environ['KAGGLE_CONFIG_DIR'] = "/content/kaggle"
```

Download Cloud image dataset

```
[7] 1 !kaggle datasets download -d nuttidalapthanachai/cloud-image-dataset  
Downloading cloud-image-dataset.zip to /content  
94% 83.0M/88.1M [00:00<00:00, 138MB/s]  
100% 88.1M/88.1M [00:00<00:00, 137MB/s]
```

สร้างโฟลเดอร์ Cloud-image-dataset และ Unzip Dataset

```
[8] 1 !mkdir cloud-image-dataset && unzip -q cloud-image-dataset.zip -d cloud-image-dataset
```

# DATA PREPARATION

Import Library ที่จำเป็น

```
[9] 1 import tensorflow as tf
2
3 Adam = tf.keras.optimizers.Adam
4 to_categorical = tf.keras.utils.to_categorical
5 ImageDataGenerator = tf.keras.preprocessing.image.ImageDataGenerator
6 ReduceLROnPlateau = tf.keras.callbacks.ReduceLROnPlateau
7 load_img = tf.keras.preprocessing.image.load_img
8 img_to_array = tf.keras.preprocessing.image.img_to_array
9 array_to_img = tf.keras.preprocessing.image.array_to_img
10
11 from sklearn.model_selection import train_test_split
12 import matplotlib.pyplot as plt
13
14 import plotly.graph_objs as go
15 from plotly import subplots
16 import plotly
17
18 from tensorflow.keras.models import Sequential
19 from tensorflow.keras.layers import Dense, GRU, LSTM, Bidirectional, Embedding, Dropout, BatchNormalization
20 from tensorflow.keras.models import load_model
21 from tensorflow.keras.callbacks import ModelCheckpoint
22
23 import pandas as pd
24 from sklearn.metrics import classification_report
25 from sklearn.metrics import confusion_matrix
26 import pickle as p
27
28 import warnings
29 warnings.filterwarnings('ignore')
```



# DATA PREPARATION



กำหนด parameter

```
[10] 1 img_rows = 32
     2 img_cols = 32
     3 EPOCHS = 100
     4 BATCH_SIZE = 32
```

Load Dataset

```
[11] 1 import cv2
     2 import os
     3 import random
     4 import matplotlib.pyplot as plt
     5 import numpy as np

[12] 1 DIRECTORY_TRAIN = r'/content/cloud-image-dataset/train'
     2 DIRECTORY_TEST = r'/content/cloud-image-dataset/test'
     3 CATEGORIES = ['Cloud','RainCloud']
```



# DATA PREPARATION



สร้างตัวแปร Train\_data และ Test\_data เพื่อเก็บข้อมูล

```
[13] 1  Train_data = []
2  Test_data = []
3
4  for category in CATEGORIES:
5      folder = os.path.join(DIRECTORY_TRAIN, category)
6      label = CATEGORIES.index(category)
7      for img in os.listdir(folder):
8          img_path = os.path.join(folder, img)
9          img_arr = cv2.imread(img_path)
10         img_arr = cv2.resize(img_arr, (img_rows, img_cols))
11         img_arr = cv2.cvtColor(img_arr, cv2.COLOR_BGR2GRAY)
12         Train_data.append([img_arr, label])
13
14     for category in CATEGORIES:
15         folder = os.path.join(DIRECTORY_TEST, category)
16         label = CATEGORIES.index(category)
17         for img in os.listdir(folder):
18             img_path = os.path.join(folder, img)
19             img_arr = cv2.imread(img_path)
20             img_arr = cv2.resize(img_arr, (img_rows, img_cols))
21             img_arr = cv2.cvtColor(img_arr, cv2.COLOR_BGR2GRAY)
22             Test_data.append([img_arr, label])
```

```
[14] 1  print(len(Train_data))
2  print(len(Test_data))
```

```
1000
200
```

```
[15] 1  random.shuffle(Train_data)
2  random.shuffle(Test_data)
```



# DATA PREPARATION



แยกข้อมูลภาพกับผลเฉลย

```
[16] 1 train_data = []
2 y = []
3 test_data = []
4 y_test = []
5
6 for features, labels in Train_data:
7     train_data.append(features)
8     y.append(labels)
9
10 for features, labels in Test_data:
11     test_data.append(features)
12     y_test.append(labels)
```

```
[17] 1 train_data = np.array(train_data)
2 y = np.array(y)
3 test_data = np.array(test_data)
4 y_test = np.array(y_test)
```

```
[18] 1 print("train dataset-rows:",train_data.shape[0]," columns:", train_data.shape[1], " rows:", train_data.shape[2])
2 print("test dataset-rows:",test_data.shape[0]," columns:", test_data.shape[1], " rows:", train_data.shape[2])
```

```
train dataset-rows: 1000 columns: 32 rows: 32
test dataset-rows: 200 columns: 32 rows: 32
```

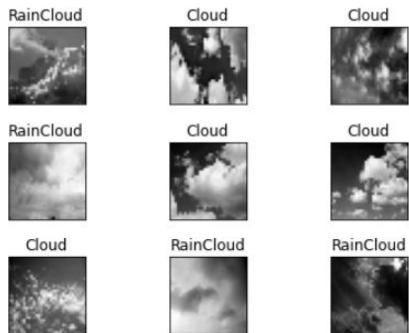


# DATA PREPARATION



แสดงภาพตัวอย่างจาก Train Dataset

```
[19] 1  for i in range(9):
2      |  ax = plt.subplot(3, 3, 1+i)
3      |  ax.set_xticks([])
4      |  ax.set_yticks([])
5      |  ax.set_title('%s'%(CATEGORIES[int(y[i]))]))
6      |  plt.imshow(train_data[i], cmap=plt.get_cmap('gray'))
7  plt.tight_layout()
8  plt.savefig('cloud.png', dpi=300)
```





# DATA PREPARATION



เข้ารหัสผลเฉลยแบบ One-hot Encoding

```
[22] 1 print(y.shape, y_test.shape)
      2 print(y[:9])
      (1000,) (200,)
      [1 0 0 1 0 0 0 1 1]
```

```
[23] 1 y = to_categorical(y)
      2 y_test = to_categorical(y_test)
      3
      4 print(y.shape, y_test.shape)
      5 y[:9]
```

```
(1000, 2) (200, 2)
array([[0., 1.],
       [1., 0.],
       [1., 0.],
       [0., 1.],
       [1., 0.],
       [1., 0.],
       [1., 0.],
       [0., 1.],
       [0., 1.]], dtype=float32)
```

แบ่งข้อมูลสำหรับ Train และ Validate โดยการสุ่มในสัดส่วน 80:20

```
[24] 1 x_train, x_val, y_train, y_val = train_test_split(train_data, y, test_size = 0.2 , random_state = 99)
      2 x_train.shape, x_val.shape, y_train.shape, y_val.shape
      ((800, 32, 32, 1), (200, 32, 32, 1), (800, 2), (200, 2))
```

# \* CLASSIFICATION MODELS \*

CLOUD  
IMAGES  
DATASET

0  
1

**BASELINE  
MODEL**

CLOUD  
IMAGES  
DATASET

# \* CLASSIFICATION MODELS \*

# BASELINE MODEL

## Define Model

```
[25] 1 # model = Sequential()
2 model = tf.keras.Sequential()
3
4 #1. CNN LAYER
5 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same', input_shape=(img_rows, img_cols, 1)))
6 model.add(tf.keras.layers.Activation("relu"))
7
8 #2. CNN LAYER
9 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same'))
10 model.add(tf.keras.layers.Activation("relu"))
11
12 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
13
14 #3. CNN LAYER
15 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
16 model.add(tf.keras.layers.Activation("relu"))
17
18 #4. CNN LAYER
19 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
20 model.add(tf.keras.layers.Activation("relu"))
21
22 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
23
24 #FULLY CONNECTED LAYER
25 model.add(tf.keras.layers.Flatten())
26 model.add(tf.keras.layers.Dense(256))
27 model.add(tf.keras.layers.Activation("relu"))
28
29 #OUTPUT LAYER
30 model.add(tf.keras.layers.Dense(2, activation='sigmoid'))
```

# BASELINE MODEL

## Compile Model

```
[26] 1 optimizer = Adam()
2 model.compile(optimizer = optimizer, loss = "binary_crossentropy", metrics=["accuracy"])
3 model.summary()
```

## Callbacks function

```
[27] 1 learning_rate_reduction = ReduceLROnPlateau(monitor='val_loss', patience = 2, verbose=1,factor=0.1, min_lr = 0.000001)
2 callbacks_list = [learning_rate_reduction]
```

## Train model

```
[28] 1 history = model.fit(x_train, y_train,
2           batch_size= BATCH_SIZE,
3           epochs= EPOCHS,
4           callbacks=callbacks_list,
5           verbose = 1,
6           validation_data=(x_val, y_val))

Epoch 1/100
25/25 [=====] - 3s 28ms/step - loss: 0.5328 - accuracy: 0.7300 - val_loss: 0.4107 - val_accuracy: 0.8150
Epoch 2/100
25/25 [=====] - 0s 18ms/step - loss: 0.2832 - accuracy: 0.8913 - val_loss: 0.3620 - val_accuracy: 0.8500
Epoch 3/100
25/25 [=====] - 0s 17ms/step - loss: 0.2475 - accuracy: 0.9050 - val_loss: 0.3781 - val_accuracy: 0.8450
Epoch 4/100
25/25 [=====] - 0s 17ms/step - loss: 0.2620 - accuracy: 0.8975 - val_loss: 0.3497 - val_accuracy: 0.8500
Epoch 5/100
25/25 [=====] - 0s 17ms/step - loss: 0.2286 - accuracy: 0.9150 - val_loss: 0.4267 - val_accuracy: 0.8650
```

# BASELINE MODEL

## นิยามกราฟ

```
[29] 1 def create_trace(x,y,label,color):
2     trace = go.Scatter(x = x,y = y,name=label,marker=dict(color=color),mode = "markers+lines",text=x)
3     return trace
4
5 def plot_accuracy_and_loss(train_model):
6     hist = train_model.history
7     acc = hist['accuracy']
8     val_acc = hist['val_accuracy']
9     loss = hist['loss']
10    val_loss = hist['val_loss']
11    epochs = list(range(1,len(acc)+1))
12
13    trace_ta = create_trace(epochs,acc,"Training accuracy", "Green")
14    trace_va = create_trace(epochs,val_acc,"Validation accuracy", "Red")
15    trace_tl = create_trace(epochs,loss,"Training loss", "Blue")
16    trace_vl = create_trace(epochs,val_loss,"Validation loss", "Magenta")
17
18    fig = subplots.make_subplots(rows=1,cols=2,
19                                subplot_titles=('Training and validation accuracy','Training and validation loss'))
20    fig.append_trace(trace_ta,1,1)
21    fig.append_trace(trace_va,1,1)
22    fig.append_trace(trace_tl,1,2)
23    fig.append_trace(trace_vl,1,2)
24    fig['layout'][['xaxis']].update(title = 'Epoch')
25    fig['layout'][['xaxis2']].update(title = 'Epoch')
26    fig['layout'][['yaxis']].update(title = 'Accuracy', range=[0,1])
27    fig['layout'][['yaxis2']].update(title = 'Loss', range=[0,1])
28
29    plotly.offline.iplot(fig, filename='accuracy-loss')
30
```

# BASELINE MODEL

## Save History

```
[31] 1  filepath_history_model = 'history_model'
[31] 2  with open(filepath_history_model, 'wb') as file :
[31] 3  |  p.dump(history.history, file)

[32] 1  filepath = 'model.h5'
[32] 2  model.save(filepath)
```

## Load History

```
[33] 1  with open(filepath_history_model, 'rb') as file :
[33] 2  |  his = p.load(file)

[34] 1  predict_model = load_model(filepath)
```

## Evaluate Model

```
[35] 1  score = predict_model.evaluate(test_data, y_test, verbose = 0)
[35] 2  print("Test Loss:",score[0])
[35] 3  print("Test Accuracy:",score[1])
```

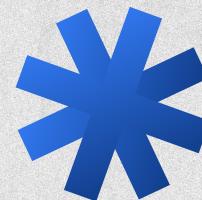
```
Test Loss: 0.2878543734550476
Test Accuracy: 0.8600000143051147
```

# \* CLASSIFICATION MODELS \*

CLOUD  
IMAGES  
DATASET

AUGMENTATION

IMAGE



0  
2

# \* CLASSIFICATION MODELS \*



# IMAGE AUGMENTATION



## นิยาม Image Augmentation

```
[41] 1  datagen = ImageDataGenerator(  
2      |     rotation_range=0.05,      #Randomly rotate images in the range  
3      |     zoom_range = 0.2,        #Randomly zoom image  
4      |     width_shift_range=0.1,   #Randomly shift images horizontally  
5      |     height_shift_range=0.1, #Randomly shift images vertically  
6      |     shear_range=0.05      #Randomly shear images  
7      |     )  
8  
9  datagen.fit(x_train)
```

## ลดมิติภาพ

```
[42] 1  x_batch = datagen.flow(x_train, y_train, batch_size=9).next()  
2  print(x_batch[0].shape)  
3  x_batch = x_batch[0].reshape((x_batch[0].shape[0], img_rows, img_cols))  
4  print(x_batch.shape)  
  
(9, 32, 32, 1)  
(9, 32, 32)
```



# IMAGE AUGMENTATION



## นิยาม Model

```
[43] 1 # model = Sequential()
2 model = tf.keras.Sequential()
3
4 #1. CNN LAYER
5 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same', input_shape=(img_rows, img_cols, 1)))
6 model.add(tf.keras.layers.Activation("relu"))
7
8 #2. CNN LAYER
9 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same'))
10 model.add(tf.keras.layers.Activation("relu"))
11
12 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
13
14 #3. CNN LAYER
15 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
16 model.add(tf.keras.layers.Activation("relu"))
17
18 #4. CNN LAYER
19 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
20 model.add(tf.keras.layers.Activation("relu"))
21
22 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
23
24 #FULLY CONNECTED LAYER
25 model.add(tf.keras.layers.Flatten())
26 model.add(tf.keras.layers.Dense(256))
27 model.add(tf.keras.layers.Activation("relu"))
28
29 #OUTPUT LAYER
30 model.add(tf.keras.layers.Dense(2, activation='sigmoid'))
```



# IMAGE AUGMENTATION



## Compile Model

```
[44] 1 optimizer = Adam()
2 model.compile(optimizer = optimizer, loss = "binary_crossentropy", metrics=["accuracy"])
3 model.summary()
```

## Train Model

```
[45] 1 history = model.fit(datagen.flow(x_train, y_train, batch_size=BATCH_SIZE),
2                               shuffle=True,
3                               epochs=EPOCHS,
4                               callbacks=callbacks_list,
5                               validation_data = (x_val, y_val),
6                               verbose=1,
7                               steps_per_epoch=x_train.shape[0] // BATCH_SIZE)

Epoch 1/100
25/25 [=====] - 2s 34ms/step - loss: 0.6106 - accuracy: 0.6263 - val_loss: 0.4483 - val_accuracy: 0.7900
Epoch 2/100
25/25 [=====] - 1s 25ms/step - loss: 0.4074 - accuracy: 0.8350 - val_loss: 0.3566 - val_accuracy: 0.8500
Epoch 3/100
25/25 [=====] - 1s 28ms/step - loss: 0.2935 - accuracy: 0.8813 - val_loss: 0.3484 - val_accuracy: 0.8300
Epoch 4/100
25/25 [=====] - 1s 26ms/step - loss: 0.2829 - accuracy: 0.8875 - val_loss: 0.3507 - val_accuracy: 0.8300
Epoch 5/100
25/25 [=====] - 1s 25ms/step - loss: 0.2798 - accuracy: 0.8913 - val_loss: 0.3715 - val_accuracy: 0.8550

Epoch 00005: ReduceLROnPlateau reducing learning rate to 0.0001000000474974513.
Epoch 6/100
25/25 [=====] - 1s 27ms/step - loss: 0.2390 - accuracy: 0.9087 - val_loss: 0.3754 - val_accuracy: 0.8550
Epoch 7/100
25/25 [=====] - 1s 27ms/step - loss: 0.2536 - accuracy: 0.8963 - val_loss: 0.3820 - val_accuracy: 0.8600
```



# IMAGE AUGMENTATION



## Save History

```
[47] 1  filepath_history_model = 'history_model'  
2  with open(filepath_history_model, 'wb') as file :  
3  |  p.dump(history.history, file)  
  
[48] 1  filepath = 'model.h5'  
2  model.save(filepath)
```

## Load History

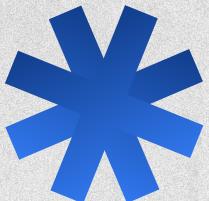
```
[49] 1  with open(filepath_history_model, 'rb') as file :  
2  |  his = p.load(file)  
  
[50] 1  predict_model = load_model(filepath)
```

## Evaluate Model

```
[51] 1  score = predict_model.evaluate(test_data, y_test, verbose = 0)  
2  print("Test Loss:",score[0])  
3  print("Test Accuracy:",score[1])  
  
Test Loss: 0.29108744859695435  
Test Accuracy: 0.8700000047683716
```

# \* CLASSIFICATION MODELS \*

0  
3



CLOUD  
IMAGES  
DATASET

BATCH  
NORMALIZATION

# \* CLASSIFICATION MODELS \*



# BATCH NORMALIZATION



## นิยาม Image Augmentation

```
[57] 1  datagen = ImageDataGenerator(  
2      |    rotation_range=0.05,      #Randomly rotate images in the range  
3      |    zoom_range = 0.2,        #Randomly zoom image  
4      |    width_shift_range=0.1,  #Randomly shift images horizontally  
5      |    height_shift_range=0.1, #Randomly shift images vertically  
6      |    shear_range=0.05      #Randomly shear images  
7      |    )  
8  
9  datagen.fit(x_train)
```

# BATCH NORMALIZATION

นิยาม Model

```
[58] 1 # model = Sequential()
2 model = tf.keras.Sequential()
3
4 #1. CNN LAYER
5 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same', input_shape=(img_rows, img_cols, 1)))
6 model.add(tf.keras.layers.BatchNormalization())
7 model.add(tf.keras.layers.Activation("relu"))
8
9 #2. CNN LAYER
10 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same'))
11 model.add(tf.keras.layers.BatchNormalization())
12 model.add(tf.keras.layers.Activation("relu"))
13
14 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
15
16 #3. CNN LAYER
17 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
18 model.add(tf.keras.layers.BatchNormalization())
19 model.add(tf.keras.layers.Activation("relu"))
20
21 #4. CNN LAYER
22 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
23 model.add(tf.keras.layers.BatchNormalization())
24 model.add(tf.keras.layers.Activation("relu"))
25
26 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
27
28 #FULLY CONNECTED LAYER
29 model.add(tf.keras.layers.Flatten())
30 model.add(tf.keras.layers.Dense(256))
31 model.add(tf.keras.layers.BatchNormalization())
32 model.add(tf.keras.layers.Activation("relu"))
33
34 #OUTPUT LAYER
35 model.add(tf.keras.layers.Dense(2, activation='sigmoid'))
```



# BATCH NORMALIZATION



## Compile Model

```
[59] 1 optimizer = Adam()
2 model.compile(optimizer = optimizer, loss = "binary_crossentropy", metrics=["accuracy"])
3
4 model.summary()
```

## Train Model

```
[60] 1 history = model.fit(datagen.flow(x_train, y_train, batch_size=BATCH_SIZE),
2           shuffle=True,
3           epochs= EPOCHS,
4           callbacks=callbacks_list,
5           validation_data = (x_val, y_val),
6           verbose=1,
7           steps_per_epoch=x_train.shape[0] // BATCH_SIZE)

Epoch 1/100
25/25 [=====] - 2s 40ms/step - loss: 0.2869 - accuracy: 0.8888 - val_loss: 0.8264 - val_accuracy: 0.5250
Epoch 2/100
25/25 [=====] - 1s 28ms/step - loss: 0.1921 - accuracy: 0.9212 - val_loss: 1.1771 - val_accuracy: 0.5250
Epoch 3/100
25/25 [=====] - 1s 29ms/step - loss: 0.1938 - accuracy: 0.9250 - val_loss: 1.3864 - val_accuracy: 0.5250

Epoch 00003: ReduceLROnPlateau reducing learning rate to 0.0001000000474974513.
Epoch 4/100
25/25 [=====] - 1s 29ms/step - loss: 0.1536 - accuracy: 0.9425 - val_loss: 1.4747 - val_accuracy: 0.5250
Epoch 5/100
25/25 [=====] - 1s 29ms/step - loss: 0.1457 - accuracy: 0.9425 - val_loss: 1.6145 - val_accuracy: 0.5250

Epoch 00005: ReduceLROnPlateau reducing learning rate to 1.000000474974514e-05.
Epoch 6/100
25/25 [=====] - 1s 28ms/step - loss: 0.1438 - accuracy: 0.9500 - val_loss: 1.7038 - val_accuracy: 0.5250
Epoch 7/100
25/25 [=====] - 1s 28ms/step - loss: 0.1421 - accuracy: 0.9488 - val_loss: 1.7803 - val_accuracy: 0.5250
```



# BATCH NORMALIZATION



## Save History

```
[62] 1  filepath_history_model = 'history_model'
[62] 2  with open(filepath_history_model, 'wb') as file :
[62] 3  |  p.dump(history.history, file)

[63] 1  filepath = 'model.h5'
[63] 2  model.save(filepath)
```

## Load History

```
[64] 1  with open(filepath_history_model, 'rb') as file :
[64] 2  |  his = p.load(file)

[65] 1  predict_model = load_model(filepath)
```

## Evaluate Model

```
[66] 1  score = predict_model.evaluate(test_data, y_test, verbose = 0)
[66] 2  print("Test Loss:",score[0])
[66] 3  print("Test Accuracy:",score[1])

Test Loss: 0.194719135761261
Test Accuracy: 0.9150000214576721
```

# \* CLASSIFICATION MODELS \*

DROP OUT

0  
4

CLOUD  
IMAGES  
DATASET

CLOUD  
IMAGES  
DATASET

# \* CLASSIFICATION MODELS \*

# DROPOUT MODEL

นิยาม Image Augmentation

```
[72] 1  datagen = ImageDataGenerator(  
2      |     | rotation_range=0.05,    #Randomly rotate images in the range  
3      |     | zoom_range=0.2,       #Randomly zoom image  
4      |     | width_shift_range=0.1, #Randomly shift images horizontally  
5      |     | height_shift_range=0.1, #Randomly shift images vertically  
6      |     | shear_range=0.05   #Randomly shear images  
7      |     )  
8  
9  datagen.fit(x_train)
```

# DROPOUT MODEL

นิยาม Model

```
[73] 1 # model = Sequential()
2 model = tf.keras.Sequential()
3
4 #1. CNN LAYER
5 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same', input_shape=(img_rows, img_cols, 1)))
6 model.add(tf.keras.layers.BatchNormalization())
7 model.add(tf.keras.layers.Activation("relu"))
8 model.add(tf.keras.layers.Dropout(0.1))
9
10 #2. CNN LAYER
11 model.add(tf.keras.layers.Conv2D(filters = 32, kernel_size = (3,3), padding = 'Same'))
12 model.add(tf.keras.layers.BatchNormalization())
13 model.add(tf.keras.layers.Activation("relu"))
14 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
15 model.add(tf.keras.layers.Dropout(0.1))
16
17 #3. CNN LAYER
18 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
19 model.add(tf.keras.layers.BatchNormalization())
20 model.add(tf.keras.layers.Activation("relu"))
21 model.add(tf.keras.layers.Dropout(0.1))
22
23 #4. CNN LAYER
24 model.add(tf.keras.layers.Conv2D(filters = 64, kernel_size = (3,3), padding = 'Same'))
25 model.add(tf.keras.layers.BatchNormalization())
26 model.add(tf.keras.layers.Activation("relu"))
27
28 model.add(tf.keras.layers.MaxPool2D(pool_size=(2, 2)))
29 model.add(tf.keras.layers.Dropout(0.1))
30
31 #FULLY CONNECTED LAYER
32 model.add(tf.keras.layers.Flatten())
33 model.add(tf.keras.layers.Dense(256))
34 model.add(tf.keras.layers.BatchNormalization())
35 model.add(tf.keras.layers.Activation("relu"))
36 model.add(tf.keras.layers.Dropout(0.1))
37
38 #OUTPUT LAYER
39 model.add(tf.keras.layers.Dense(2, activation='sigmoid'))
```

# DROPOUT MODEL

## Compile Model

```
[74] 1 optimizer = Adam()
2 model.compile(optimizer = optimizer, loss = "binary_crossentropy", metrics=["accuracy"])
3 model.summary()
```

## Train Model

```
[75] 1 history = model.fit(datagen.flow(x_train, y_train, batch_size=BATCH_SIZE),
2                                         shuffle=True,
3                                         epochs=EPOCHS,
4                                         callbacks=callbacks_list,
5                                         validation_data = (x_val, y_val),
6                                         verbose = 1,
7                                         steps_per_epoch=x_train.shape[0] // BATCH_SIZE)

Epoch 1/100
25/25 [=====] - 2s 39ms/step - loss: 0.3659 - accuracy: 0.8500 - val_loss: 0.7357 - val_accuracy: 0.5250
Epoch 2/100
25/25 [=====] - 1s 31ms/step - loss: 0.2289 - accuracy: 0.9038 - val_loss: 1.2812 - val_accuracy: 0.5250
Epoch 3/100
25/25 [=====] - 1s 31ms/step - loss: 0.2078 - accuracy: 0.9250 - val_loss: 1.6813 - val_accuracy: 0.5250

Epoch 00003: ReduceLROnPlateau reducing learning rate to 0.00010000000474974513.
Epoch 4/100
25/25 [=====] - 1s 29ms/step - loss: 0.1698 - accuracy: 0.9413 - val_loss: 1.8992 - val_accuracy: 0.5250
Epoch 5/100
25/25 [=====] - 1s 29ms/step - loss: 0.1573 - accuracy: 0.9400 - val_loss: 2.1175 - val_accuracy: 0.5250

Epoch 00005: ReduceLROnPlateau reducing learning rate to 1.0000000474974514e-05.
Epoch 6/100
25/25 [=====] - 1s 28ms/step - loss: 0.1561 - accuracy: 0.9375 - val_loss: 2.2541 - val_accuracy: 0.5250
Epoch 7/100
25/25 [=====] - 1s 29ms/step - loss: 0.1567 - accuracy: 0.9400 - val_loss: 2.3562 - val_accuracy: 0.5250
```

# DROPOUT MODEL

## Save History

```
[77] 1  filepath_history_model = 'history_model'
[77] 2  with open(filepath_history_model, 'wb') as file :
[77] 3  |  p.dump(history.history, file)

[78] 1  filepath = 'model.h5'
[78] 2  model.save(filepath)
```

## Load History

```
[79] 1  with open(filepath_history_model, 'rb') as file :
[79] 2  |  his = p.load(file)

[80] 1  predict_model = load_model(filepath)
```

## Evaluate Model

```
[81] 1  score = predict_model.evaluate(test_data, y_test, verbose = 0)
[81] 2  print("Test Loss:",score[0])
[81] 3  print("Test Accuracy:",score[1])
```

```
Test Loss: 0.17420116066932678
Test Accuracy: 0.9200000166893005
```

CLOUD  
IMAGES  
DATASET

# COMPARE MODELS

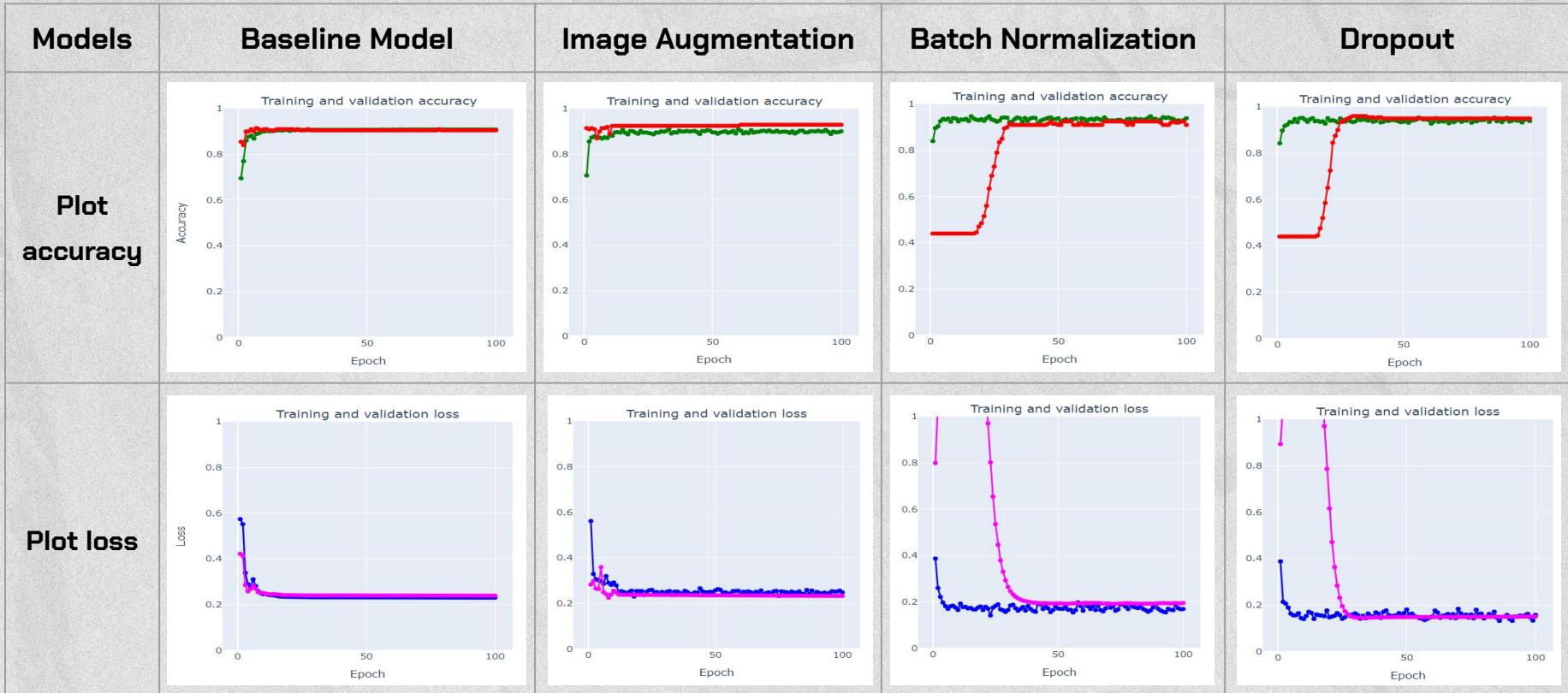
\* CLASSIFICATION \*

\* CLASSIFICATION \*



# COMPARE MODELS

- Training accuracy
- Validation accuracy
- Training loss
- Validation loss



# COMPARE MODELS

Models	Baseline Model	Image Augmentation	Batch Normalization	* Dropout *																																																																																																
Confusion matrix	 <pre> [[93, 7],  [21, 79]] </pre>	 <pre> [[91, 9],  [17, 83]] </pre>	 <pre> [[93, 7],  [10, 90]] </pre>	 <pre> [[93, 7],  [9, 91]] </pre>																																																																																																
Classification Report	<table border="1"> <thead> <tr> <th></th> <th>precision</th> <th>recall</th> <th>f1-score</th> </tr> </thead> <tbody> <tr> <td>Cloud</td> <td>0.8158</td> <td>0.9300</td> <td>0.8692</td> </tr> <tr> <td>RainCloud</td> <td>0.9186</td> <td>0.7900</td> <td>0.8495</td> </tr> <tr> <td>accuracy</td> <td>0.8672</td> <td>0.8600</td> <td>0.8600</td> </tr> <tr> <td>macro avg</td> <td>0.8672</td> <td>0.8600</td> <td>0.8593</td> </tr> <tr> <td>weighted avg</td> <td>0.8672</td> <td>0.8600</td> <td>0.8593</td> </tr> </tbody> </table>		precision	recall	f1-score	Cloud	0.8158	0.9300	0.8692	RainCloud	0.9186	0.7900	0.8495	accuracy	0.8672	0.8600	0.8600	macro avg	0.8672	0.8600	0.8593	weighted avg	0.8672	0.8600	0.8593	<table border="1"> <thead> <tr> <th></th> <th>precision</th> <th>recall</th> <th>f1-score</th> </tr> </thead> <tbody> <tr> <td>Cloud</td> <td>0.8426</td> <td>0.9100</td> <td>0.8750</td> </tr> <tr> <td>RainCloud</td> <td>0.9022</td> <td>0.8300</td> <td>0.8646</td> </tr> <tr> <td>accuracy</td> <td>0.8724</td> <td>0.8700</td> <td>0.8698</td> </tr> <tr> <td>macro avg</td> <td>0.8724</td> <td>0.8700</td> <td>0.8698</td> </tr> <tr> <td>weighted avg</td> <td>0.8724</td> <td>0.8700</td> <td>0.8698</td> </tr> </tbody> </table>		precision	recall	f1-score	Cloud	0.8426	0.9100	0.8750	RainCloud	0.9022	0.8300	0.8646	accuracy	0.8724	0.8700	0.8698	macro avg	0.8724	0.8700	0.8698	weighted avg	0.8724	0.8700	0.8698	<table border="1"> <thead> <tr> <th></th> <th>precision</th> <th>recall</th> <th>f1-score</th> </tr> </thead> <tbody> <tr> <td>Cloud</td> <td>0.9029</td> <td>0.9300</td> <td>0.9163</td> </tr> <tr> <td>RainCloud</td> <td>0.9278</td> <td>0.9000</td> <td>0.9137</td> </tr> <tr> <td>accuracy</td> <td>0.9150</td> <td>0.9150</td> <td>0.9150</td> </tr> <tr> <td>macro avg</td> <td>0.9154</td> <td>0.9150</td> <td>0.9150</td> </tr> <tr> <td>weighted avg</td> <td>0.9154</td> <td>0.9150</td> <td>0.9150</td> </tr> </tbody> </table>		precision	recall	f1-score	Cloud	0.9029	0.9300	0.9163	RainCloud	0.9278	0.9000	0.9137	accuracy	0.9150	0.9150	0.9150	macro avg	0.9154	0.9150	0.9150	weighted avg	0.9154	0.9150	0.9150	<table border="1"> <thead> <tr> <th></th> <th>precision</th> <th>recall</th> <th>f1-score</th> </tr> </thead> <tbody> <tr> <td>Cloud</td> <td>0.9118</td> <td>0.9300</td> <td>0.9208</td> </tr> <tr> <td>RainCloud</td> <td>0.9286</td> <td>0.9100</td> <td>0.9192</td> </tr> <tr> <td>accuracy</td> <td>0.9202</td> <td>0.9200</td> <td>0.9200</td> </tr> <tr> <td>macro avg</td> <td>0.9202</td> <td>0.9200</td> <td>0.9200</td> </tr> <tr> <td>weighted avg</td> <td>0.9202</td> <td>0.9200</td> <td>0.9200</td> </tr> </tbody> </table>		precision	recall	f1-score	Cloud	0.9118	0.9300	0.9208	RainCloud	0.9286	0.9100	0.9192	accuracy	0.9202	0.9200	0.9200	macro avg	0.9202	0.9200	0.9200	weighted avg	0.9202	0.9200	0.9200
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weighted avg	0.9202	0.9200	0.9200																																																																																																	

# THANKS!

## PRESENTED BY

Group : Today is sunday!

620710405 นางสาวณัฐริดา ลาภวนชัย

620710745 นางสาวอาทิตยา ชมทอง

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IMAGES  
DATASET

