

IMAGE RECOGNITION WITH IBM CLOUD

VISUAL RECOGNITION

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PHASE 3

PROGRAM FOR IMAGE RECOGNITION

GOAL:

1. The main goal and purpose of the program is to perform image classification on a user-provided image using a pre-trained InceptionV3 model.

2. The program combines image processing, a pre-trained deep learning model, and visualization techniques to classify an image provided by the user through a URL. It then displays the top predictions and the image itself.

Here are the steps and objectives :

1. User Interaction:

The program prompts the user to enter a URL of an image. This is done using the `input()` function which waits for user input.

2. Image Retrieval:

The `requests.get(url)` function sends a GET request to the provided URL. It retrieves the image data from the internet.

3. Image Processing:

`'BytesIO(response.content)'` creates a byte stream from the image content obtained in the response. This stream-like object allows us to treat the binary image data as a file.

`'Image.open(...)'` opens the image from the byte stream.

`'.convert("RGB")'` converts the image to RGB mode, which is a standard color mode for images.

4. Image Resizing:

`img.resize((299, 299))` resizes the image to a square of 299x299 pixels.

This is a requirement for the InceptionV3 model.

5. Image to Array:

`'img_to_array(img)'` converts the image to a numpy array. This format is compatible with the InceptionV3 model.

6. Preprocessing for Model Input:

`'preprocess_input(img)'` applies necessary preprocessing to the image array, like mean subtraction, scaling, etc. This prepares the image for input into the neural network.

7. Tensor Conversion:

`'tf.convert_to_tensor(img)'` converts the numpy array to a TensorFlow tensor. TensorFlow uses tensors as the fundamental data structure for computations.

8. Resizing for Model Input:

`'tf.image.resize(img, (299, 299))'` resizes the image tensor to match the input size expected by the InceptionV3 model.

9. Adding Batch Dimension:

`'img[tf.newaxis, ...]'` adds an extra dimension at the beginning to represent a batch of images. This is needed for compatibility with the model.

10. Loading Pre-trained Model:

`'InceptionV3(weights='imagenet')'` loads the InceptionV3 model pre-trained on the ImageNet dataset.

11. Model Prediction:

`'model.predict(img)'` passes the preprocessed image through the neural network to get predictions.

12. Prediction Decoding:

`'decode_predictions(predictions)'` translates the raw class probabilities into human-readable labels.

13. Display Top Predictions:

The program prints out the top 5 predicted labels along with their associated probabilities.

14. Display Image:

`'plt.imshow(...)'` displays the downloaded image using matplotlib. The image is shown without axis information.

15. Display the Result:

The program outputs the predictions and displays the image.

PROGRAM :

```
from PIL import Image
import requests
from io import BytesIO
import tensorflow as tf
from tensorflow.keras.applications.inception_v3 import InceptionV3,
preprocess_input, decode_predictions
import matplotlib.pyplot as plt
url = input("Please enter the URL of the image: ")
response = requests.get(url)
img = Image.open(BytesIO(response.content)).convert("RGB")
img = img.resize((299, 299))
img = tf.keras.preprocessing.image.img_to_array(img)
img = tf.keras.applications.inception_v3.preprocess_input(img)
img = tf.convert_to_tensor(img)
img = tf.image.resize(img, (299, 299))
img = img[tf.newaxis, ...]
model = InceptionV3(weights='imagenet')
predictions = model.predict(img)
decoded_predictions = decode_predictions(predictions)[0]
print("Top predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions[:5]):
    print(f'{i + 1}: {label} ({score:.2f})')
plt.imshow(Image.open(BytesIO(response.content)))
plt.axis('off')
plt.show()
```