Al-powered Nutrition Analyzer For Fitness Enthusiasts

Sprint - 4

Date :	17/11/2022
Team Id:	PNT2022TMID37915
Domain Name :	Al-powered Nutrition Analyzer
	for Fitness Enthusiasts

```
"nbformat": 4,
 "nbformat_minor": 0,
 "metadata": {
  "colab": {
   "provenance": [],
   "collapsed_sections": []
  "kernelspec": {
   "name": "python3",
   "display_name": "Python 3"
  "language_info": {
   "name": "python"
 },
 "cells": [
   "cell_type": "code",
   "source": [
     "#Importing Neccessary Libraries\n",
     "\n",
     "import numpy as np\n",
     "#used for numerical analysis\n",
     "import tensorflow #open source used for both ML and DL for computation\n",
     "from tensorflow.keras.models import Sequential #it is a plain stack of layers\n",
     "from tensorflow.keras import layers # a layer consists of a tensor-in tensor-out computation
function\n",
     "#Dense layer is the regular deeply connected neural network layer\n",
     "from tensorflow.keras.layers import Dense, Flatten \n",
     "#Flatten-used fot flattering the input or change the dimension\n",
     "from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dropout #convolutional layer\n",
     "#MaxPooling2D-for downsampling the image\n",
     "from keras.preprocessing.image import ImageDataGenerator"
   ],
   "metadata": {
    "id": "XdON9-MKqt-K"
```

```
},
 "execution count": 1,
 "outputs": []
 "cell_type": "code",
 "source": [
  "from google.colab import drive\n",
  "drive.mount('/content/drive')"
 ],
 "metadata": {
  "colab": {
   "base_uri": "https://localhost:8080/"
  },
  "id": "Yac7zN7nb907",
  "outputId": "1e2e3a47-587f-4485-8aab-1f5aa1945444"
 },
 "execution_count": 18,
 "outputs": [
   "output_type": "stream",
   "name": "stdout",
   "text": [
    "Mounted at /content/drive\n"
   1
  }
]
 "cell_type": "code",
 "source": [
  "import tensorflow as tf\n",
  ''\n'',
  "from tensorflow.keras import datasets, layers, models\n",
  "import matplotlib.pyplot as plt"
 "metadata": {
  "id": "NbjcLABwTD-f"
 "execution_count": 5,
 "outputs": []
},
 "cell_type": "code",
 "source": [
  "(train_images, train_labels), (test_images, test_labels) = datasets.cifar10.load_data()\n",
  "# Normalize pixel values to be between 0 and 1\n",
  "train_images, test_images = train_images / 255.0, test_images / 255.0"
 ],
 "metadata": {
  "colab": {
```

```
"base uri": "https://localhost:8080/"
  },
  "id": "fOxgYCO6THhC",
  "outputId": "8b9319aa-ca4c-465a-f05c-2df9cbd49f5a"
 "execution_count": 6,
 "outputs": [
   "output_type": "stream",
   "name": "stdout",
   "text": [
    "Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz\n",
    }
1
},
 "cell_type": "code",
 "source": [
  "#Creating the model\n",
  "model = models.Sequential()\n",
  "model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)))\n",
  "model.add(layers.MaxPooling2D((2, 2)))\n",
  "model.add(layers.Conv2D(64, (3, 3), activation='relu'))\n",
  "model.add(layers.MaxPooling2D((2, 2)))\n",
  "model.add(layers.Conv2D(64, (3, 3), activation='relu'))\n",
  "model.add(layers.Flatten())\n",
  "model.add(layers.Dense(64, activation='relu'))\n",
  "model.add(layers.Dense(10))\n"
],
 "metadata": {
  "id": "169yIjH3lrlT"
 },
 "execution count": 7,
 "outputs": []
},
 "cell_type": "code",
 "source": [
  "model.summary()"
],
 "metadata": {
  "colab": {
   "base_uri": "https://localhost:8080/"
  "id": "q5fdM_m2lvrx",
  "outputId": "69314f45-2b78-460b-cb5e-893840945ff1"
 "execution_count": 8,
 "outputs": [
```

```
"output_type": "stream",
   "name": "stdout",
   "text": [
    "Model: \"sequential\"\n",
                                                                                      _\n'',
    " Layer (type)
                            Output Shape
                                                   Param# \n",
    " conv2d (Conv2D)
                                (None, 30, 30, 32)
                                                        896
                                                               \n'',
                                              n''
                                                                     \n'',
    " max_pooling2d (MaxPooling2D (None, 15, 15, 32)
                                              \n'',
                                              \n",
    " conv2d_1 (Conv2D)
                                 (None, 13, 13, 64)
                                                         18496
                                                                  n'',
                                              \n'',
    " max_pooling2d_1 (MaxPooling (None, 6, 6, 64)
                                                             0
                                                                   n''
    " 2D)
                                                n'',
                                              \n'',
    " conv2d_2 (Conv2D)
                                 (None, 4, 4, 64)
                                                        36928
                                                                 n'',
                                              \n'',
    " flatten (Flatten)
                            (None, 1024)
                                                  0
                                                         n'',
                                              \n",
    " dense (Dense)
                             (None, 64)
                                                  65600
                                                           \n'',
                                              n'',
    " dense_1 (Dense)
                                                            n'',
                               (None, 10)
                                                    650
                                              \n'',
    "Total params: 122,570\n",
    "Trainable params: 122,570\n",
    "Non-trainable params: 0\n",
                                                                                       _\n''
]
"cell_type": "code",
"source": [
 "#Compiling the model\n",
 "model.compile(optimizer='adam',\n",
           loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),\n",
      metrics=['accuracy'])\n",
 "#Fitting the model\n",
 "history = model.fit(train_images, train_labels, epochs=10, \n",
               validation data=(test images, test labels))"
],
"metadata": {
 "colab": {
  "base_uri": "https://localhost:8080/"
 },
 "id": "7TWsgR2kl1K5",
},"outputId": "cd1fd463-08eb-43b2-a3eb-12ca48b1d855"
```

```
"execution_count": 9,
  "outputs": [
   "output type": "stream",
   "name": "stdout",
   "text": [
    "Epoch 1/10\n",
    accuracy:
0.4449 - val loss: 1.2775 - val accuracy: 0.5395\n",
"Epoch 2/10\n",
    accuracy:
0.5837 - val loss: 1.2141 - val accuracy: 0.5757\n",
"Epoch 3/10\n",
    accuracy:
0.6392 - val loss: 0.9934 - val accuracy: 0.6493\n",
"Epoch 4/10\n",
    accuracy:
0.6738 - val loss: 0.9645 - val accuracy: 0.6568\n",
"Epoch 5/10\n",
    accuracy:
0.6982 - val_loss: 0.8927 - val_accuracy: 0.6906\n",
"Epoch 6/10\n",
    accuracy:
0.7185 - val loss: 0.8897 - val accuracy: 0.6925\n",
"Epoch 7/10\n",
    "1563/1563 [=======] - 77s 49ms/step - loss: 0.7532 -
accuracy:
0.7351 - val_loss: 0.9193 - val_accuracy: 0.6885\n",
"Epoch 8/10\n",
    accuracy:
0.7508 - val loss: 0.8996 - val accuracy: 0.6962\n",
"Epoch 9/10\n",
    "1563/1563 [======] - 76s 48ms/step - loss: 0.6726 -
accuracy:
0.7641 - val loss: 0.8864 - val accuracy: 0.6996\n",
"Epoch 10/10\n",
    "1563/1563 [======] - 76s 49ms/step - loss: 0.6358 -
accuracy:
0.7755 - val loss: 0.9306 - val accuracy: 0.6936\n"
   1
   }
  ]
 },
  "cell_type": "code",
```

```
"source": [
     "#Saving our model\n",
     "model.save('nutrition.h5')"
   ],
    "metadata": {
     "id": "WLvInNYXpQTz"
    "execution count": 11,
    "outputs": []
   },
    "cell_type": "code",
    "source": [
     "#Prediciting our results\n",
     "from tensorflow.keras.models import load_model\n",
     "from tensorflow.keras.preprocessing import image\n",
     "model=load model('nutrition.h5')"
   ],
    "metadata": {
     "id": "SW74x3bapQiW"
    "execution count": 16,
    "outputs": []
    "cell_type": "code",
    "source": [
     "img=image.load img('/content/drive/MyDrive/Nutrition Image Analysis using CNN and Rapid
API-20221106T044103Z-001/Nutrition
                                           Image
                                                      Analysis
                                                                    using
                                                                              CNN
                                                                                        and
                                                                                                 Rapid
API/Dataset/TRAIN_SET/APPLES/n07740461_10065.jpg',target_size=(70,70))\n'',
     "img"
   ],
    "metadata": {
     "colab": {
      "base_uri": "https://localhost:8080/",
      "height": 87
     "id": " OrLiTErq2xo",
     "outputId": "af059ad7-d898-4517-8f0f-5c16c437e8a0"
    },
    "execution_count": 19,
    "outputs": [
      "output_type": "execute_result",
      "data": {
       "text/plain": [
       "image/png":
      "metadata": {},
      "execution_count": 19
```

```
]
},
 "cell_type": "code",
 "source": [
  "x= image.img_to_array(img)\n"
 ],
 "metadata": {
  "id": "LN8xqzTar0Lb"
 "execution_count": 20,
 "outputs": []
 "cell type": "code",
 "source": [
  x = \text{np.expand\_dims}(x, axis=0)
 "metadata": {
  "id": "p8lCho1Rr-E9"
 "execution_count": 21,
 "outputs": []
 "cell_type": "code",
 "source": [
  "\n",
  "index=['APPLES', 'BANANA', 'ORANGE', 'PINEAPPLE', 'WATERMELON']\n", "result=str(index[0])\n",
  "result"
 ],
 "metadata": {
  "colab": {
   "base_uri": "https://localhost:8080/",
   "height": 36
  "id": "2fI9nwY7sSvT",
  "outputId": "acade409-66c4-4d6f-c974-7e3edb8f5570"
 },
 "execution_count": 27,
 "outputs": [
   "output_type": "execute_result",
   "data": {
     "text/plain": [
      "APPLES"
     "application/vnd.google.colaboratory.intrinsic+json": {
      "type": "string"
     }
   },
```