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      "# **Trained by Team ID : PNT2022TMID17050**"
    ],
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forcibly remount, call drive.mount(\"/content/drive\",
force remount=True).\n"
      ],
      "source": [
       "from google.colab import drive\n",
        "drive.mount('/content/drive')"
      1
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          "text": [
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[A]11, [N]one, [r]ename: N\n"
        }
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        "# **STEP 2 Image** **Augumentation**"
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        "from tensorflow.keras.preprocessing.image import
ImageDataGenerator"
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        "train datagen=ImageDataGenerator(rescale=1./255,
zoom range=0.2,horizontal flip=True,vertical flip=False)"
      ],
      "metadata": {
        "id": "9yZUiTxnr0UN"
      "execution count": 5,
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       "test datagen=ImageDataGenerator(rescale=1./255)"
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      "execution count": 6,
```

```
"outputs": []
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"x train=train datagen.flow from directory(r\"/content/drive/MyDrive/AI IBM
/flowers\", target size=(64,64), class mode='categorical', batch size=24)"
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      },
      "execution count": 7,
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          1
        }
      ]
    },
      "cell type": "code",
      "source": [
"x test=test datagen.flow from directory(r\"/content/drive/MyDrive/AI IBM/f
lowers\",target size=(64,64),class mode='categorical',batch size=24)"
      ],
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            "Found 4317 images belonging to 5 classes.\n"
        }
      1
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      "source": [
        "x train.class indices"
      ],
```

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'tulip': 4}"
          },
          "metadata": {},
          "execution_count": 9
        }
      ]
    },
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      "source": [
        "# **Step -3 Initializing CNN And Create Model**"
      "metadata": {
        "id": "05cz-9q0JM s"
    },
      "cell type": "code",
      "source": [
        "from tensorflow.keras.models import Sequential\n",
        "from tensorflow.keras.layers import
Dense, Convolution2D, MaxPooling2D, Flatten"
      "metadata": {
        "id": "QAUHi2otRcoC"
      },
      "execution count": 10,
      "outputs": []
    },
      "cell type": "markdown",
      "source": [
       "# **Step -4 Add layers**"
      "metadata": {
        "id": "xew7skua3a0z"
    },
      "cell type": "code",
      "source": [
        "model=Sequential()"
```

```
],
      "metadata": {
       "id": "dack9NXYR2t6"
      "execution count": 11,
      "outputs": []
    },
      "cell type": "markdown",
      "source": [
       "# **4.1 Input Layers (Convolution ,MaxPooling,Flatten) **"
      "metadata": {
       "id": "SzIvL8Q52DFR"
      }
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      "cell type": "code",
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"model.add(Convolution2D(32,(3,3),input shape=(64,64,3),activation='relu'))
      "metadata": {
       "id": "qPUbKxHGR7EX"
      "execution_count": 12,
      "outputs": []
    },
      "cell_type": "code",
      "source": [
       "model.add(MaxPooling2D(pool size=(2,2)))"
      "metadata": {
       "id": "IBGMZ7sSSAIB"
      "execution count": 13,
      "outputs": []
    },
    {
      "cell type": "code",
      "source": [
       "model.add(Flatten())"
      "metadata": {
       "id": "c65fXm9KSErL"
      "execution_count": 14,
      "outputs": []
   },
      "cell_type": "code",
      "source": [
        "model.summary()"
      "metadata": {
```

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          "Model: \"sequential\"\n",
                                                        _\n",
          " Layer (type)
                                  Output Shape
                                                          Param #
n",
"-----\n",
          " conv2d (Conv2D)
                                   (None, 62, 62, 32)
                                                         896
n",
\n",
          " max pooling2d (MaxPooling2D (None, 31, 31, 32) 0
\n",
          ")
\n",
\n",
          " flatten (Flatten) (None, 30752)
n",
\n",
"==========\n",
         "Total params: 896\n",
          "Trainable params: 896\n",
          "Non-trainable params: 0\n",
      }
     ]
   },
     "cell type": "markdown",
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      "# **4.2 Hidden Layers**"
     "metadata": {
      "id": "f4ZSQPIFJaeb"
     }
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     "cell type": "code",
     "source": [
      "model.add(Dense(300, activation='relu')) \n",
```

```
"model.add(Dense(150,activation='relu'))"
      ],
      "metadata": {
        "id": "x8MIUG1PSZ21"
      "execution count": 16,
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      "cell type": "markdown",
      "source": [
       "# **4.3 Output Layers**"
      ],
      "metadata": {
       "id": "PNLk8KHHJf3K"
    },
      "cell_type": "code",
      "source": [
        "model.add(Dense(5,activation='softmax'))"
      "metadata": {
       "id": "grI0IbuwSeg0"
      "execution_count": 17,
      "outputs": []
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      "cell type": "code",
      "source": [
"model.compile(loss='categorical crossentropy',optimizer='adam',metrics=['a
ccuracy'])"
      ],
      "metadata": {
       "id": "144vMW4QShaw"
      "execution count": 18,
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      "cell type": "code",
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       "len(x_train)"
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         "execution count": 19
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       "# **Step -5 Train the Model**"
     "metadata": {
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   },
     "cell type": "code",
     "source": [
       "model.fit generator(x train, steps per epoch=len(x train),
validation data=x test, validation steps=len(x test), epochs= 30)"
     ],
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packages/ipykernel launcher.py:1: UserWarning: `Model.fit generator` is
deprecated and will be removed in a future version. Please use `Model.fit`,
which supports generators.\n",
           " \"\"Entry point for launching an IPython kernel.\n"
       },
         "output type": "stream",
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           "Epoch 1/30\n",
           "180/180 [============= ] - 393s 2s/step -
loss: 1.3213 - accuracy: 0.4714 - val loss: 1.1275 - val accuracy:
0.5532\n",
           "Epoch 2/30\n",
           loss: 1.0600 - accuracy: 0.5854 - val loss: 0.9406 - val accuracy:
0.6301\n'',
```

```
"Epoch 3/30\n",
          "180/180 [============= ] - 73s 405ms/step -
loss: 0.9678 - accuracy: 0.6247 - val loss: 0.9603 - val accuracy:
0.6203\n'',
          "Epoch 4/30\n",
          "180/180 [============ ] - 77s 429ms/step -
loss: 0.8884 - accuracy: 0.6546 - val loss: 0.8187 - val accuracy:
0.6938\n",
          "Epoch 5/30\n",
          loss: 0.8358 - accuracy: 0.6787 - val loss: 0.7393 - val accuracy:
0.7225\n'',
          "Epoch 6/30\n",
          "180/180 [============== ] - 75s 418ms/step -
loss: 0.7924 - accuracy: 0.6965 - val loss: 0.8389 - val accuracy:
0.6928\n",
          "Epoch 7/30\n",
          "180/180 [============= ] - 73s 405ms/step -
loss: 0.7521 - accuracy: 0.7158 - val loss: 0.8503 - val accuracy:
0.6789\n",
          "Epoch 8/30\n",
          "180/180 [============== ] - 74s 411ms/step -
loss: 0.7048 - accuracy: 0.7313 - val loss: 0.6492 - val accuracy:
0.7521\n'',
          "Epoch 9/30\n",
          "180/180 [============= ] - 72s 400ms/step -
loss: 0.6502 - accuracy: 0.7521 - val loss: 0.6458 - val accuracy:
0.7438\n'',
          "Epoch 10/30\n",
          loss: 0.6182 - accuracy: 0.7684 - val loss: 0.5721 - val accuracy:
0.7818\n'',
          "Epoch 11/30\n",
          "180/180 [============= ] - 72s 402ms/step -
loss: 0.5662 - accuracy: 0.7931 - val loss: 0.5968 - val accuracy:
0.7725\n'',
          "Epoch 12/30\n",
          "180/180 [============= ] - 72s 401ms/step -
loss: 0.5600 - accuracy: 0.7908 - val loss: 0.6907 - val accuracy:
0.7612\n'',
          "Epoch 13/30\n",
          "180/180 [============= ] - 72s 399ms/step -
loss: 0.5064 - accuracy: 0.8138 - val loss: 0.5185 - val accuracy:
0.8117\n",
          "Epoch 14/30\n",
          "180/180 [============ ] - 71s 394ms/step -
loss: 0.4830 - accuracy: 0.8249 - val loss: 0.3613 - val accuracy:
0.8673\n",
          "Epoch 15/30\n",
          loss: 0.4650 - accuracy: 0.8196 - val loss: 0.3396 - val accuracy:
0.8768\n",
          "Epoch 16/30\n",
          "180/180 [=========== ] - 71s 393ms/step -
loss: 0.4117 - accuracy: 0.8559 - val loss: 0.3472 - val accuracy:
0.8738\n'',
          "Epoch 17/30\n",
```

```
loss: 0.3892 - accuracy: 0.8631 - val loss: 0.3314 - val accuracy:
0.8826\n",
         "Epoch 18/30\n",
          "180/180 [============= ] - 70s 389ms/step -
loss: 0.3441 - accuracy: 0.8726 - val loss: 0.4008 - val accuracy:
0.8589\n'',
         "Epoch 19/30\n",
         "180/180 [============ ] - 73s 404ms/step -
loss: 0.3467 - accuracy: 0.8719 - val loss: 0.2484 - val accuracy:
0.9060\n",
         "Epoch 20/30\n",
          "180/180 [============= ] - 72s 398ms/step -
loss: 0.3327 - accuracy: 0.8758 - val loss: 0.2234 - val accuracy:
0.9210\n'',
          "Epoch 21/30\n",
          loss: 0.2807 - accuracy: 0.9009 - val loss: 0.2830 - val accuracy:
0.9036\n",
         "Epoch 22/30\n",
          "180/180 [============= ] - 70s 392ms/step -
loss: 0.2751 - accuracy: 0.9013 - val loss: 0.2392 - val accuracy:
0.9141\n'',
          "Epoch 23/30\n",
          "180/180 [============= ] - 73s 404ms/step -
loss: 0.2549 - accuracy: 0.9097 - val loss: 0.2221 - val accuracy:
0.9189\n",
         "Epoch 24/30\n",
          "180/180 [============= ] - 72s 399ms/step -
loss: 0.2412 - accuracy: 0.9243 - val loss: 0.2029 - val accuracy:
0.9291\n'',
         "Epoch 25/30\n",
         "180/180 [============== ] - 72s 402ms/step -
loss: 0.2360 - accuracy: 0.9199 - val loss: 0.1965 - val accuracy:
0.9307 \n'',
          "Epoch 26/30\n",
         "180/180 [============ ] - 72s 401ms/step -
loss: 0.2199 - accuracy: 0.9201 - val loss: 0.1919 - val accuracy:
0.9331\n",
          "Epoch 27/30\n",
          loss: 0.2008 - accuracy: 0.9363 - val loss: 0.1218 - val accuracy:
0.9560\n",
         "Epoch 28/30\n",
          loss: 0.1889 - accuracy: 0.9310 - val loss: 0.2838 - val accuracy:
0.9108\n",
          "Epoch 29/30\n",
         "180/180 [============= ] - 70s 389ms/step -
loss: 0.2046 - accuracy: 0.9275 - val loss: 0.2116 - val accuracy:
0.9307\n",
         "Epoch 30/30\n",
          "180/180 [============ ] - 70s 392ms/step -
loss: 0.1886 - accuracy: 0.9372 - val_loss: 0.2091 - val_accuracy:
0.9280\n"
      },
```

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 ]
},
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 "source": [
   "# **Step -6 Save The model**"
 "metadata": {
   "id": "1uK880jw9Kru"
  }
},
{
  "cell type": "code",
 "source": [
   "model.save('Flowers classification model1.h5')"
 "metadata": {
   "id": "scoaKurE9FZk"
 "execution_count": 21,
 "outputs": []
},
  "cell_type": "markdown",
 "source": [
   "# **Step -7 Test The model**"
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  }
},
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   "ls"
 "metadata": {
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   "id": "Z-co6hBAEmzq",
    "outputId": "bf8a661d-3210-4695-dcb7-48e6f365dfce"
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Flowers classification model1.h5 Flowers-Dataset.zip video.mp4\n"
    },
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        "from tensorflow.keras.models import load model\n",
        "from tensorflow.keras.preprocessing import image"
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        "id": "mJvRRo7VvkeO"
      "execution count": 23,
      "outputs": []
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      "cell type": "code",
      "source": [
       "# Load the model\n",
        "model=load model('Flowers classification model1.h5')"
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        "id": "xo6F 4jw9KBZ"
      "execution count": 24,
      "outputs": []
    },
      "cell type": "code",
      "source": [
        "img=image.load img(r\"/content/s3.jpg\",target size=(64,64))\n",
        "x=image.img to array(img)\n",
        "x=np.expand dims(x,axis=0)\n",
        "y=np.argmax(model.predict(x),axis=1)\n",
        "# x train.class_indices\n",
        "index=['daisy','dandelion','rose','sunflower','tulip']\n",
        "index[y[0]]"
      ],
      "metadata": {
        "colab": {
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          "height": 35
        "id": "2rnrfMAf-AB9",
        "outputId": "c6357a8b-5163-4884-c82e-05651a65571c"
      "execution count": 38,
      "outputs": [
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           "'sunflower'"
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           "type": "string"
        },
        "metadata": {},
        "execution_count": 38
     }
    ]
  },
    "cell type": "markdown",
    "source": [
     "# **We Achieved 93 percent of accuracy with this model** \n",
      "# **Trained by Team ID : PNT2022TMID17050**"
    "metadata": {
     "id": "2f85wU8fL0Si"
 }
]
```