

The image consists of three vertically stacked screenshots of a Jupyter Notebook interface, all titled "DMproject1MAGelabelling".

Screenshot 1: The first screenshot shows the initial setup of the project. It includes importing necessary libraries like numpy, matplotlib.pyplot, os, tensorflow, and tensorflow.keras. It also mounts Google Drive and lists the contents of the mounted directory.

```

import numpy as np
import matplotlib.pyplot as plt
import os
import tensorflow as tf

from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, GlobalAveragePooling2D
from tensorflow.keras.applications import MobileNetV2

```

```

from google.colab import drive
drive.mount('/content/drive')

root_path = '/content/drive/MyDrive/'

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

data_path = root_path + 'DM-PROJECT/Uttrakhand_dishes'
data_path

```

```

'/content/drive/MyDrive/DM-PROJECT/Uttrakhand_dishes'

os.listdir(data_path)

['Thechhwani',
 'Dosa']

```

Screenshot 2: The second screenshot shows the creation of an ImageDataGenerator object. It defines various parameters for data augmentation, including rescale, validation split, rotation range, zoom range, width shift range, height shift range, and horizontal flip.

```

train_datagen = ImageDataGenerator(
    rescale=1./255,
    validation_split=0.25,
    rotation_range=10,
    zoom_range=0.1,
    width_shift_range=0.05,
    height_shift_range=0.05,
    horizontal_flip=True
)

```

```

training_set = train_datagen.flow_from_directory(
    data_path,
)

```

Screenshot 3: The third screenshot shows the creation of training and test datasets using the flow_from_directory method. It also prints the number of images found in each class.

```

training_set = train_datagen.flow_from_directory(
    data_path,
    target_size=(160, 160),
    batch_size=16,
    class_mode='categorical',
    subset='training'
)

test_set = train_datagen.flow_from_directory(
    data_path,
    target_size=(160, 160),
    batch_size=16,
    class_mode='categorical',
    subset='validation'
)

Found 14567 images belonging to 20 classes.
Found 4845 images belonging to 20 classes.

```

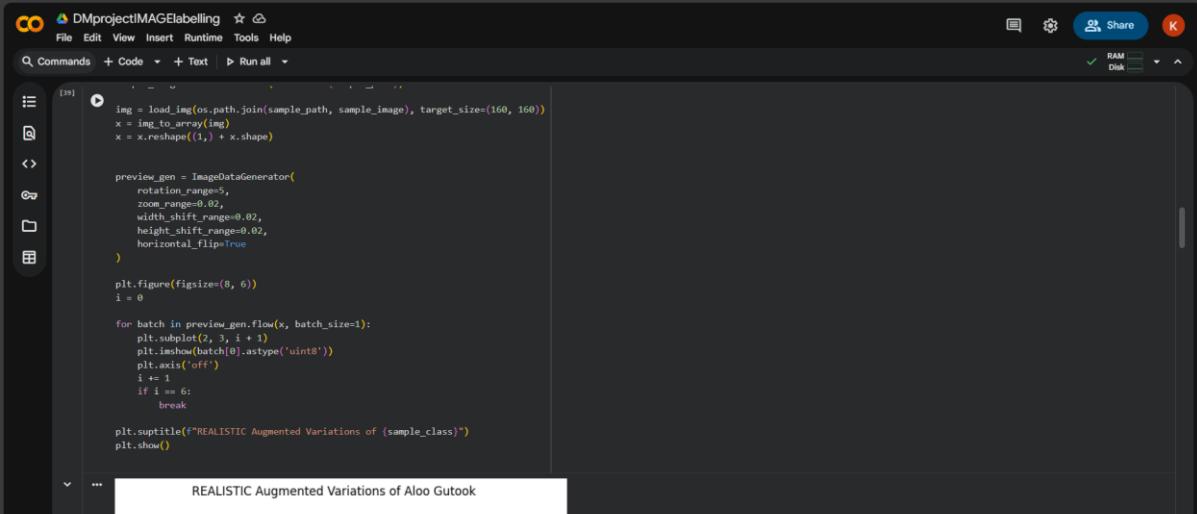
```

import matplotlib.pyplot as plt
import random
from tensorflow.keras.preprocessing.image import ImageDataGenerator, load_img, img_to_array

sample_class = random.choice(os.listdir(data_path))
sample_path = os.path.join(data_path, sample_class)
sample_image = random.choice(os.listdir(sample_path))

img = load_img(os.path.join(sample_path, sample_image), target_size=(160, 160))
x = img_to_array(img)
x = x.reshape((1,) + x.shape)

```



```
[39] img = load_img(os.path.join(sample_path, sample_image), target_size=(160, 160))
x = img_to_array(img)
x = x.reshape((1,) + x.shape)

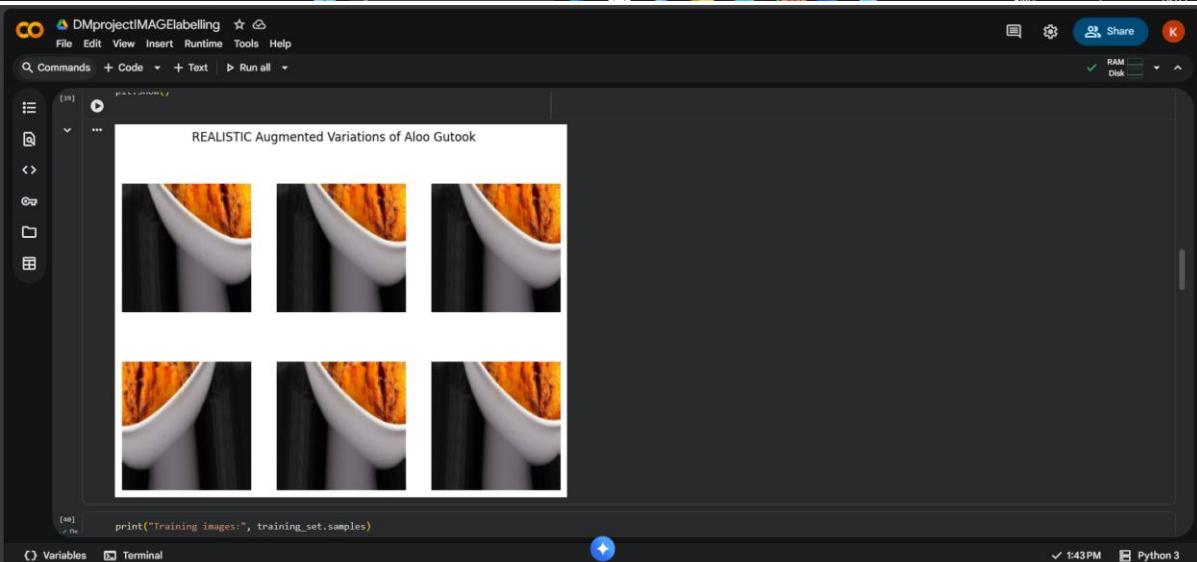
preview_gen = ImageDataGenerator(
    rotation_range=5,
    zoom_range=0.02,
    width_shift_range=0.02,
    height_shift_range=0.02,
    horizontal_flip=True
)

plt.figure(figsize=(8, 6))

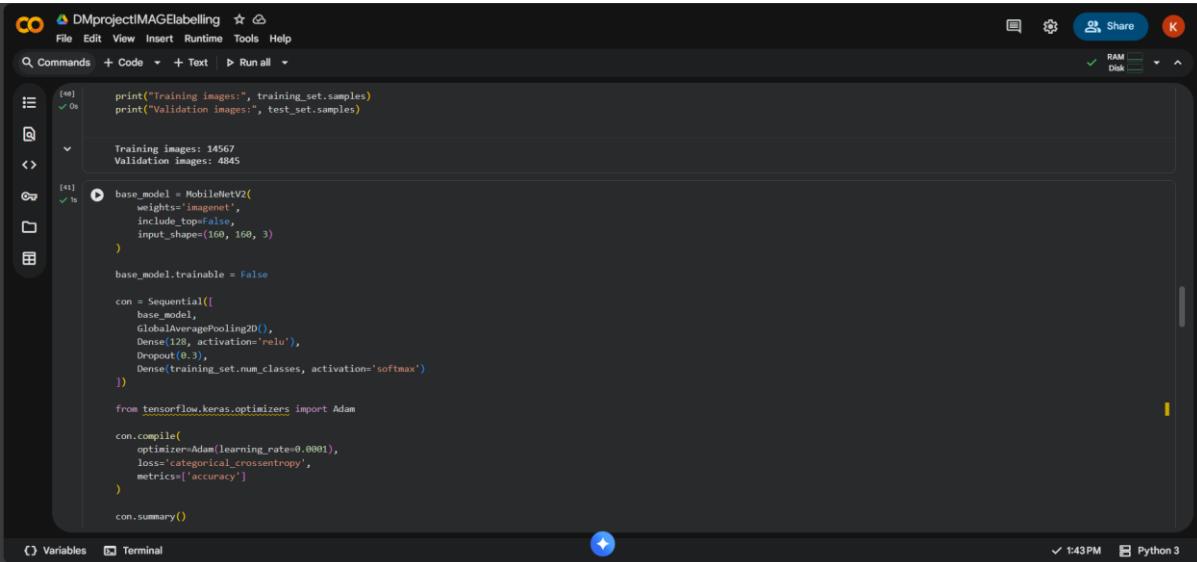
for batch in preview_gen.flow(x, batch_size=1):
    plt.subplot(2, 3, i + 1)
    plt.imshow(batch[0].astype('uint8'))
    plt.axis('off')
    i += 1
    if i == 6:
        break

plt.suptitle("REALISTIC Augmented Variations of " + sample_class)
plt.show()
```

REALISTIC Augmented Variations of Aloo Gutook



```
[40] print("Training images:", training_set.samples)
```



```
[40] Training images: 14567
Validation images: 4845

[41] base_model = MobileNetV2(
    weights='imagenet',
    include_top=False,
    input_shape=(160, 160, 3)
)

base_model.trainable = False

con = Sequential([
    base_model,
    GlobalAveragePooling2D(),
    Dense(128, activation='relu'),
    Dropout(0.3),
    Dense(training_set.num_classes, activation='softmax')
])

from tensorflow.keras.optimizers import Adam

con.compile(
    optimizer=Adam(learning_rate=0.0001),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)

con.summary()
```

```
[41] ✓ 1s
  con.summary()

Model: "sequential_3"
+-----+-----+-----+
| Layer (type) | Output Shape | Param # |
+-----+-----+-----+
| mobilenetv2_1.00_160 (Functional) | (None, 1, 1, 1280) | 2,257,964 |
| global_average_pooling2d_3 (GlobalAveragePooling2D) | (None, 1280) | 0 |
| dense_6 (Dense) | (None, 128) | 163,968 |
| dropout_3 (Dropout) | (None, 128) | 0 |
| dense_7 (Dense) | (None, 10) | 2,500 |
+-----+-----+-----+
Total params: 2,424,532 (9.25 MB)
Trainable params: 166,545 (656.58 KB)
Non-trainable params: 2,257,984 (8.61 MB)

[42] ✓ 29m
  history = con.fit(
    training_set,
    validation_data=test_set,
    epochs=12,
    steps_per_epoch=88,      # MUCH faster
    validation_steps=28,
    verbose=1
  )

... Epoch 1/12
88/88 - 281s 4s/step - accuracy: 0.7278 - loss: 0.9506 - val_accuracy: 0.8438 - val_loss: 0.7791
1:43PM Python 3

Variables Terminal
```



```
[43] ✓ 29m
  ...
  ... Epoch 1/12
  80/80 - 281s 4s/step - accuracy: 0.7278 - loss: 0.9506 - val_accuracy: 0.8438 - val_loss: 0.7791
  Epoch 2/12
  80/80 - 233s 3s/step - accuracy: 0.7651 - loss: 0.8174 - val_accuracy: 0.8531 - val_loss: 0.6354
  Epoch 3/12
  80/80 - 216s 3s/step - accuracy: 0.7947 - loss: 0.7386 - val_accuracy: 0.8969 - val_loss: 0.4907
  Epoch 4/12
  80/80 - 184s 2s/step - accuracy: 0.8475 - loss: 0.5779 - val_accuracy: 0.9187 - val_loss: 0.4126
  Epoch 5/12
  80/80 - 158s 2s/step - accuracy: 0.8385 - loss: 0.5361 - val_accuracy: 0.9344 - val_loss: 0.4027
  Epoch 6/12
  80/80 - 126s 2s/step - accuracy: 0.8934 - loss: 0.4376 - val_accuracy: 0.9344 - val_loss: 0.3529
  Epoch 7/12
  80/80 - 140s 2s/step - accuracy: 0.8907 - loss: 0.4281 - val_accuracy: 0.9594 - val_loss: 0.2339
  Epoch 8/12
  80/80 - 98s 1s/step - accuracy: 0.9042 - loss: 0.3566 - val_accuracy: 0.9844 - val_loss: 0.1960
  Epoch 9/12
  80/80 - 110s 1s/step - accuracy: 0.9361 - loss: 0.2872 - val_accuracy: 0.9656 - val_loss: 0.1765
  Epoch 10/12
  80/80 - 95s 1s/step - accuracy: 0.9337 - loss: 0.2825 - val_accuracy: 0.9781 - val_loss: 0.1832
  Epoch 11/12
  80/80 - 86s 1s/step - accuracy: 0.9375 - loss: 0.2724 - val_accuracy: 0.9812 - val_loss: 0.1647
  Epoch 12/12
  31/80 - 9s 196ms/step - accuracy: 0.9257 - loss: 0.2739/usr/local/lib/python3.12/dist-packages/keras/src/trainers/epoch_iterator.py:116: UserWarning: Your input ran out of data
  self._interrupted_warning()
  37s 459ms/step - accuracy: 0.9288 - loss: 0.2645 - val_accuracy: 0.9875 - val_loss: 0.1166
1:43PM Python 3

Variables Terminal
```



```
[44] ✓ 6m
  test_loss, test_accuracy = con.evaluate(test_set)
  print("Test Accuracy:", test_accuracy)

1:43PM Python 3
```



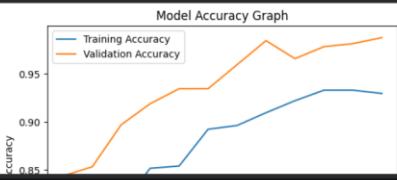
```
[45] ✓ 6m
  test_loss, test_accuracy = con.evaluate(test_set)
  print("Test Accuracy:", test_accuracy)

  303/303 - 392s 1s/step - accuracy: 0.9778 - loss: 0.1544

[46] ✓ 0s
  plt.plot(history.history['accuracy'])
  plt.plot(history.history['val_accuracy'])

  plt.xlabel('Epochs')
  plt.ylabel('Accuracy')
  plt.title('Model Accuracy Graph')
  plt.legend(['Training Accuracy', 'Validation Accuracy'])

  plt.show()
```



Model Accuracy Graph

Epoch	Training Accuracy	Validation Accuracy
1	~0.85	~0.85
2	~0.85	~0.88
3	~0.88	~0.90
4	~0.90	~0.92
5	~0.90	~0.93
6	~0.91	~0.94
7	~0.91	~0.95
8	~0.92	~0.96
9	~0.92	~0.97
10	~0.92	~0.98
11	~0.92	~0.98
12	~0.92	~0.98

1:43PM Python 3

