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      "# **Trained by Team ID : PNT2022TMID17050**"
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            "Drive already mounted at /content/drive; to attempt to
forcibly remount, call drive.mount(\"/content/drive\",
force_remount=True).\n"
        }
      ],
      "source":
        "from google.colab import drive\n",
        "drive.mount('/content/drive')"
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        "# **STEP 1 UNZIP FILES**"
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```

```
}
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            "replace flowers/daisy/100080576 f52e8ee070 n.jpg? [y]es,
[n]o, [A]ll, [N]one, [r]ename: N\n"
        }
      ]
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        "# **STEP 2 Image** **Augumentation**"
      ],
      "metadata": {
        "id": "hvG8h70rrphg"
      }
    },
      "cell_type": "code",
      "source": [
        "from tensorflow.keras.preprocessing.image import
ImageDataGenerator"
      ],
      "metadata": {
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      },
      "execution count": 4,
      "outputs": []
    },
```

```
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zoom range=0.2,horizontal flip=True,vertical flip=False)"
      ],
      "metadata": {
        "id": "9vZUiTxnr0UN"
      "execution count": 5,
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      "source": [
        "test datagen=ImageDataGenerator(rescale=1./255)"
      "metadata": {
        "id": "zD7ristVr3F3"
      "execution count": 6,
      "outputs": []
    },
      "cell type": "code",
      "source": [
"x train=train datagen.flow from directory(r\"/content/drive/MyDrive/
AI IBM/
flowers\", target size=(64,64), class mode='categorical', batch size=24)"
      "metadata": {
        "colab": {
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        "id" "BiQo5zGHuHN4",
        "outputId": "d3d1e296-e74d-4e52-cce8-8d26459d10f1"
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            "Found 4317 images belonging to 5 classes.\n"
        }
      ]
    },
      "cell_type": "code".
```

```
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"x test=test 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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        "outputId": "e71a3e44-6642-4592-fa96-7af9c6edb08f"
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          "name": "stdout",
          "text": [
            "Found 4317 images belonging to 5 classes.\n"
        }
      ]
    },
      "cell type": "code",
      "source": [
        "x train.class indices"
      ],
      "metadata": {
        "colab": {
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        "id": "EaBhHHYTuv4X",
        "outputId": "8a9f62e0-7d2b-4138-c5ce-4ca16b78fbd1"
      },
      "execution count": 9,
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              "{'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3,
'tulip': 4}"
          },
          "metadata": {},
          "execution count": 9
        }
      1
```

```
},
      "cell_type": "markdown",
      "source": [
        "# **Step -3 Initializing CNN And Create Model**"
      ],
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        "id": "05cz-9q0JM s"
    },
      "cell_type": "code",
      "source": [
        "from tensorflow.keras.models import Sequential\n",
        "from tensorflow.keras.layers import
Dense, Convolution 2D, Max Pooling 2D, Flatten"
      ],
      "metadata": {
        "id": "OAUHi2otRcoC"
      },
      "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": "SzIvL8052DFR"
```

```
}
   },
      "cell_type": "code",
      "source": [
        "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": {
        "colab": {
          "base uri": "https://localhost:8080/"
        "id": "-go5E-VbSIau",
        "outputId": "1f46f35d-1950-4456-bce9-16a06053d40f"
      },
      "execution count": 15,
      "outputs": [
```

```
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                                                                   \n",
                                           Output Shape
            " Layer (type)
Param #
          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)
                                                                     0
n",
n",
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            "Trainable params: 896\n",
            "Non-trainable params: 0\n",
                                                                   \n"
        }
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        "# **4.2 Hidden Layers**"
      ],
      "metadata": {
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      "cell type": "code",
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        "model.add(Dense(300,activation='relu'))\n",
        "model.add(Dense(150,activation='relu'))"
```

```
],
      "metadata": {
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      "source": [
        "# **4.3 Output Layers**"
      "metadata": {
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    },
{
      "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',metric
s=['accuracy'])"
      ],
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        "id": "l44vMW4QShaw"
      "execution count": 18,
      "outputs": []
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      "cell_type": "code",
      "source": [
        "len(x train)"
      "metadata": {
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        "id" "Beiar4NESkL4",
```

```
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              "180"
            ]
          },
          "metadata": {},
          "execution count": 19
        }
      1
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      "cell_type": "markdown",
      "source": [
        "# **Step -5 Train the Model**"
      "metadata": {
        "id": "Y9f3ElSv3Nc6"
      }
    },
      "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)"
      ],
      "metadata": {
        "colab": {
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        },
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        "outputId": "734d2b05-c864-450f-a46f-8ce129904306"
      },
      "execution_count": 20,
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          "name": "stderr",
          "text": [
"/usr/local/lib/python3.7/dist-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"
```

```
1
      },
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        "name": "stdout",
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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",
         - 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",
         - 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".
          "180/180 [============= ] - 71s 397ms/step
- loss: 0.3892 - accuracy: 0.8631 - val loss: 0.3314 - val accuracy:
0.8826\n".
          "Epoch 18/30\n".
          - 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",
         - loss: 0.2360 - accuracy: 0.9199 - val loss: 0.1965 - val accuracy:
0.9307\n",
         "Epoch 26/30\n".
         - 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",
         "180/180 [============ ] - 73s 406ms/step
- 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",
         - loss: 0.1886 - accuracy: 0.9372 - val loss: 0.2091 - val accuracy:
0.9280\n"
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           "<keras.callbacks.History at 0x7f3e15438e50>"
        },
        "metadata": {},
        "execution count": 20
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    1
   },
    "cell type": "markdown",
    "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**"
      ],
      "metadata": {
        "id": "YAH2UVpi9RMV"
    },
      "cell type": "code",
      "source": [
        "ls"
      ],
      "metadata": {
        "colab": {
          "base_uri": "https://localhost:8080/"
        "id": "Z-co6hBAEmzg",
        "outputId": "bf8a661d-3210-4695-dcb7-48e6f365dfce"
      "execution count": 22,
      "outputs": [
        {
          "output type": "stream",
          "name": "stdout",
          "text": [
            "\u001b[0m\u001b[01;34mflowers\u001b[0m/
Flowers classification model1.h5 Flowers-Dataset.zip video.mp4\n"
        }
      ]
    },
      "cell_type": "code",
      "source": [
        "import numpy as np\n",
        "from tensorflow.keras.models import load model\n",
        "from tensorflow.keras.preprocessing import image"
      ],
```

```
"metadata": {
        "id": "mJvRRo7Vvke0"
      "execution count": 23,
      "outputs": []
    },
      "cell type": "code",
      "source": [
        "# Load the model\n",
        "model=load_model('Flowers classification model1.h5')"
      ],
      "metadata": {
        "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[v[0]]"
      ],
      "metadata": {
        "colab": {
          "base uri": "https://localhost:8080/",
          "height": 35
        },
        "id": "2rnrfMAf-AB9",
        "outputId": "c6357a8b-5163-4884-c82e-05651a65571c"
      },
      "execution_count": 38,
      "outputs": [
          "output type": "execute result",
          "data": {
            "text/plain": [
              "'sunflower'"
            "application/vnd.google.colaboratory.intrinsic+json": {
              "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"
    }
}
```