

PORTEND: A Joint Performance Model for Partitioned Early-Exiting DNNs

Maryam Ebrahimi

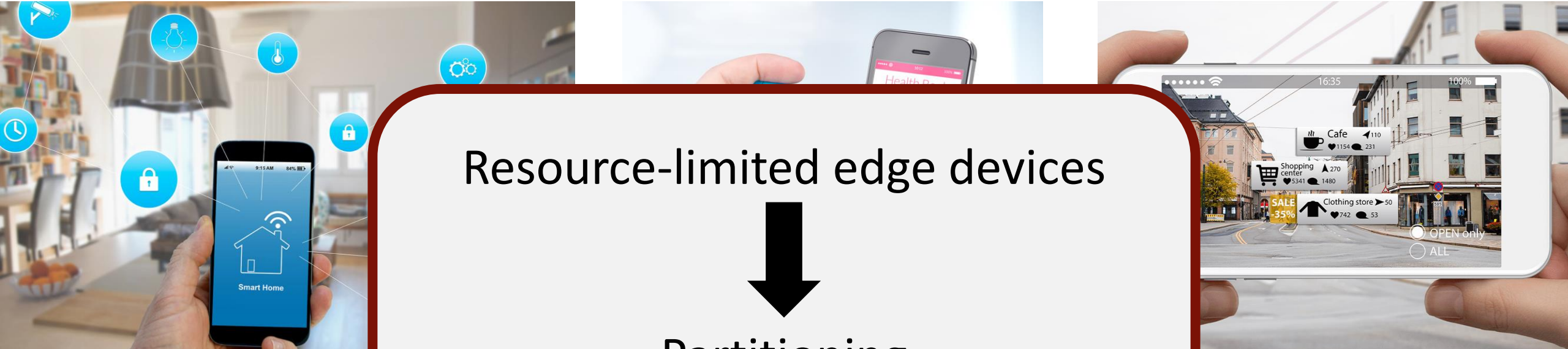
Moshe Gabel

Alexandre da Silva Veith

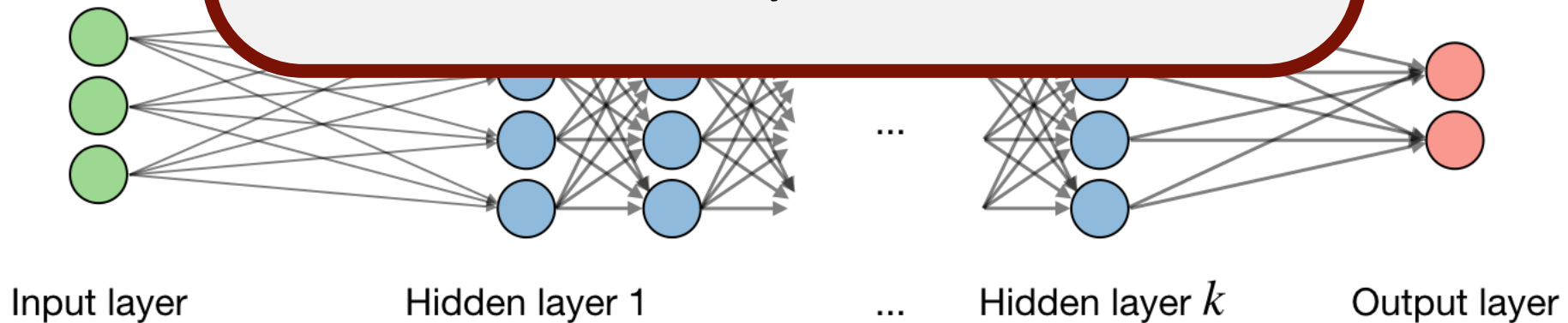
Eyal de Lara



Edge and mobile applications, need fast DNN inference

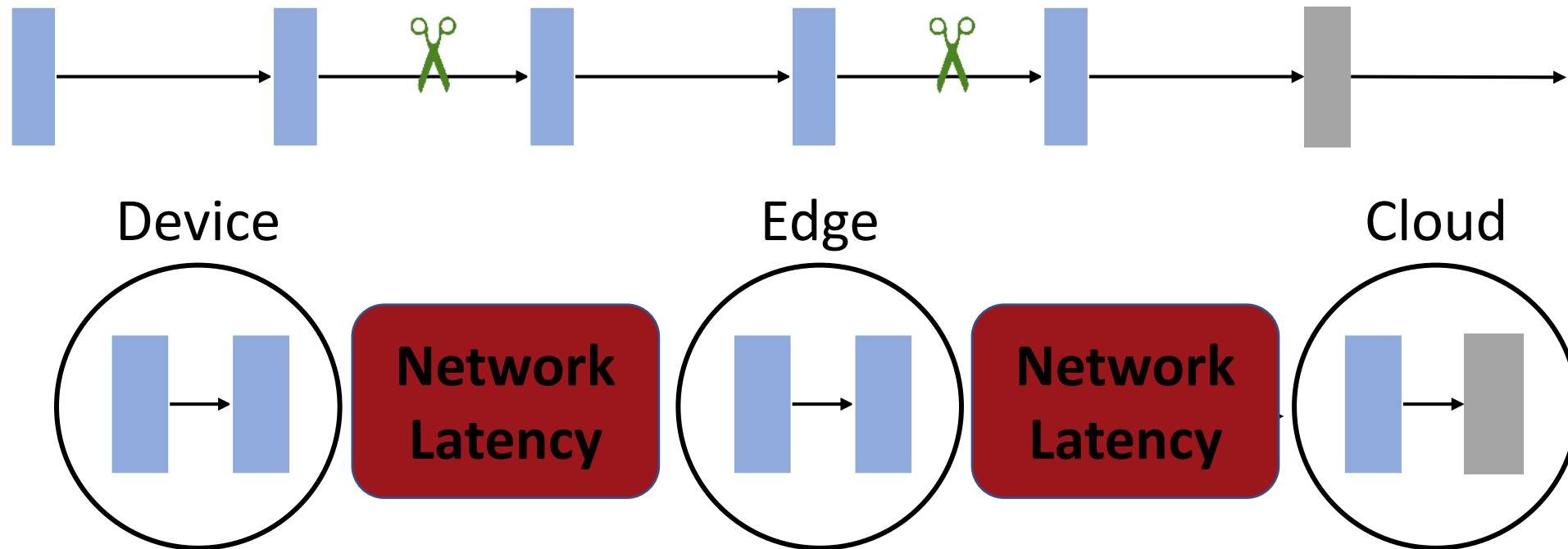


Resource-limited edge devices
↓
Partitioning
Early Exit



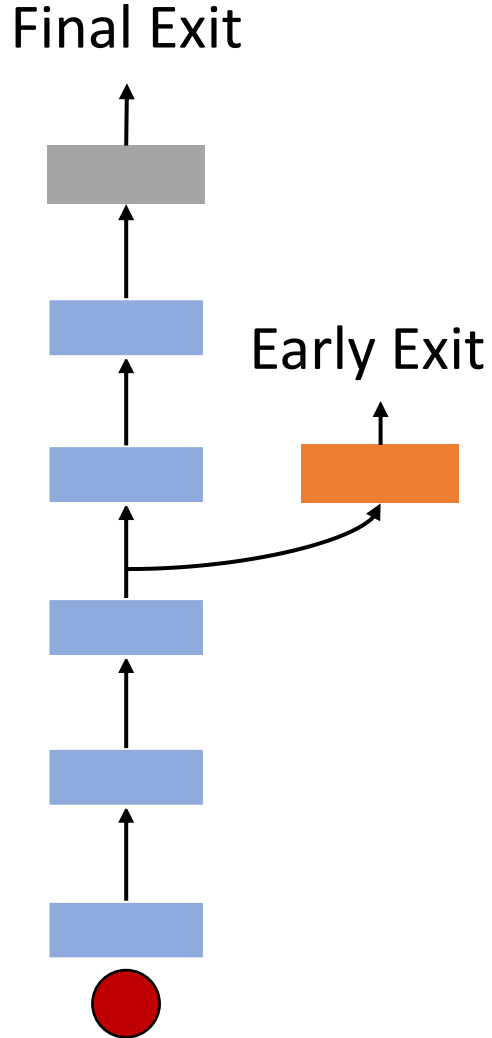
Background: Partitioning

Split pre-trained DNN across multiple servers



This approach trades-off computation time for network latency

Background: Early Exit



Add early exit branches as auxiliary classifiers

Use entropy to measure the confidence of the prediction

$\text{Entropy}(\text{output}) < \text{Threshold} \longrightarrow \text{Exit}$

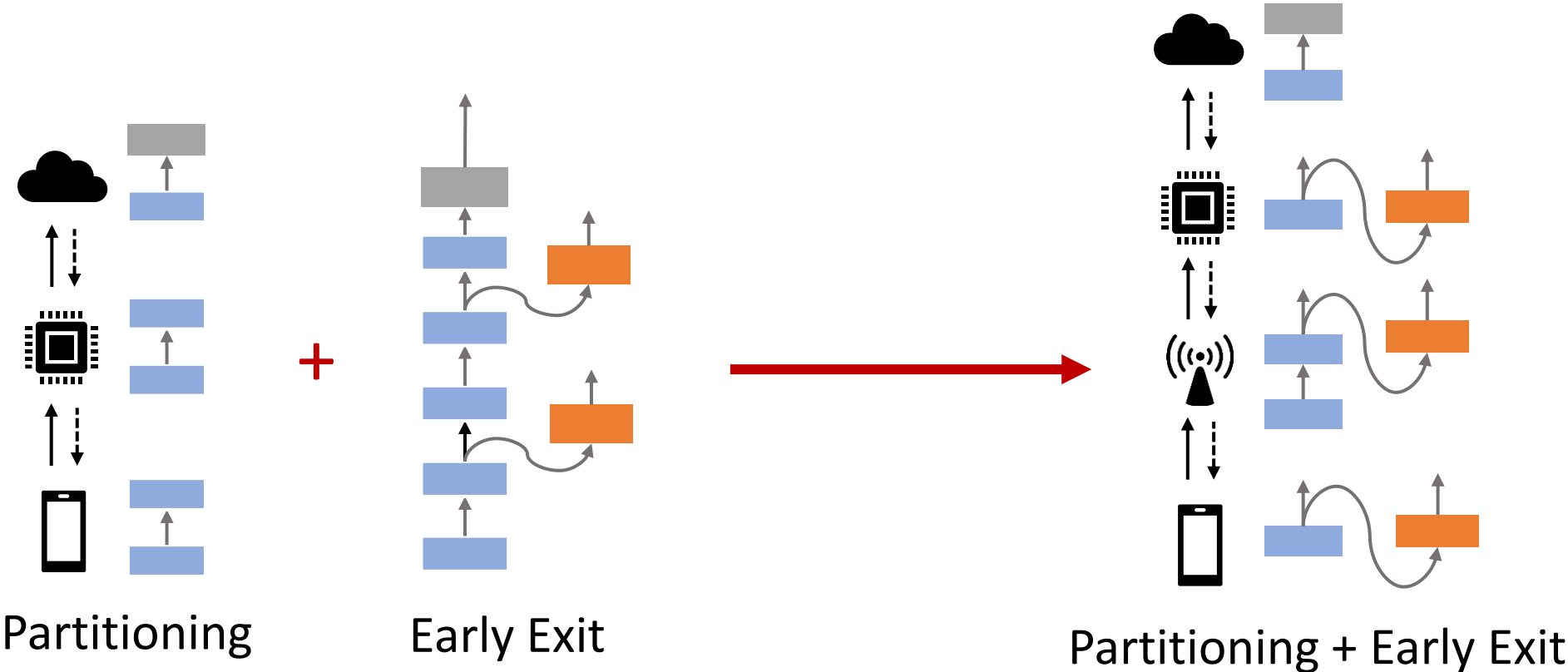
$\text{Entropy}(\text{output}) > \text{Threshold} \longrightarrow \text{Stay}$

This approach trades-off accuracy for computation latency

Combining DNN partitioning and early exit

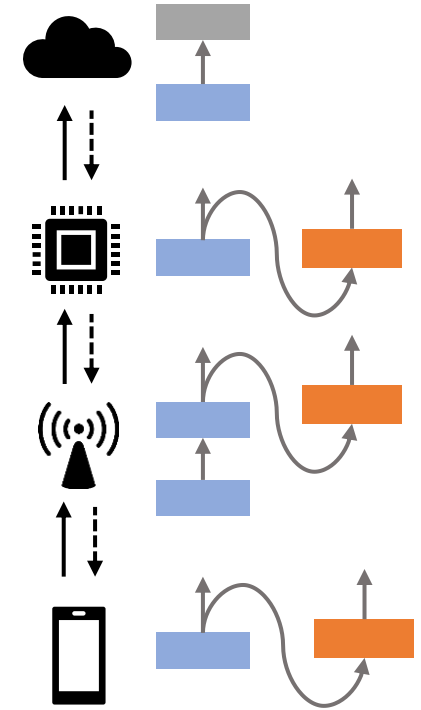
Consider partitioning and early exit holistically

A performance model that jointly optimizes them



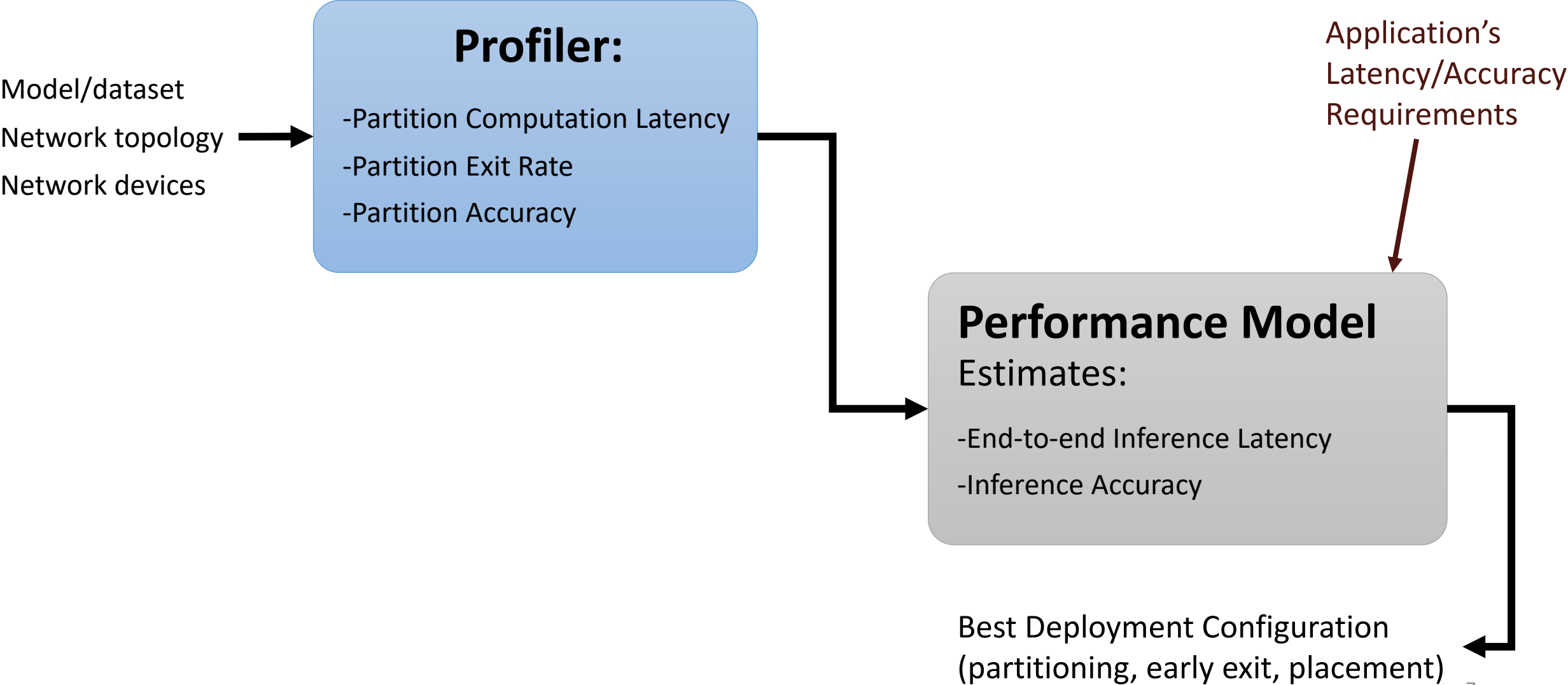
Combining DNN partitioning and early exit

1. Where to *partition* the model and attach the early exit branches?
 2. How to *place* these partitions on network devices?
- Calculate the inference latency and accuracy
 - Objective: minimizing latency/maximizing accuracy

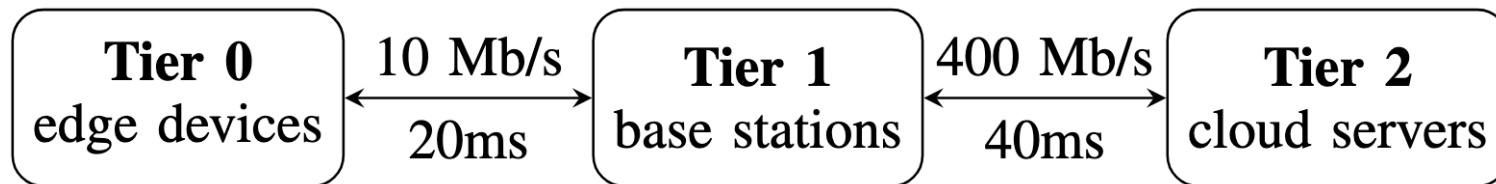


Partitioning + Early Exit

PORTEND



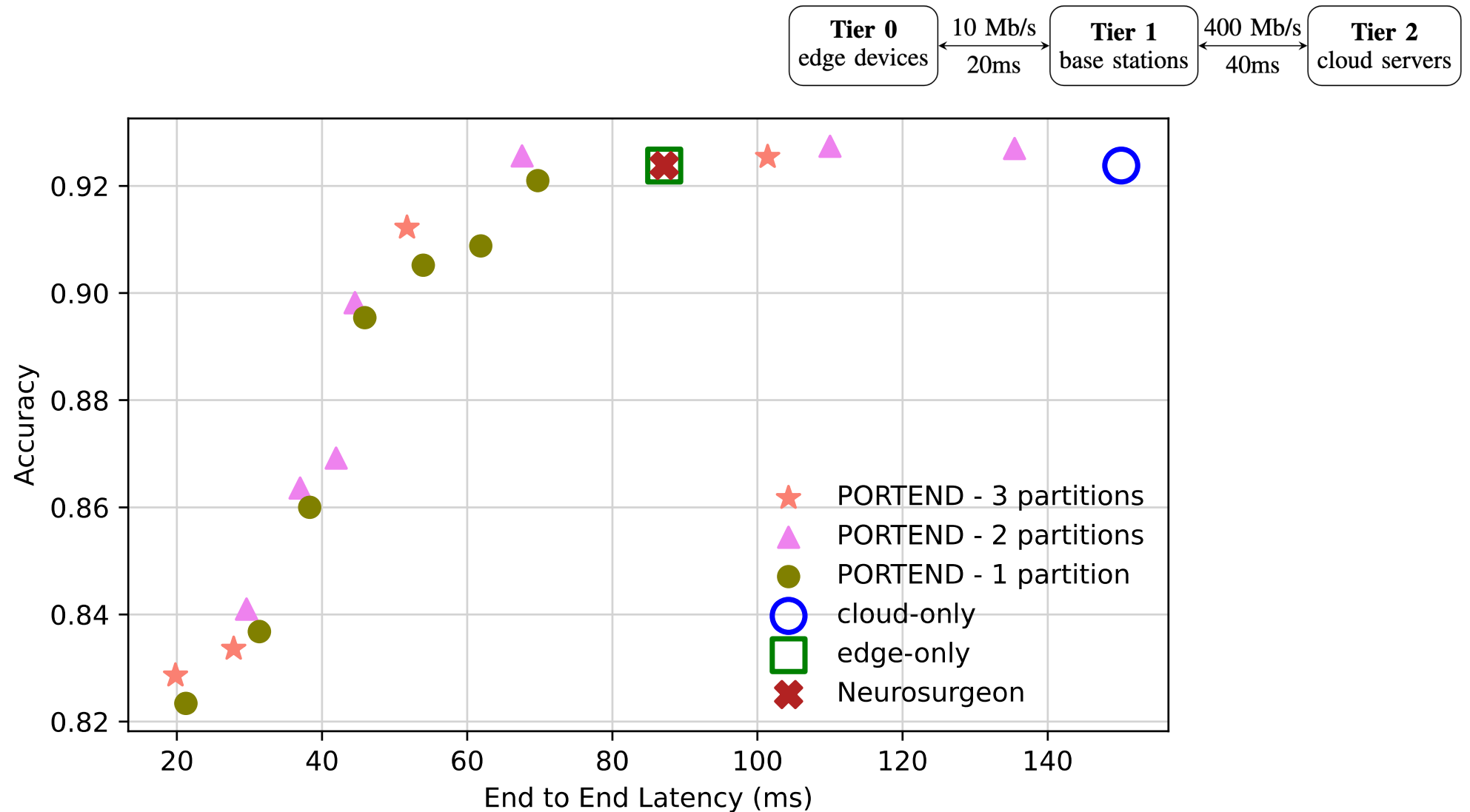
Experimental Setup



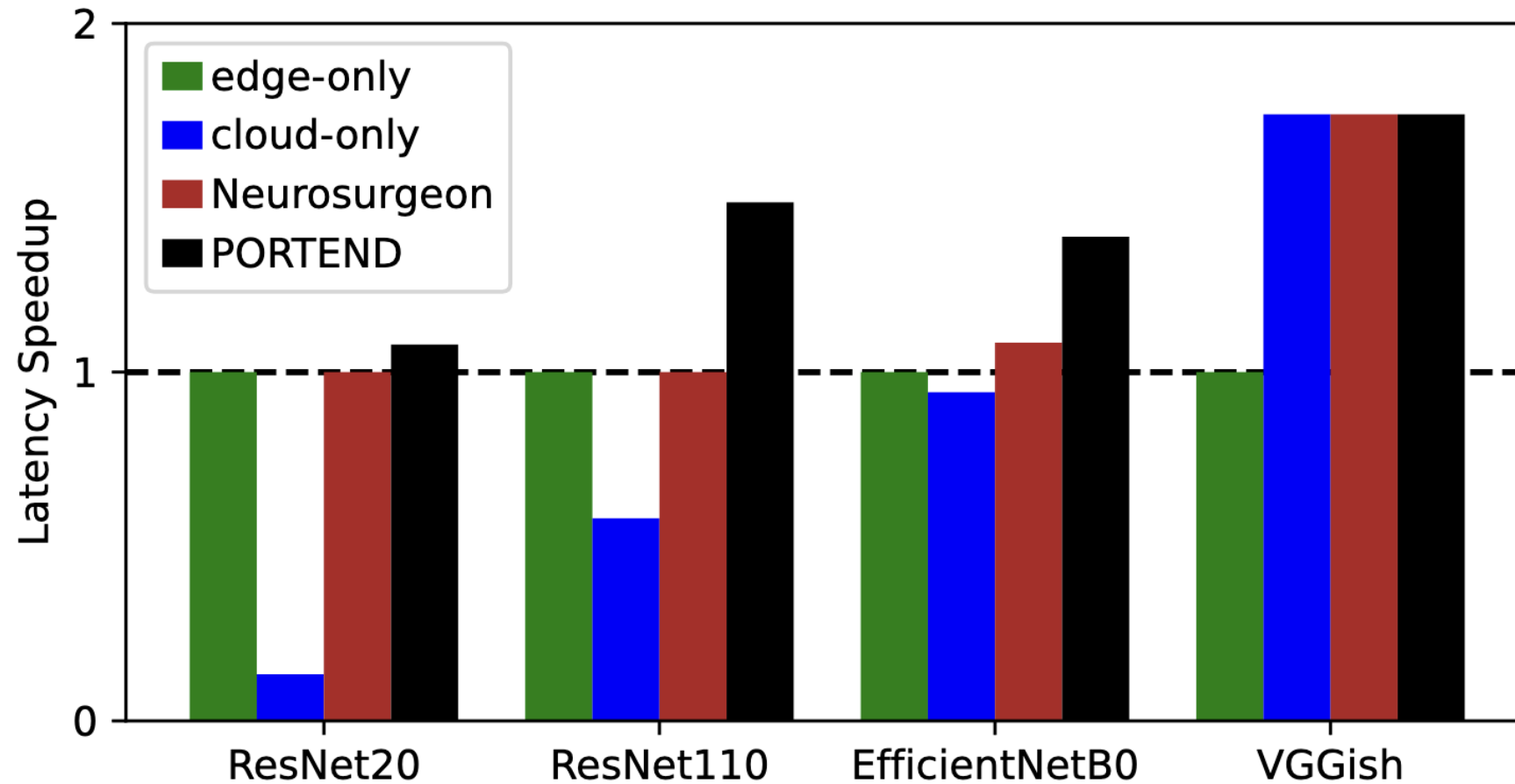
Tier	EC2 Type	CPU	Cores	GPU
0 (edge device)	a1.medium*	Graviton (ARM)	1	–
1 (base station)	m4.large	Intel Xeon	2	–
2 (cloud server)	g4dn.xlarge	Intel Xeon	4	T4

Model	Dataset	Blocks
ResNet-20	CIFAR-10	10
ResNet-110	CIFAR-10	55
EfficientNet-B0	ImageNet	8
VGGish	AudioSet	4

Benefits of Multiple Partitions - ResNet-110

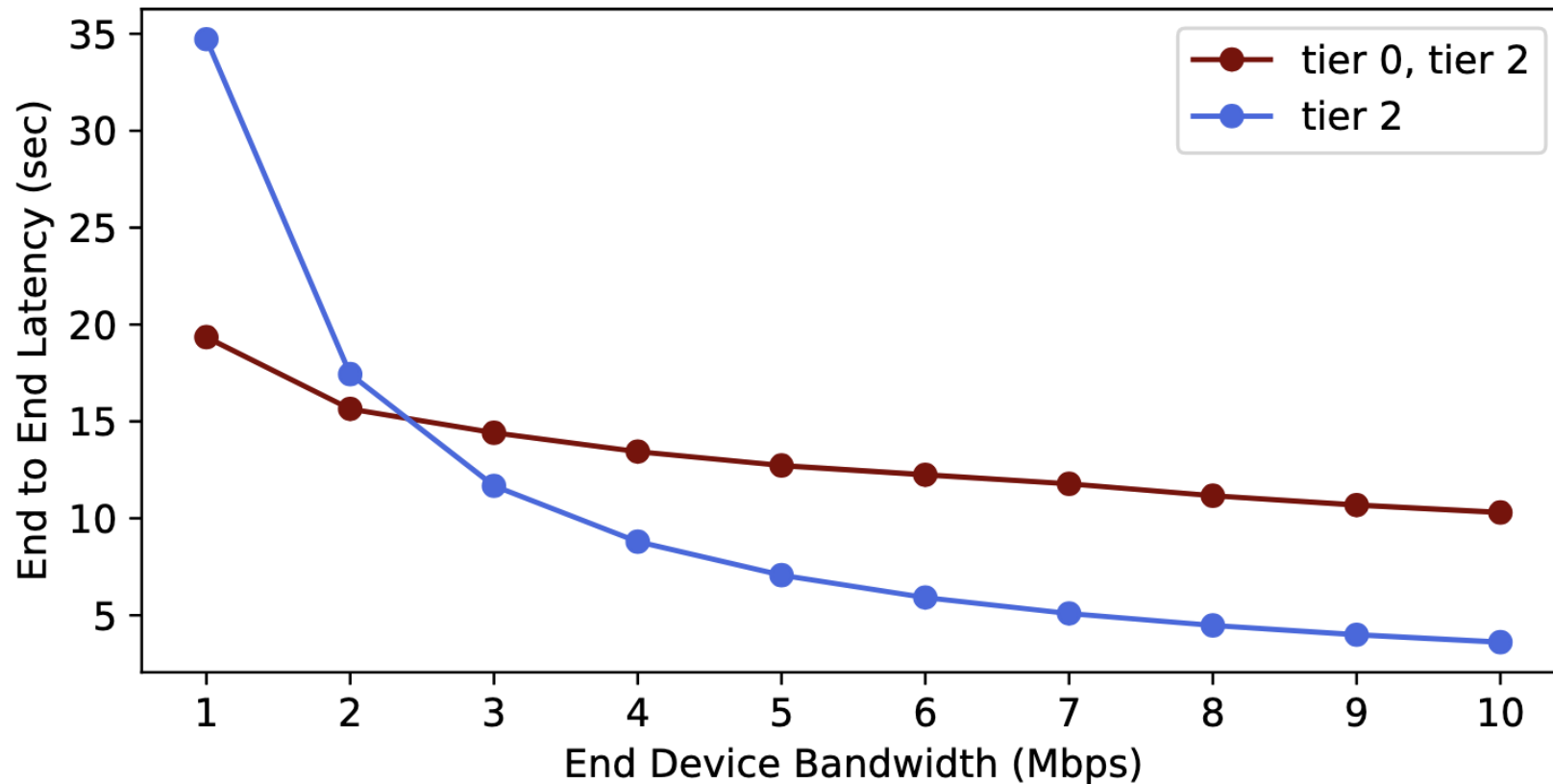


PORTEND Latency Speed-up



Exploring Hypothetical Scenarios

Reducing Bandwidth – EfficientNetB7



Conclusions & Future Work:

1. Considering multiple partitions with early exits improves the latency-accuracy trade-off.
2. Considering flexible placement is necessary.
3. Future work:
 1. Model compression and quantization
 2. Dynamic scheduling environment

Thank you!

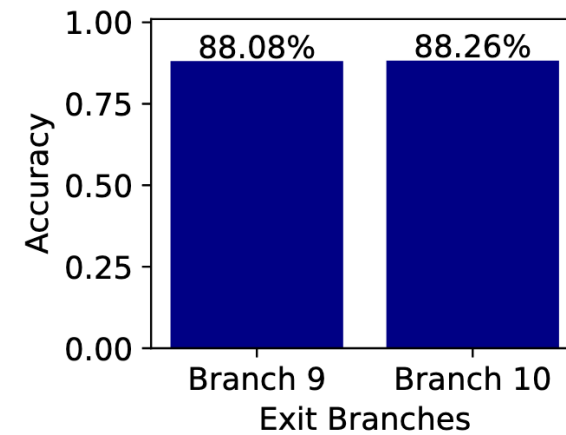
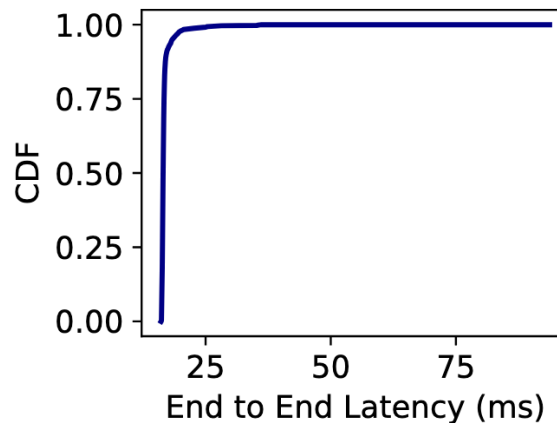
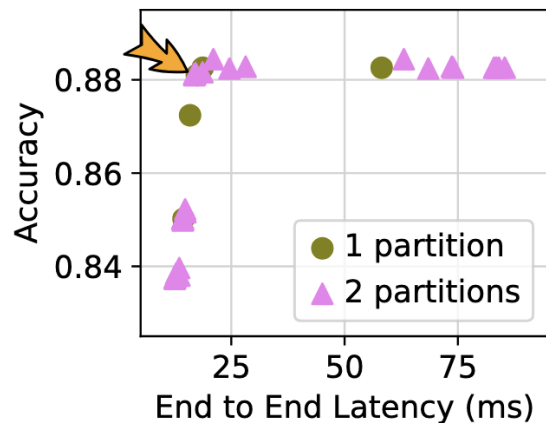
Questions?

Experimental Setup

Image and Audio Classification Models

Model	Dataset	Blocks	Optimization Time
ResNet-20	CIFAR-10	10	21.2s
ResNet-110	CIFAR-10	55	30m
EfficientNet-B0	ImageNet	8	16.6s
VGGish	AudioSet	4	12.2s

Finding Optimal Configurations- ResNet20



Model	Tiers [blocks]	Threshold	Exit rates
ResNet-20	0 [1-9] , 1 [10]	0.7	99% , 1%

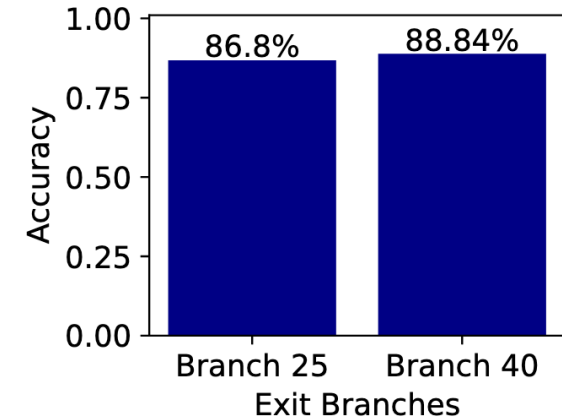
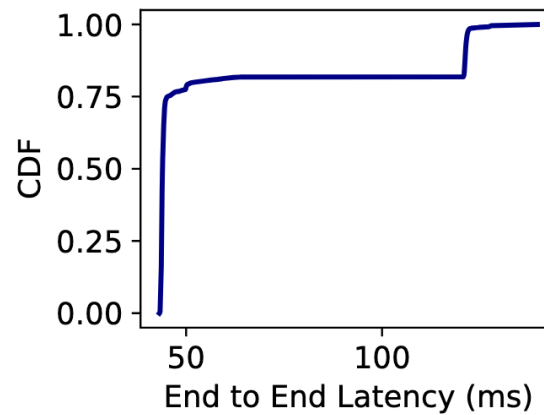
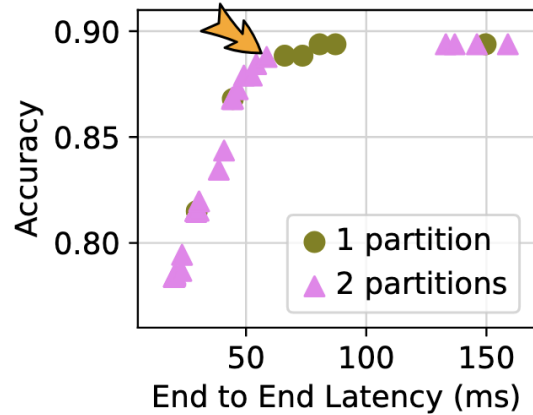
Partition1 (on tier0)

Partition2 (on tier1)

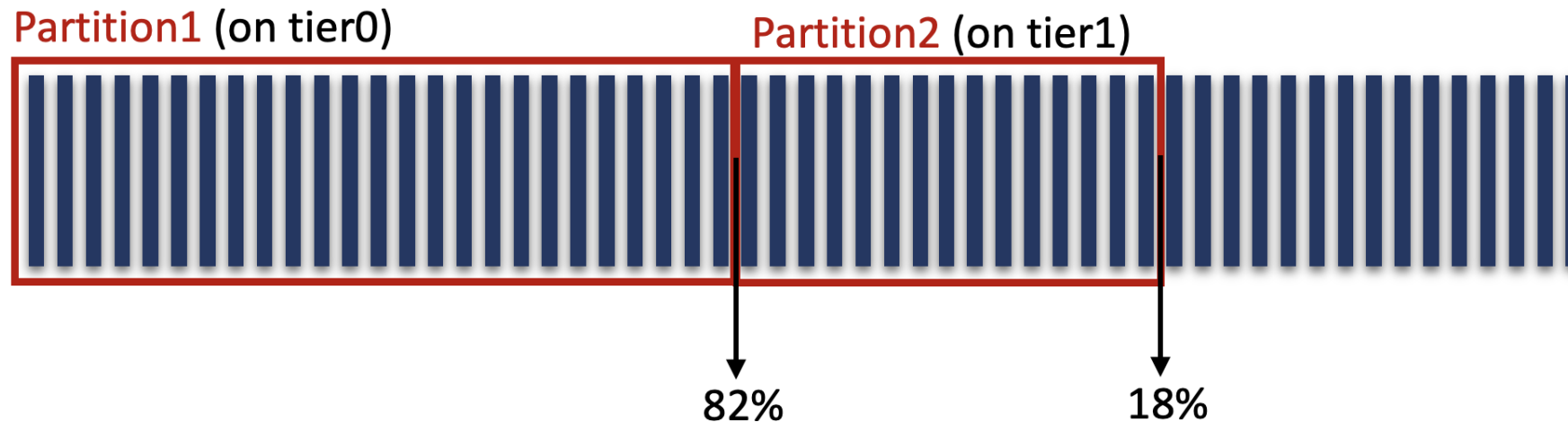


99% 1%

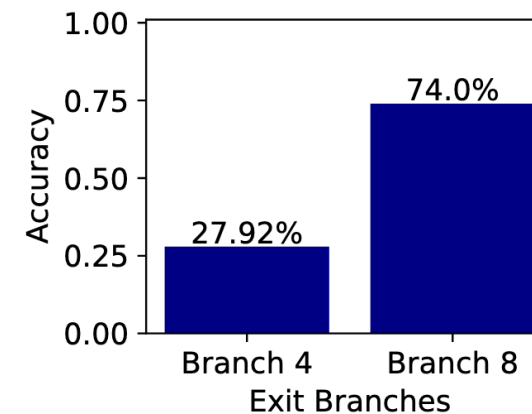
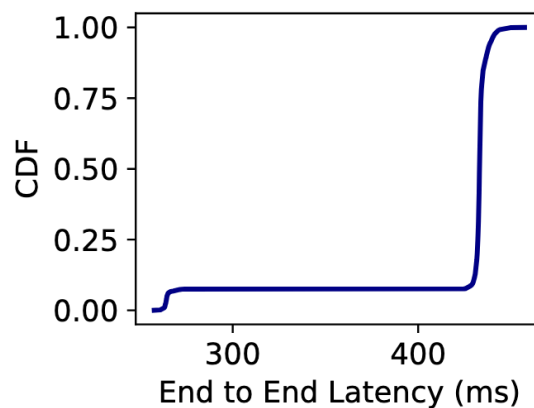
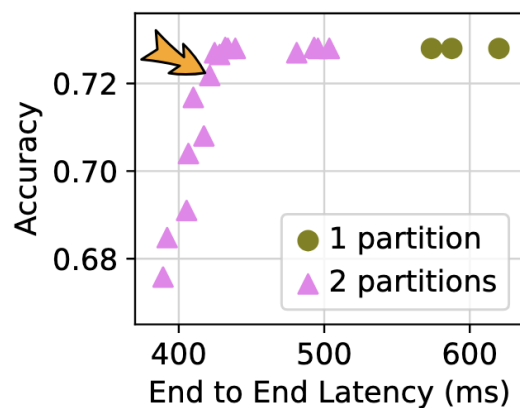
Finding Optimal Configurations- ResNet110



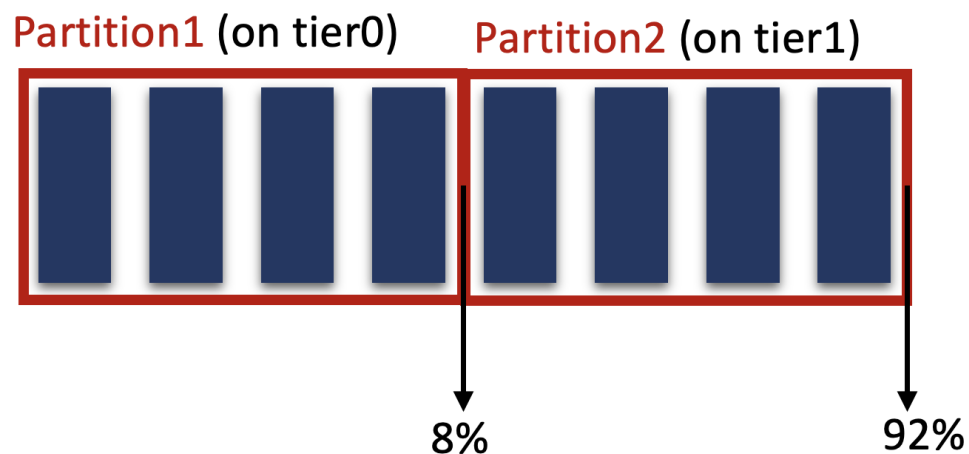
Model	Tiers [blocks]	Threshold	Exit rates
ResNet-110	0 [1-25] , 1 [26-40]	0.1	82% , 18%



Finding Optimal Configurations- EfficientNetB0

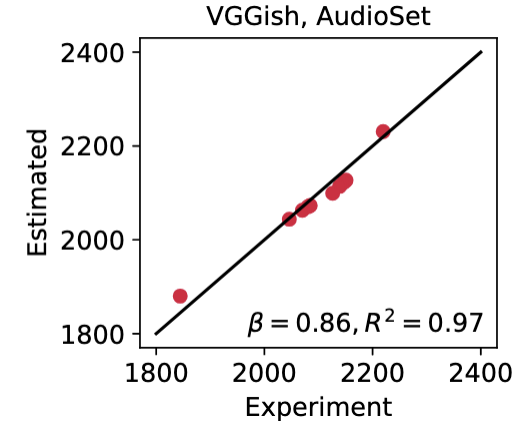
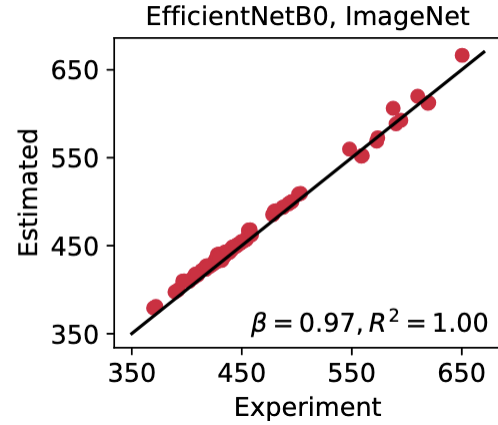
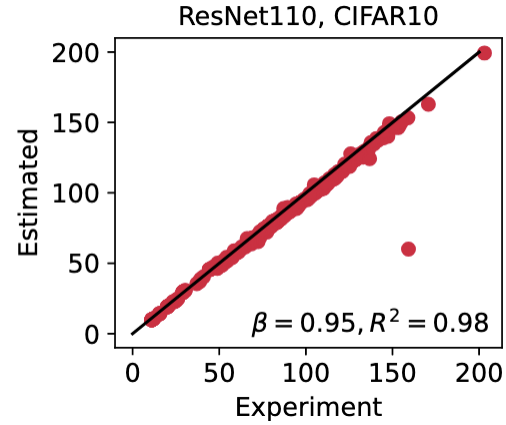
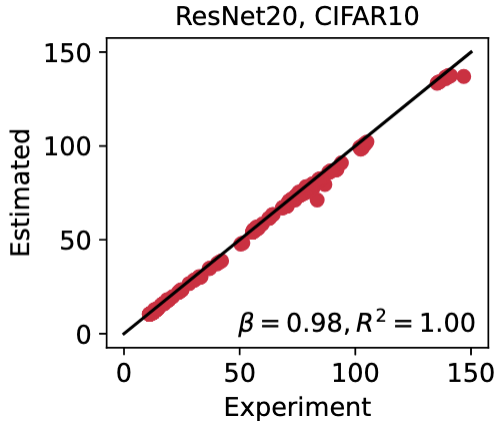


Model	Tiers [blocks]	Threshold	Exit rates
EfficientNet-B0	0 [1-4] , 1 (5-8)	0.3	8% , 92%

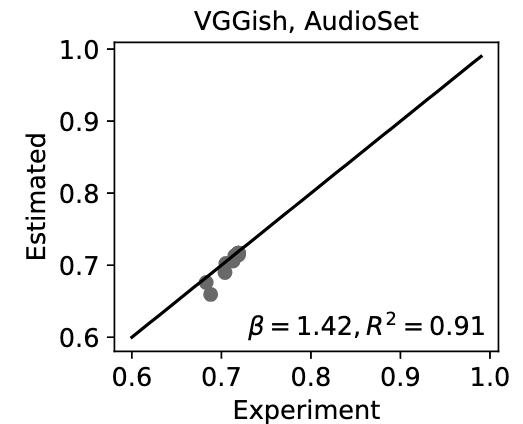
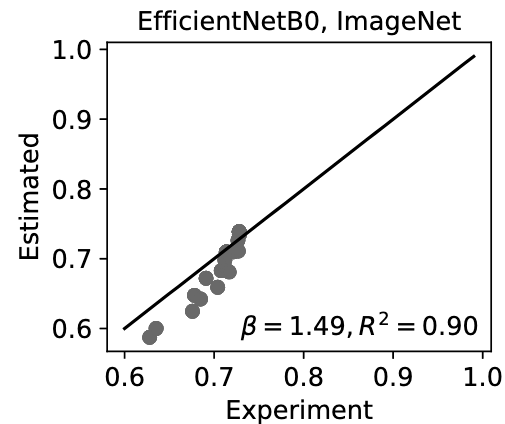
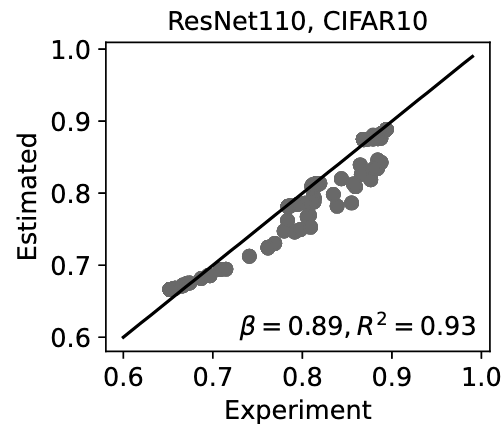
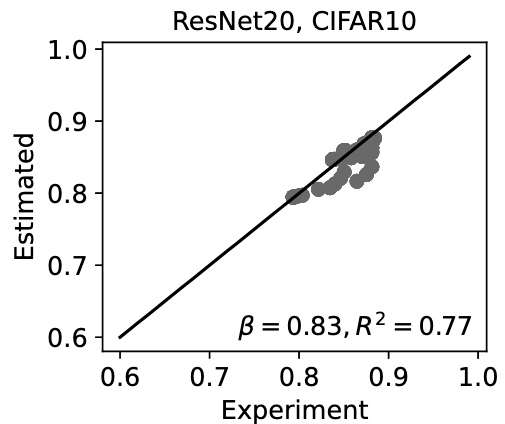


Validity of the Performance Model

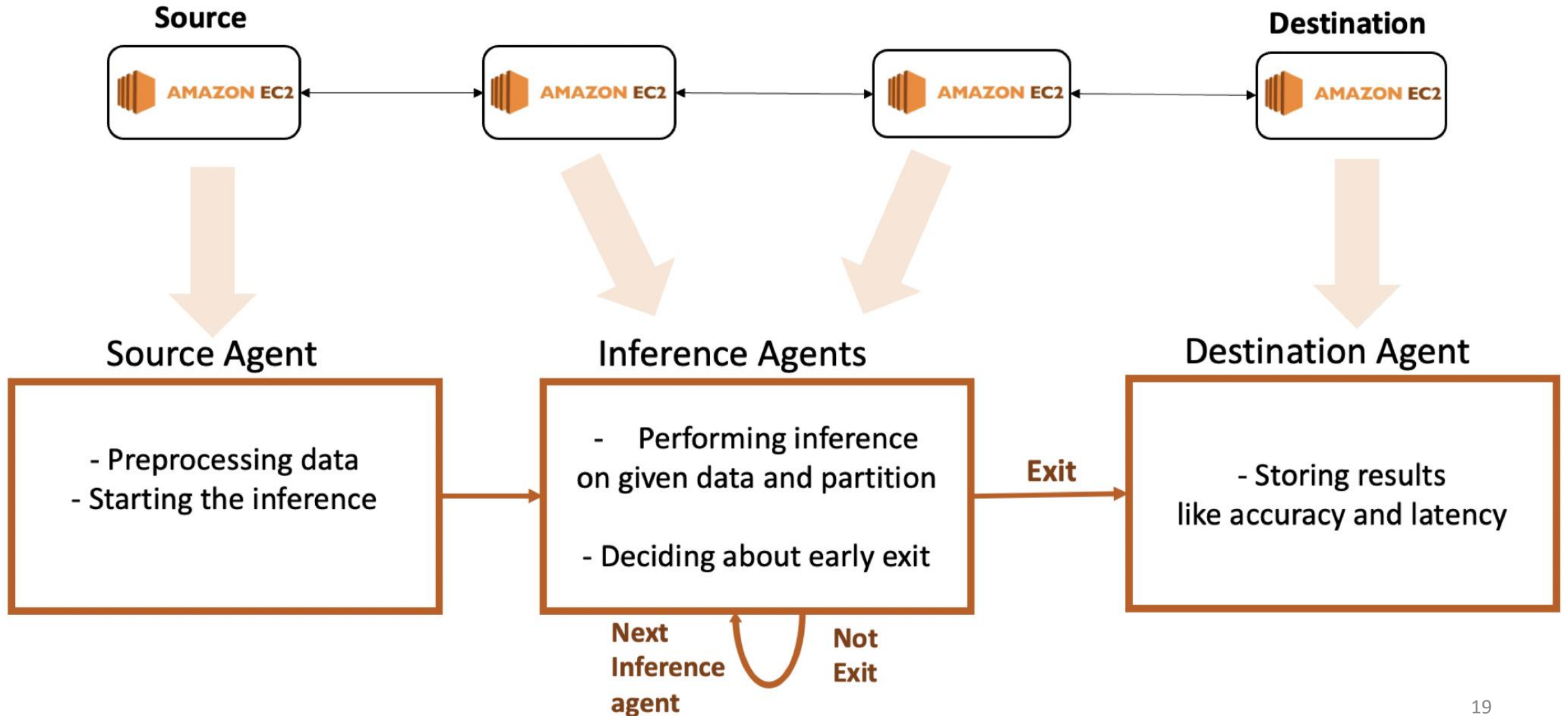
End-to-End Inference Latency



Inference Accuracy

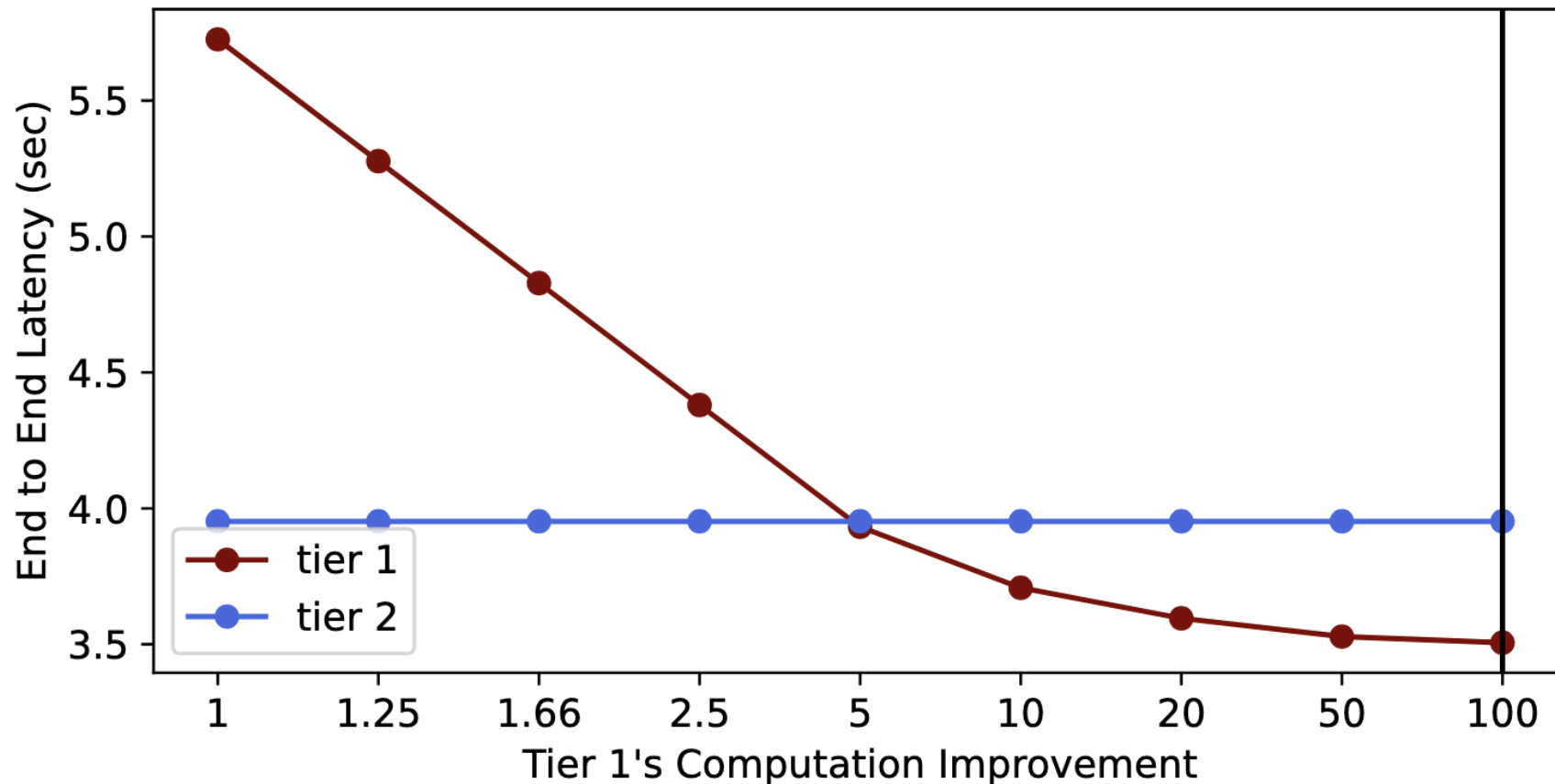


Experimental Setup



Exploring Hypothetical Scenarios

Increasing Compute Power – EfficientNetB7



Effect of Fine-Tuning on final models

