VIDEO ANALYSIS

|  |  |
| --- | --- |
| TEAM ID | PNT2022TMID48278 |
| PROJECT NAME | EMERGING METHODS FOR EARLY DEDECTION OF FOREST FIRES |

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"# \*\*DATA COLLECTION\*\*"

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"import tensorflow as tf\n",

"import numpy as np\n",

"from tensorflow import keras\n",

"import os\n",

"import cv2\n",

"from tensorflow.keras.preprocessing.image import ImageDataGenerator\n",

"from tensorflow.keras.preprocessing import image\n",

"import matplotlib.pyplot as plt"

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"train = ImageDataGenerator(rescale=1/255)\n",

"test = ImageDataGenerator(rescale=1/255)\n",

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"train\_dataset = train.flow\_from\_directory(\"/content/drive/MyDrive/IBM/Dataset/train\_set\",\n",

" target\_size=(150,150),\n",

" batch\_size = 32,\n",

" class\_mode = 'binary')\n",

"\n",

"test\_dataset = test.flow\_from\_directory(\"/content/drive/MyDrive/IBM/Dataset/test\_set\",\n",

" target\_size=(150,150),\n",

" batch\_size =32,\n",

" class\_mode = 'binary')"

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"{'Fire': 0, 'NoFire': 1}"

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"# \*\*MODEL BUILDING AND FITTING\*\*"

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"model = keras.Sequential()\n",

"model.add(keras.layers.Conv2D(32,(3,3),activation='relu',input\_shape=(150,150,3)))\n",

"model.add(keras.layers.MaxPool2D(2,2))\n",

"model.add(keras.layers.Conv2D(64,(3,3),activation='relu'))\n",

"model.add(keras.layers.MaxPool2D(2,2))\n",

"model.add(keras.layers.Conv2D(128,(3,3),activation='relu'))\n",

"model.add(keras.layers.MaxPool2D(2,2))\n",

"model.add(keras.layers.Conv2D(128,(3,3),activation='relu'))\n",

"model.add(keras.layers.MaxPool2D(2,2))\n",

"model.add(keras.layers.Flatten())\n",

"model.add(keras.layers.Dense(512,activation='relu'))\n",

"model.add(keras.layers.Dense(1,activation='sigmoid'))"

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"\*\*COMPILE THE MODEL ✈\*\*"

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"model.compile(optimizer='adam',loss='binary\_crossentropy',metrics=['accuracy'])"

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"\*\*FIT THE MODEL ⚡\*\*"

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"r = model.fit(train\_dataset,\n",

" epochs = 10,\n",

" validation\_data = test\_dataset)"

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"Epoch 2/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.6388 - accuracy: 0.5500 - val\_loss: 0.9704 - val\_accuracy: 0.5000\n",

"Epoch 3/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.9761 - accuracy: 0.5000 - val\_loss: 0.6340 - val\_accuracy: 0.5000\n",

"Epoch 4/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.6478 - accuracy: 0.5000 - val\_loss: 0.6171 - val\_accuracy: 0.9500\n",

"Epoch 5/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.6174 - accuracy: 0.8500 - val\_loss: 0.6139 - val\_accuracy: 0.7500\n",

"Epoch 6/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.6048 - accuracy: 0.8000 - val\_loss: 0.5628 - val\_accuracy: 0.8500\n",

"Epoch 7/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.5632 - accuracy: 0.8500 - val\_loss: 0.4429 - val\_accuracy: 1.0000\n",

"Epoch 8/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.4820 - accuracy: 0.8500 - val\_loss: 0.2908 - val\_accuracy: 1.0000\n",

"Epoch 9/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.3741 - accuracy: 0.9000 - val\_loss: 0.1425 - val\_accuracy: 1.0000\n",

"Epoch 10/10\n",

"1/1 [==============================] - 2s 2s/step - loss: 0.2571 - accuracy: 0.9000 - val\_loss: 0.0666 - val\_accuracy: 1.0000\n"

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"\*\*LOSS VS ITERATION GRAPH ⤴\*\*"

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"import seaborn as plt\n",

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"plt.relplot(data=r.history, kind=\"line\")"

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"\*\*ACCURACY/ITERATION ⤴\*\*"

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"\*\*FUNCTION - RETURNS IMAGE WITH LABEL 🌴\*\*"

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"import matplotlib.pyplot as plt\n",

"import random\n",

"\n",

"def predictImage(filename, message = False):\n",

" img1 = image.load\_img(filename,target\_size=(150,150))\n",

" plt.imshow(img1)\n",

" Y = image.img\_to\_array(img1)\n",

" X = np.expand\_dims(Y,axis=0)\n",

" val = model.predict(X)\n",

" print(val)\n",

" if val == 0:\n",

" if message: sendMessage(\"Fire Detected\")\n",

" return plt.xlabel(\"Fire\",fontsize=30)\n",

" elif val == 1:\n",

" return plt.xlabel(\"No Fire\",fontsize=30)"

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"def loadAndPredictImage(className):\n",

" local\_path = \"/content/drive/MyDrive/IBM/Dataset/test\_set/\" + className + \"/\"\n",

" files = []\n",

" for filename in os.listdir(local\_path):\n",

" if filename.endswith(\"jpg\"): \n",

" files.append(local\_path+filename)\n",

"\n",

" ran = random.choice(files)\n",

" predictImage(ran)"

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"# \*\*VIDEO PROCESSING\*\*"

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"from google.colab.patches import cv2\_imshow\n",

"\n",

"#define video from file\n",

"video=cv2.VideoCapture(\"/content/Pexels Videos 2330708.mp4\") \n",

"#define the features \n",

"name=['forest','with fire']\n",

"\n",

"while(1):\n",

" success, frame= video.read() \n",

" cv2.imwrite(\"image.jpg\",frame)\n",

" img=image.load\_img(\"image.jpg\",target\_size=(150,150)) \n",

" x=image.img\_to\_array(img) \n",

" x=np.expand\_dims(x,axis=0)\n",

" pred=model.predict(x) \n",

" print(pred[0])\n",

" cv2.putText(frame,\"predicted class=\"+str(name[int(pred[0])]),(100,100),cv2.FONT\_HERSHEY\_SIMPLEX,1, (0,0,0), 1) \n",

" \n",

" if pred[0]==1:\n",

" # sendMessage(\"Forest Fire is detected, stay alert\")\n",

" print(message.sid)\n",

" print('Fire Detected') \n",

" print('SMS sent!')\n",

" else:\n",

" print('No Danger') \n",

"video.release() \n",

"cv2.destryoAllWindows()"

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