Image Pre Processing

```
import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3
def __iter__(self): return 0
# @hidden cell
# The following code accesses a file in your IBM Cloud Object Storage.
It includes your credentials.
# You might want to remove those credentials before you share the
notebook.
cos client = ibm boto3.client(service name='s3',
    ibm api key id='vRbVaygGdy71MU1EguFc 8FXxSewcIbSFXvAd9xXQBJg',
    ibm auth endpoint="https://iam.cloud.ibm.com/oidc/token",
    config=Config(signature version='oauth'),
    endpoint url='https://s3.private.us.cloud-object-
storage.appdomain.cloud')
bucket = 'emergingmethodsforearlydetectiono-donotdelete-pr-
meoatyzjzinmpg'
object key = 'Dataset.zip'
streaming body 1 = cos client.get object(Bucket=bucket,
Key=object key)['Body']
# Your data file was loaded into a botocore.response.StreamingBody
object.
# Please read the documentation of ibm boto3 and pandas to learn more
about the possibilities to load the data.
# ibm boto3 documentation: https://ibm.github.io/ibm-cos-sdk-python/
# pandas documentation: http://pandas.pydata.org/
from io import BytesIO
import zipfile
unzip = zipfile.ZipFile(BytesIO(streaming body 1.read()),'r')
file paths = unzip.namelist()
for path in file paths:
    unzip.extract(path)
pwd
'/home/wsuser/work'
bwd
'/home/wsuser/work'
```

```
import os
filenames=os.listdir('/home/wsuser/work/Dataset/train set')
```

Import Keras libraries

import keras

Importing Image Data Generator from Keras

from matplotlib import pyplot as plt
from keras.preprocessing.image import ImageDataGenerator

Defining the parameter for image generator class

```
train_datagen=ImageDataGenerator(rescale=1./255,shear_range=0.2,rotati
on_range=180,zoom_range=0.2,horizontal_flip=True)
test_datagen=ImageDataGenerator(rescale=1./255,shear_range=0.2,rotatio
n_range=180,zoom_range=0.2,horizontal_flip=True)

#Applying image data generator functionality to trainset
x_train=train_datagen.flow_from_directory('/home/wsuser/work/Dataset/
train_set/',
target_size=(150,150),batch_size=32,class_mode='binary')

Found 436 images belonging to 2 classes.

#Applying image data generator functionality to trainset
x_test=test_datagen.flow_from_directory('/home/wsuser/work/Dataset/
test_set',
target_size=(150,150),batch_size=32,class_mode='binary')

Found 121 images belonging to 2 classes.
```

MODEL BUILDING

```
#To define linear intialisation import Sequential
from keras.models import Sequential
#To add layers import Dense
from keras.layers import Dense
#To creat Convolution kernal import Convolution2D
from keras.layers import Convolution2D
#import Maxpooling layer
from keras.layers import MaxPooling2D
#import Flatten layer
from keras.layers import Flatten
import warnings
warnings.filterwarnings('ignore')
```

```
Initializing the model and Adding CNN and Dense layers
#initializing the model
model=Sequential()
#add convolution laver
model.add(Convolution2D(32,
(3,3),input shape=(150,150,3),activation='relu'))
#add maxpooling layer
model.add(MaxPooling2D(pool size=(2,2)))
#add convolution layer
model.add(Convolution2D(64,(3,3),activation='relu'))
#add maxpooling layer
model.add(MaxPooling2D(pool size=(2,2)))
#add convolution layer
model.add(Convolution2D(128,(3,3),activation='relu'))
#add maxpooling layer
model.add(MaxPooling2D(pool_size=(2,2)))
#add convolution layer
model.add(Convolution2D(128,(3,3),activation='relu'))
#add maxpooling layer
model.add(MaxPooling2D(pool size=(2,2)))
#add flatten layer
model.add(Flatten())
model.add(Dense(512.activation='relu'))
model.add(Dense(1,activation='sigmoid'))
Configuring the learning process
model.compile(loss='binary crossentropy',optimizer="adam",metrics=["ac
curacy"])
Training the model
r=model.fit(x train,epochs=10,validation data=x test)
Epoch 1/10
accuracy: 0.7936 - val loss: 0.2293 - val accuracy: 0.8926
Epoch 2/10
accuracy: 0.8601 - val loss: 0.2015 - val accuracy: 0.9752
Epoch 3/10
accuracy: 0.9266 - val loss: 0.0283 - val accuracy: 0.9917
Epoch 4/10
accuracy: 0.9518 - val loss: 0.0387 - val accuracy: 0.9917
Epoch 5/10
```

```
accuracy: 0.9427 - val loss: 0.0265 - val accuracy: 1.0000
Epoch 6/10
accuracy: 0.9381 - val loss: 0.0399 - val accuracy: 1.0000
Epoch 7/10
accuracy: 0.9404 - val loss: 0.0292 - val accuracy: 0.9917
Epoch 8/10
accuracy: 0.9197 - val loss: 0.0668 - val accuracy: 0.9752
Epoch 9/10
accuracy: 0.9243 - val loss: 0.0404 - val accuracy: 1.0000
Epoch 10/10
accuracy: 0.9427 - val loss: 0.0528 - val accuracy: 0.9917
Save the model
model.save("forestalert.h5")
!tar -zcvf image-classification-model new.tqz forestalert.h5
forestalert.h5
ls -1
Dataset/
forestalert.h5
image-classification-model new.tgz
!pip install watson-machine-learning-client --upgrade
Collecting watson-machine-learning-client
 Downloading watson_machine_learning_client-1.0.391-py3-none-any.whl
(538 kB)
ent already satisfied: ibm-cos-sdk in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (2.11.0)
Requirement already satisfied: tabulate in /opt/conda/envs/Python-
3.9/lib/python3.9/site-packages (from watson-machine-learning-client)
(0.8.9)
Requirement already satisfied: pandas in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (1.3.4)
Requirement already satisfied: certifi in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (2022.9.24)
Requirement already satisfied: requests in /opt/conda/envs/Python-
3.9/lib/python3.9/site-packages (from watson-machine-learning-client)
(2.26.0)
```

```
Requirement already satisfied: boto3 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (1.18.21)
Requirement already satisfied: tgdm in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (4.62.3)
Requirement already satisfied: urllib3 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (1.26.7)
Requirement already satisfied: lomond in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from watson-
machine-learning-client) (0.3.3)
Requirement already satisfied: s3transfer<0.6.0,>=0.5.0 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from boto3-
>watson-machine-learning-client) (0.5.0)
Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from boto3-
>watson-machine-learning-client) (0.10.0)
Requirement already satisfied: botocore<1.22.0,>=1.21.21 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from boto3-
>watson-machine-learning-client) (1.21.41)
Requirement already satisfied: python-dateutil<3.0.0,>=2.1 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from
botocore<1.22.0,>=1.21.21->boto3->watson-machine-learning-client)
(2.8.2)
Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-
3.9/lib/python3.9/site-packages (from python-dateutil<3.0.0,>=2.1-
>botocore<1.22.0,>=1.21.21->boto3->watson-machine-learning-client)
(1.15.0)
Requirement already satisfied: ibm-cos-sdk-s3transfer==2.11.0 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-
sdk->watson-machine-learning-client) (2.11.0)
Requirement already satisfied: ibm-cos-sdk-core==2.11.0 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-
sdk->watson-machine-learning-client) (2.11.0)
Requirement already satisfied: charset-normalizer~=2.0.0 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests-
>watson-machine-learning-client) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-
3.9/lib/python3.9/site-packages (from requests->watson-machine-
learning-client) (3.3)
Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-
3.9/lib/python3.9/site-packages (from pandas->watson-machine-learning-
client) (2021.3)
Requirement already satisfied: numpy>=1.17.3 in
/opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas-
>watson-machine-learning-client) (1.20.3)
Installing collected packages: watson-machine-learning-client
Successfully installed watson-machine-learning-client-1.0.391
```

Replacing the credendials that were got from Watson Machine Learning service

```
from ibm watson machine learning import APIClient
wml credentials={
                 "url": "https://us-south.ml.cloud.ibm.com",
"apikey" :"4edltW4RsVTMfg Ni5NrIwkltXTmh3Y8xg3ms3Ysrt1d"
client = APIClient(wml credentials)
client = APIClient(wml credentials)
def guid_from_space_name(client, space_name):
    space = client.spaces.get details()
    #print(space)
    return(next(item for item in space['resources'] if item['entity']
['name'] == space name)['metadata']['id'])
space uid = guid from space name(client, 'Forest fire detection')
print("Space UID = " + space uid)
Space UID = 4b776d80-ba2d-460b-ac44-dbbc2508e7d5
client.set.default space(space uid)
'SUCCESS'
client.software specifications.list()
NAME
                               ASSET ID
TYPE
default py3.6
                               0062b8c9-8b7d-44a0-a9b9-46c416adcbd9
base
kernel-spark3.2-scala2.12
                               020d69ce-7ac1-5e68-ac1a-31189867356a
base
pytorch-onnx 1.3-py3.7-edt
                               069ea134-3346-5748-b513-49120e15d288
base
scikit-learn 0.20-py3.6
                               09c5a1d0-9c1e-4473-a344-eb7b665ff687
spark-mllib 3.0-scala 2.12
                               09f4cff0-90a7-5899-b9ed-lef348aebdee
base
                               0b848dd4-e681-5599-be41-b5f6fccc6471
pytorch-onnx_rt22.1-py3.9
base
                               Ocdb0f1e-5376-4f4d-92dd-da3b69aa9bda
ai-function 0.1-py3.6
base
                               0e6e79df-875e-4f24-8ae9-62dcc2148306
shiny-r3.6
base
tensorflow 2.4-py3.7-horovod
                               1092590a-307d-563d-9b62-4eb7d64b3f22
base
```

pytorch_1.1-py3.6	10ac12d6-6b30-4ccd-8392-3e922c096a92
base tensorflow 1.15-py3.6-ddl	111e41b3-de2d-5422-a4d6-bf776828c4b7
base	11164103-0620-3422-8400-017700200407
autoai-kb_rt22.2-py3.10	125b6d9a-5b1f-5e8d-972a-b251688ccf40
base	
runtime-22.1-py3.9	12b83a17-24d8-5082-900f-0ab31fbfd3cb
<pre>base scikit-learn 0.22-py3.6</pre>	154010fa-5b3b-4ac1-82af-4d5ee5abbc85
base	13401014-3535-4401-0241-44300345503
default_r3.6	1b70aec3-ab34-4b87-8aa0-a4a3c8296a36
base	
pytorch-onnx_1.3-py3.6	1bc6029a-cc97-56da-b8e0-39c3880dbbe7
<pre>base kernel-spark3.3-r3.6</pre>	1c9e5454-f216-59dd-a20e-474a5cdf5988
base	1C9EJ4J4-1210-J9UU-820E-4748JCU1J900
pytorch-onnx rt22.1-py3.9-edt	1d362186-7ad5-5b59-8b6c-9d0880bde37f
base	
tensorflow_2.1-py3.6	1eb25b84-d6ed-5dde-b6a5-3fbdf1665666
base	20047472 0-00 50-7 0445 -774012-4045
<pre>spark-mllib_3.2 base</pre>	20047f72-0a98-58c7-9ff5-a77b012eb8f5
tensorflow 2.4-py3.8-horovod	217c16f6-178f-56bf-824a-b19f20564c49
base	21,01010 1,01 3001 0210 013120301013
runtime-22.1-py3.9-cuda	26215f05-08c3-5a41-a1b0-da66306ce658
base	
do_py3.8	295addb5-9ef9-547e-9bf4-92ae3563e720
<pre>base autoai-ts_3.8-py3.8</pre>	2aa0c932-798f-5ae9-abd6-15e0c2402fb5
base	2dd0C332-7301-3dC3-dbd0-13C0C2+021b3
tensorflow_1.15-py3.6	2b73a275-7cbf-420b-a912-eae7f436e0bc
base	
kernel-spark3.3-py3.9	2b7961e2-e3b1-5a8c-a491-482c8368839a
<pre>base pytorch_1.2-py3.6</pre>	2c8ef57d-2687-4b7d-acce-01f94976dac1
base	200e1374-2007-4074-acce-01134970dac1
spark-mllib_2.3	2e51f700-bca0-4b0d-88dc-5c6791338875
base	
pytorch-onnx_1.1-py3.6-edt	32983cea-3f32-4400-8965-dde874a8d67e
base	36507ebe-8770-55ba-ab2a-eafe787600e9
spark-mllib_3.0-py37 base	30307ebe-6770-33ba-ab2a-ea1e767000e9
spark-mllib 2.4	390d21f8-e58b-4fac-9c55-d7ceda621326
base	
autoai-ts_rt22.2-py3.10	396b2e83-0953-5b86-9a55-7ce1628a406f
base	39e31acd-5f30-41dc-ae44-60233c80306e
xgboost_0.82-py3.6 base	336310C0-3130-410C-0644-00733C003006
pytorch-onnx_1.2-py3.6-edt	40589d0e-7019-4e28-8daa-fb03b6f4fe12
base	

```
40e73f55-783a-5535-b3fa-0c8b94291431
pytorch-onnx rt22.2-py3.10
base
default_r36py38
                              41c247d3-45f8-5a71-b065-8580229facf0
base
autoai-ts rt22.1-py3.9
                              4269d26e-07ba-5d40-8f66-2d495b0c71f7
base
                              42b92e18-d9ab-567f-988a-4240ba1ed5f7
autoai-obm 3.0
base
pmml-3.0 4.3
                              493bcb95-16f1-5bc5-bee8-81b8af80e9c7
base
spark-mllib 2.4-r 3.6
                              49403dff-92e9-4c87-a3d7-a42d0021c095
base
xgboost 0.90-py3.6
                              4ff8d6c2-1343-4c18-85e1-689c965304d3
base
pytorch-onnx 1.1-py3.6
                              50f95b2a-bc16-43bb-bc94-b0bed208c60b
base
autoai-ts 3.9-py3.8
                              52c57136-80fa-572e-8728-a5e7cbb42cde
                              55a70f99-7320-4be5-9fb9-9edb5a443af5
spark-mllib 2.4-scala 2.11
base
spark-mllib 3.0
                              5c1b0ca2-4977-5c2e-9439-ffd44ea8ffe9
base
autoai-obm 2.0
                              5c2e37fa-80b8-5e77-840f-d912469614ee
base
                              5c3cad7e-507f-4b2a-a9a3-ab53a21dee8b
spss-modeler 18.1
base
cuda-py3.8
                              5d3232bf-c86b-5df4-a2cd-7bb870a1cd4e
base
runtime-22.2-py3.10-xc
                              5e8cddff-db4a-5a6a-b8aa-2d4af9864dab
base
autoai-kb 3.1-py3.7
                              632d4b22-10aa-5180-88f0-f52dfb6444d7
base
_____
                              ______
Note: Only first 50 records were displayed. To display more use
'limit' parameter.
software spec uid =
client.software specifications.get uid by name("tensorflow rt22.1-
py3.9")
software spec uid
'acd9c798-6974-5d2f-a657-ce06e986df4d'
model details = client.repository.store model(model='image-
classification-model new.tgz',meta props={
client.repository.ModelMetaNames.NAME:"CNN",
client.repository.ModelMetaNames.TYPE: "tensorflow rt22.1",
client.repository.ModelMetaNames.SOFTWARE SPEC UID:software spec uid}
model id = client.repository.get model uid(model details)
```

```
This method is deprecated, please use get model id()
model id
'a4275dc7-5c37-4454-afe9-42efaa67bb58'
client.repository.download("a4275dc7-5c37-4454-afe9-42efaa67bb58",
'forestfire1.tar.gz')
Successfully saved model content to file: 'forestfire1.tar.gz'
'/home/wsuser/work/forestfire1.tar.gz'
Predictions(with fire)
#import load model from keras.model
from keras.models import load model
#import image from keras
from tensorflow.keras.preprocessing import image
import numpy as np
#import cv2
import cv2
#load the saved model
model=load model("forestalert.h5")
img=image.load img('/home/wsuser/work/Dataset/test set/with
fire/FORESTFIRE (1).jpg')
x=image.img to array(img)
res=cv2.resize(x,dsize=(150,150),interpolation=cv2.INTER CUBIC)
#expand the image shape
x=np.expand dims(res,axis=0)
pred=model.predict(x)
pred = int(pred[0][0])
pred
int(pred)
1
if pred==1:
    print('Forest Fire')
elif pred==0:
    print('No Fire')
Forest Fire
Predictions(without fire)
#import load model from keras.model
from keras.models import load model
#import image from keras
from tensorflow.keras.preprocessing import image
```

```
import numpy as np
#import cv2
import cv2
#load the saved model
model=load model("forestalert.h5")
img=image.load_img('/home/wsuser/work/Dataset/test_set/forest/55967210
1517195076987621071193712n.jpg')
x=image.img to array(img)
res=cv2.resize(x,dsize=(150,150),interpolation=cv2.INTER CUBIC)
#expand the image shape
x=np.expand_dims(res,axis=0)
pred=model.predict(x)
pred = int(pred[0][0])
pred
int(pred)
if pred==1:
    print('Forest Fire')
elif pred==0:
    print('No Fire')
No Fire
```