Fog Computing-Based Object Detection Project

Project Overview:

Our project is designed to address the need for real-time object detection in images within a distributed fog computing system. This system enables efficient and parallel processing of incoming images from multiple clients while ensuring fast response times.

Project Objectives:

- Real-time object detection
- Utilizing fog computing
- Scalability and parallel processing

Our project is designed to address the need for real-time object detection in images within a distributed fog computing system. This system enables efficient and parallel processing of incoming images from multiple clients while ensuring fast response times.

Technologies Used:

- Python
- Flask-SocketIO
- Ultralytics YOLOv8 model for objects detection
- Socket programming
- Threading and parallelism
- Networking and communication

How It Works?

Step 1 : Image Upload

Clients, using Flutter applications, upload images to our central server. These
images can be of various types and sizes, and clients can upload them
concurrently.

Step 2 : Server

- The server acts as the central hub of our system. It listens for incoming connections from clients, and upon receiving an image, it handles the following tasks:
 - 1. Receives and stores the image.
 - 2. Distributes the image to multiple fog nodes for object detection.
 - 3. Sends the prediction result back to the client.

Step 3: Fog Nodes

- Our fog nodes are distributed computing units each listening on a socket, and each running the powerful YOLOv8 model for real-time object detection.
 These nodes are continuously available and responsive to new tasks.
- When a server sends an image to a fog node, it processes the image, detecting objects within it, and then returns the results to the server.

Step 4: Parallel Processing

 One of the key strengths of our system is parallel processing. As multiple clients upload images, the server can send these images to multiple fog nodes simultaneously. This means that different images can be processed at the same time on different fog nodes.

Step 5: Results

 As fog nodes complete object detection tasks, they send the results back to the server. The server accumulates the results as they arrive.

Step 6: Feedback to Clients

• Finally, the server sends the object detection results back to the clients who uploaded the images. Clients receive real-time feedback, allowing them to take further actions based on the detected objects.

Implementation Overview:

1. Server Code Overview:

The initial section of the server code imports the necessary Python modules and libraries:

```
import os
import socket
import concurrent.futures
import threading
import pickle
import datetime
```

- os (Operating System Interface): Provides functions for interacting with the operating system, including file and directory operations.
- socket (Socket Programming): Enables network communication by creating and managing network sockets for sending and receiving data.
- concurrent.futures: Part of the Python standard library and used for working with asynchronous tasks, such as thread and process pools.
- threading (Threading): Provides a way to create and manage threads, allowing for concurrent execution of code within a single process.

- pickle (Object Serialization): A module for serializing and deserializing Python objects, used for data storage and communication between processes.
- datetime (Date and Time Operations): Offers classes and functions to work with dates, times, and time intervals, including formatting and parsing date and time values.

This part of the code sets up various global variables and configuration parameters:

```
9  UPLOADS_FOLDER = 'uploads/' # Specify the folder
10  # Create a dictionary to store locks for each image
11  image_locks = {}
12  image_list = []
13  node_availability = [True]*2
14  imagesSent={}
15  predictions = {}
16  imageConnection={}
17  counter=0
18  counter_lock = threading.Lock() # Create a lock in node_availability_locks = threading.Lock()
```

- UPLOADS FOLDER defines the folder where received images are stored.
- A dictionary, image_locks, is used to store locks for each image, ensuring that images are processed and deleted in a thread-safe manner.
- image list is a list to maintain received images waiting to be processed.
- node_availability is a list to track the availability of fog nodes.
- *imagesSent* is used to associate images with the fog node to which they are sent.
- *predictions* stores the prediction results for images.
- imageConnection associates each image with the connection from which it was received.
- count is a global variable used to avoid image storage with the same name by two
 threads.
- counter_lock is the lock used to let access only by one thread.
- node_availability_locks is the lock used to avoid two threads uses the same node at the same time.

This part of the code specify the different functions implemented :

```
def run_app():
    with socket.socket(socket.AF_INET, socket.SOCK_STREAM) as server_socket:
        server_socket.bind(('192.168.1.18', 100))
        server_socket.listen()

print(f"Server listening on {'192.168.1.18'}:{100}")

while True:
        connection, address = server_socket.accept()
        threading.Thread(target=handle_connection, args=(connection, address)).start[)
```

- This function is responsible for listening to client connections and launching a new thread to handle each incoming connection.
- It binds the server socket to a specified IP address and port and listens for incoming connections.

```
def handle_connection(connection, address):
   print(f"Connected to {address}")
   image_data = b''
   receiving = False
       data = connection.recv(1024)
       if not data:
           break
       if b'ImageStart' in data:
           receiving = True
           data = data.replace(b'ImageStart', b'')
       if b'ImageEnd' in data:
           receiving = False
           data = data.replace(b'ImageEnd', b'')
           image_data += data
           timestamp = datetime.datetime.now().strftime("%Y%m%d_%H%M%S")
           global counter
           image_path = os.path.join(UPLOADS_FOLDER, f'received_image_{timestamp}_{counter}.jpg')
           with counter_lock:
               counter += 1
           with open(image_path, 'wb') as image_file:
               image_file.write(image_data)
            image_list.append(image_path)
            imageConnection[image_path]=connection
            image_data = b''
        if receiving:
            image_data += data
```

- This function is executed in a separate thread for each client connection.
- It receives image data from clients and saves it to a file with the use of the counter_lock to make sure
 of a unique image name.
- maintains the image_list for images that are ready for processing and updates imageConnection to keep track of which connection is associated with each image.

```
def process_images():
   num nodes = 2
   node_addresses = [("192.168.1.18", 200), ("192.168.1.18", 300)]
   with concurrent.futures.ThreadPoolExecutor(max_workers=num_nodes) as executor:
       futures = []
       print(image_list)
            for j in range(num_nodes):
                if not image_list:
                    break
                with node_availability_locks:
                    if node availability[j]:
                       image_path = image_list.pop(0)
                       imagesSent[image_path] = j
                       image_locks[image_path] = threading.Lock()
                        node_availability[j] = False
                        future = executor.submit(send_image_to_node, node_addresses[j], image_path)
                        futures.append((future, image_path))
            for future, image_path in futures:
                if future.done():
                   response = future.result()
                    node_index = imagesSent[image_path]
                   print(f"Objects detected by Node {node_index+1}: {response}")
                   futures.remove((future, image_path))
                   node_availability[node_index] = True
                   predictions[image_path] = response
            # Acquire the lock before deleting the image
                   with image_locks[image_path]:
                       if os.path.exists(image_path):
                           os.remove(image_path)
                # Send acknowledgment
                       if image_path in imageConnection:
                            imageConnection[image_path].send(bytes('\n'.join(response), 'utf-8'))
                            del imageConnection[image_path]
```

- This function manages the distribution of images to available fog nodes using a thread pool.
- It sends images to nodes as they become available and we use node_availability_locks lock to avoid
 two threads using the same node, and then waits for results. results are stored in futures List: which is
 a container that holds future objects, each representing an ongoing asynchronous task for sending
 images to fog nodes and receiving predictions.
- It uses the image_locks dictionary to ensure that images are processed and deleted in a thread-safe manner.
- When results are received, As futures in the futures list complete, their results are received by checking
 their status, allowing the system to collect and process the predictions made by fog nodes. and then
 predictions are sent back to the client and the image is deleted from the server

```
def send_image_to_node(node_address, image_data):
    with socket.socket(socket.AF_INET, socket.SOCK_STREAM) as s:
        s.connect(node_address)
        s.sendall(b'ImageStart') # Start the image transmission
        s.sendall(image_data) # Send the image data
        s.sendall(b'ImageEnd') # End the image transmission
        response = s.recv(1024)
        return pickle.loads(response)
```

- This function is responsible for establishing a socket connection with a fog node and sending an image path for processing.
- It sends the image data to the fog node, waits for a response, and returns the prediction results.

- In the main block, the script starts the run_app function in a separate thread to listen for client connections.
- It then calls the process_images function to manage image distribution and processing.

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2- Node Code Overview:

```
import socket
import pickle
from ultralytics import YOLO
```

• Utilize the Ultralytics YOLOv8 model for object detection. YOLO (You Only Look Once) is a popular deep learning model for real-time object detection in images.

- This function is responsible for performing object detection using the YOLOv8 model.
- It takes image data as input and returns a list of detected objects found in the image.

```
node_function(node_address):
with socket.socket(socket.AF_INET, socket.SOCK_STREAM) as s:
     s.bind(node_address)
    s.listen()
print("Listening on address:", node_address[0], "PORT:", node_address[1])
         conn, addr = s.accept()
              image_data = b'
receiving = False
                   data = conn.recv(1024)
if not data:
                   if b'ImageStart' in data:
    receiving = True
                        data = data.replace(b'ImageStart', b'')
                    if b'ImageEnd' in data:
    receiving = False
                        data = data.replace(b'ImageEnd', b'')
                        image_data += data
                        # Convert received image data to a PIL Image object
image = Image.open(BytesIO(image_data))
                        detected_objects = object_detection(image)
                        conn.sendall(pickle.dumps(detected_objects))
                        break
                    if receiving:
                         image_data += data
```

This function encapsulates the main functionality of the fog node.

 It listens for incoming connections on a specific IP address and port, receives image data from the server, and performs object detection using the YOLOv8 model.

- In the main block of the fog node code, we specify the node's IP address and port to listen for connections.
- Then we start the **node_function**, which listens for incoming image data from the server, processes it, and sends the detected objects back to the server.

3- Flutter Code Overview:

```
Future getImage(bool isCamera) async {
    final pickedFile = await picker.pickImage(
        source: isCamera ? ImageSource.camera : ImageSource.gallery);

setState(() {
    if (pickedFile != null) {
        image = File(pickedFile.path);
    } else {
        print('No image selected.');
    }
};

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```

- This function is responsible for capturing or selecting an image. It uses the ImagePicker plugin to access the device's camera or image gallery based on the *isCamera* parameter.
- The selected image is stored in the image variable, and the user interface is updated accordingly.

```
Future sendImageToServer() async {
         if (image == null) return;
         try {
           socket = await Socket.connect('192.168.1.18', 100);
           listenForMessages();
           print(e.toString());
         if (image == null) {
           print('No image selected');
           return;
         final File imageFile = File(image!.path);
         final List<int> imageBytes = await imageFile.readAsBytes();
         socket?.add(utf8.encode('ImageStart'));
         socket?.add(imageBytes);
         socket?.add(utf8.encode('ImageEnd'));
56
       3
```

- This function is used to send the selected image to the server. It establishes a socket connection to the server at the specified IP address and port.
- It converts the image file into bytes, adds delimiters ('ImageStart' and 'ImageEnd'), and sends the image data to the server through the socket connection.

```
void listenForMessages() {

socket?.listen((List<int> event) {

final message = String.fromCharCodes(event);

setState(() {

acknowledgmentMessage = message;
});

4 });

5 }
```

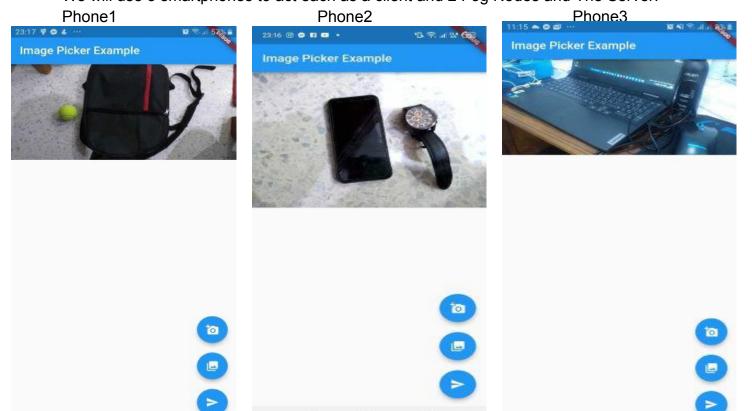
- This function sets up a listener on the socket connection. It listens for incoming messages from the server
- When a message is received, it converts the byte data into a string and updates the
 acknowledgmentMessage variable, which can be displayed to the user to show the acknowledgment or
 response from the server.

Benefits:

- Real-time object detection: Clients receive results in near real-time.
- **Scalability**: Our system can efficiently handle a large number of clients and images.
- *Efficient parallel processing*: Multiple images can be processed simultaneously on multiple fog nodes, ensuring a responsive system.
- Optimized resource usage: Images are removed from the server once they are successfully predicted and sent back to the client, ensuring efficient use of server storage resources.

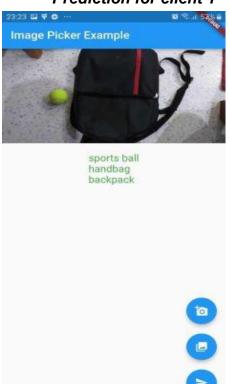
Simulation:

We will use 3 smartphones to act each as a client and 2 Fog Nodes and The Server:



- Like we see each client has captured an image and will now send it to the server by pressing the send button and it will wait for the prediction.
- Image Processing will happen simultaneously.
- → After Processing is Complete

Prediction for client 1



Prediction for client 2



Prediction for client3



Conclusion: Like we can see for each client: Image objects are detected by the Fog Nodes and sent to the client.

Time Gain:

In the example of simulation when there are 3 clients and 2 Fog nodes:

- -Using parallel Processing: Time to process the three images is 8 seconds.
- -Using One node(no parallel processing): Time to process the three images is 15 seconds.

Video Explaining the work of Server and Fog Nodes:

Watch The video Sent separately.