MicroOpt: Model-driven Slice Resource **Optimization in 5G and Beyond Networks** Muhammad Sulaiman, Mahdieh Ahmadi, Bo Sun, Mohammad A. Salahuddin,

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RESEARCH PROBLEM

BACKGROUND

- **5G Network:** Comprises Virtual Network Functions (VNFs) within the Radio Access Network (RAN), Transport, and Core
- Network Slicing: Involves creating isolated, virtual networks tailored to specific use-cases (e.g., eMBB slices tailored of VR/XR, URLLC slice for remote driving, telesurgery and mMTC for IOT applications)
- Slice QoS & SLA: Service Level Agreements (SLAs) define the minimum Quality of Service (QoS) 5G slices must receive
- Resource Allocation vs. QoS: The amount of resources allocated to VNFs directly impacts the slice's QoS

MOTIVATION

- Dynamic Traffic Patterns: Slices experience timevarying traffic
- Resource Efficiency: Resource allocation for peak traffic leads to resource over-provisioning

PROBLEM STATEMENT & CHALLENGES

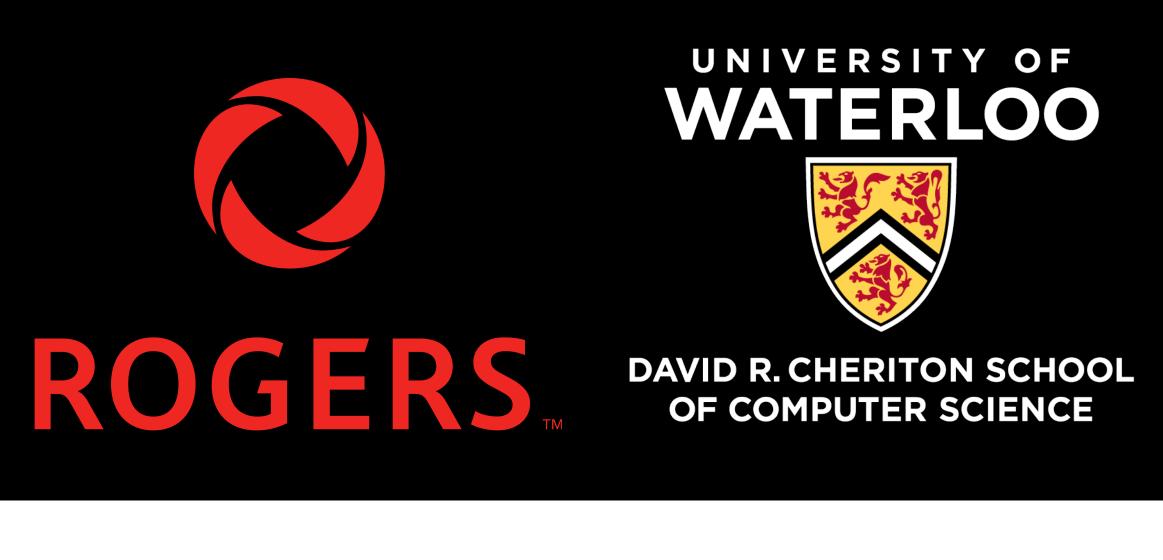
- Resource Optimization: Minimize resource allocation while meeting QoS requirements
 - **Network Complexity:** Accurately modeling complex, multi-VNF slices
 - **Real-Time Adaptation:** Fast, efficient resource adjustment to varying slice traffic

SOLUTION

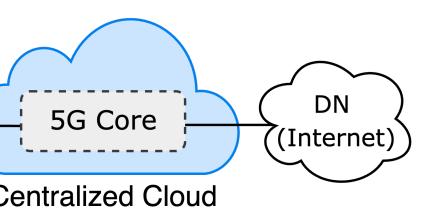
- ML-Based VNF & Slice Modeling: Leveraging machine learning for modeling VNFs that can be composed to create end-to-end 5G slice models
- Resource Allocation Algorithm: Using Lagraingian primal-dual algorithm coupled with gradient descent for fast, near-optimal resource scaling under QoS constraints

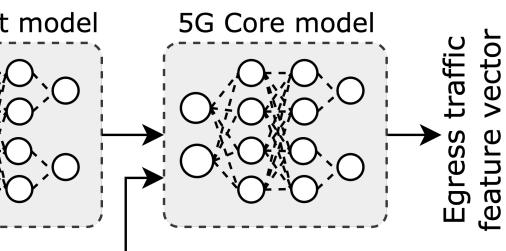
ML-BASED VNF & SLICE MODELING VNF MODELING Input/Output: Predicts the egress traffic feature vector given input traffic trace (i.e., pcap) and VNF configuration & resource allocation • **Dataset:** Uses data from in-lab 5G testbed and partner network operator's real network for training VNF models **Differentiability:** Leverages reparameterization trick to maintain VNF model differentiability **Composability:** Allows stacking VNF models to form end-to-end slice model **SLICE MODELING Construction:** Composed using individual VNF models Backhaul _ _ _ _ _ _ _ _ _ _ _ _ DN 5G Core 5G RAN transport ---------**Centralized Cloud Distributed Cloud** Transport Net 5G RAN mode Fransport mode Pkt stream VNF config / Resource allocation 0.8 CDF \mathbf{O} — Ground Truth Predicted Delay (ms) Throughput (Mbps)

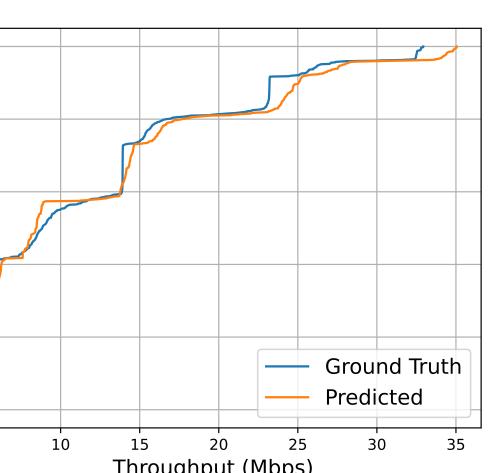
Fig: Ground truth vs. predicted throughput and delay for end-to-end slice model











RESOURCE ALLOCATION ALGORITHM

ALGORITHM

- with Gradient descent
 - **Relaxed Lagrangian formulation:**

- (i.e., $QoS_{required} QoS_{predicted}$)

Algorithm 1 MicroOpt Algorithm $\tau_{1,max}, \tau_{2,max}, \alpha_1, \alpha_2, \alpha_3, \epsilon_1, \epsilon_2$ 2: while $\frac{\text{UB}-\text{LB}}{\text{UB}} > \epsilon_1$ or $\tau_1 < \tau_{1,max}$ do $\mathbf{r} \leftarrow [\mathbf{r} - \alpha_1 \nabla_r \hat{\mathcal{L}}]^+$ $\tau_2 \leftarrow \tau_2 + 1$ end while $\lambda_s \leftarrow [\lambda_s + \alpha_2(\beta_i^s - \beta_{thresh}^s)]^+, \forall s$ $\mu_k \leftarrow [\mu_k + \alpha_3(\sum_{s \in \mathbf{S}} r^{s,k} - R^k)]^+, \forall k$ $LB = max(LB, \mathcal{L}(\mathbf{r}, \boldsymbol{\mu}, \boldsymbol{\lambda}))$ 10: $UB = min(UB, \sum_{s \in S} \eta^{\mathsf{T}} \mathbf{r}^s)$ $\tau_1 \leftarrow \tau_1 + 1$ 13: end while

14: return r

RESULTS

- **Resource saving:** SOTA and peak resource allocation
- compared to previous SOTA

Lagrangian primal dual algorithm coupled

 $\mathcal{L} = \mathbf{r} + \boldsymbol{\lambda} \cdot (QoS_{required} - QoS_{predicted})$

Outer loop: Updates Lagrangian variables (λ) based on QoS constraint violation

Inner loop: Utilizes gradient descent to minimize the relaxed Lagrangian, with gradients from the differentiable slice model

Input: Traffic x_i^s , Slice Model $f_{OoS}^s(x_i^s, r_i^s)$, QoS threshold q_{thresh}^{s} , QoS degradation threshold β_{thresh}^{s} , **Output:** Optimal resource allocation vector r_i^s 1: Initialize $\lambda, \mu, LB = 0, UB = \infty, \tau_1 = 0, \tau_2 = 0$ $\mathbf{r} \leftarrow \text{Initialization}(\mathbf{x}_i^s, f_{QoS}(\mathbf{x}_i^s, \mathbf{r}))$ while $|\nabla_r \hat{\mathcal{L}}| > \epsilon_2$ or $\tau_2 < \tau_{2,max}$ do 0 ter

<u>14.60% and 20.74% improvement over previous</u>

Significantly faster resource scaling: <u>2-3 orders of magnitude</u> faster resource scaling