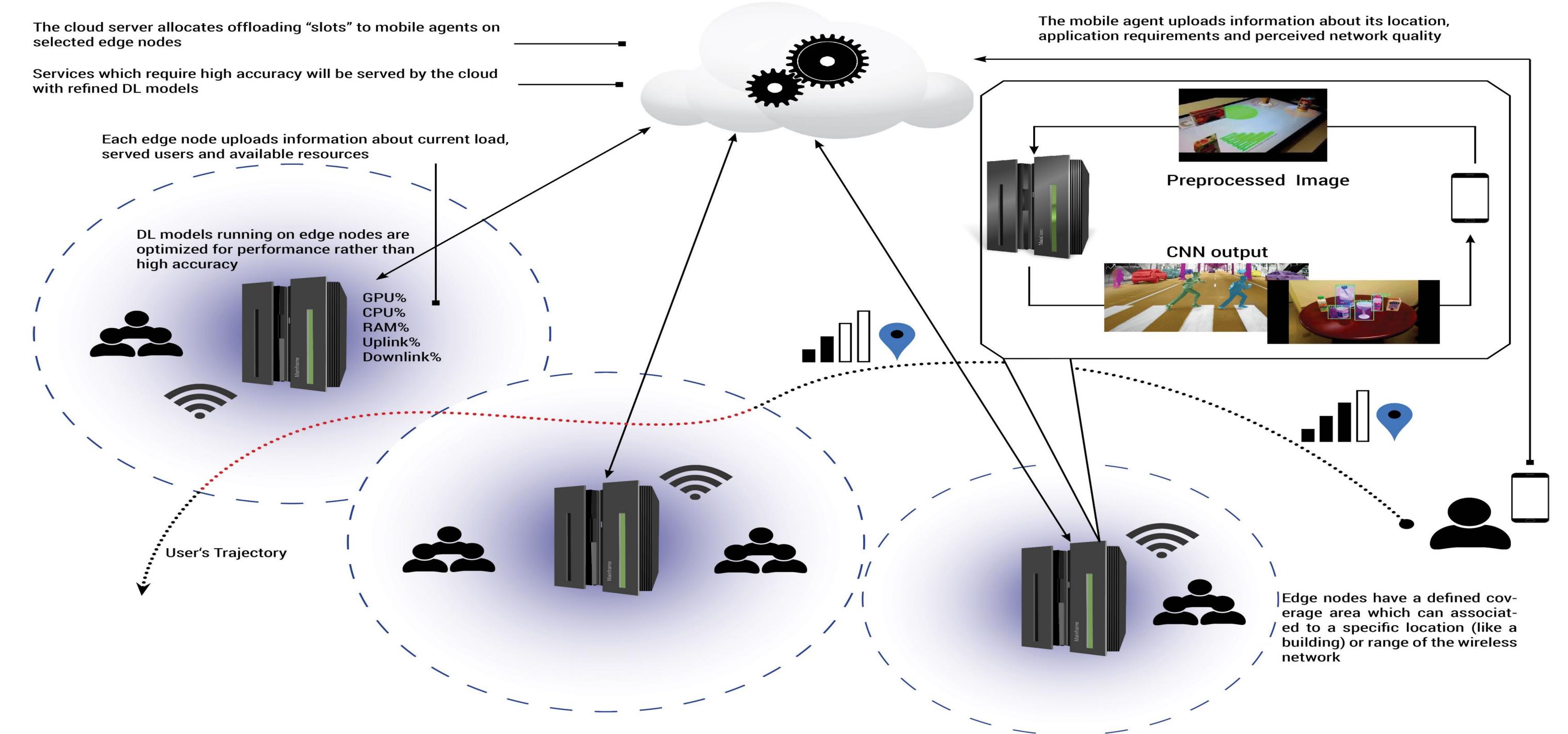
mDOCS: Mobile Deep-Neural-Networks Offloading Edge-Cloud Scheduler Vittorio Cozzolino, Leonardo Tonetto {vittorio.cozzolino; tonetto}@in.tum.de Chair for Connected Mobility



Overview

- Computer Vision powers every mobile Augmented
 Reality (AR) application available on mobile devices
- Machine Learning for mobile AR applications increases immersiveness (e.g. Pokemon GO, PinAR):
 - SLAM (sparse or dense), SIFT
 Image classification

Our Approach

- Exploit nearby infrastructure by offloading computation to hardware:
 - In the Cloud
 - At the **Edge**
- Optimize for network condition, mobility, available hardware resource, required accuracy, etc.

- Object recognition
- Image segmentation
- Panoptic segmentation
- Such models either run directly on the phone (downgraded version) or make use of cloud resources

Problem

- Smartphones are either not powerful enough or the applications too demanding in terms of energy drain
- Support old smartphone which do not have GPU acceleration, AI chips or octa-core CPUs
- Mobile Data Traffic is often expensive
- AR applications are highly demanding on data traffic
- Deep Learning models on mobile devices deliver reduced accuracy due to simplifying optimization steps like pruning, layers fusion, quantization etc.

Centralized decision making for better resource allocation

Edge-Cloud Scheduling

- Centralized, cloud-based (overseer) which decides:
 - How/which edge node should serve a mobile agent in proximity following our heuristic
 - when to offload at the edge, when to use the cloud or if to just run everything locally
- Edge nodes and mobile agents periodically send updates to notify their current status (e.g. position, current load)
- Two macro-sets of parameters influence:
 - Network and Hardware constraints
 - Human Mobility constraints

Network and Hardware Constraints



- High performance mobile AR experience in a multitenant scenario regardless of mobile devices hardware
- Provide a scheduling algorithm based on a multivariate, multi-constrained heuristic with a trifecta target function:
 - maximization of throughput (FPS)
 - minimization of latency
 - extension of battery life
- Proactively allocate resources as users move in-andout from edge nodes' coverage area
- Optimize inter-node scheduling for efficient GPU sharing

- Available edge nodes resources: GPU, CPU, RAM
- Network conditions and fluctuations
- Application requirements: accuracy, recall, speed
- Impact of multi-tenancy
- Performance impact of sharing a GPU across multiple users
- Heterogeneous devices: Jetson Nano, TX2, server GPUs
 High-precision networks vs quantized optimized networks

Human Mobility Constraints

- User's position and network coverage (WiFi/Mobile Network)
- Trajectory prediction for preemptive resource allocations
- Mobility and Data Traffic must show strong correlation*