IMAGE RECOGNITION WITH IBM CLOUD VISUAL RECOGNITION

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PHASE 4

GOAL:

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

STEP 20: I have given urls of few images as input in the runtime, and got the output successfully. It will give the top five predictions about the image as captions.

Sample output 1:

Sample output 2:

Sample output 3:

CONCLUSION:

- In conclusion, the Python program developed serves as a powerful tool for image classification using deep learning techniques. Leveraging the InceptionV3 model and TensorFlow, the program provides accurate predictions for a wide range of objects present in images. The user-friendly interface allows for dynamic input via URLs, making it a versatile solution for real-time image analysis.
- Furthermore, the seamless integration with IBM Cloud enhances the program's scalability and accessibility. By deploying the application on the IBM Cloud platform, it gains the advantage of robust infrastructure and can be easily accessed from anywhere. This not only ensures reliability but also lays the foundation for potential future enhancements and collaborations within the IBM Cloud ecosystem.
- In summary, the combination of advanced image classification capabilities and the IBM Cloud platform empowers this program to be a valuable asset for various applications, from e-commerce to content moderation and beyond. It represents a significant step towards harnessing the potential of AI and cloud computing for impactful solutions.