



Agua: A Concept-Based Explainer for Learning Enabled Systems

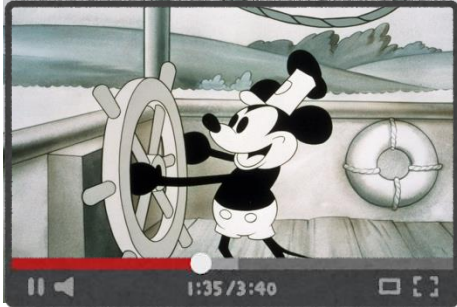
Sagar Patel, Dongsu Han, Nina Narodystka, Sangeetha Abdu Jyothi

UC Irvine



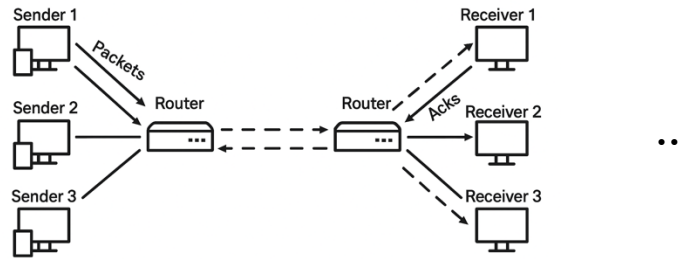
vmware[®]
by Broadcom

Promise of Learning Enabled Systems



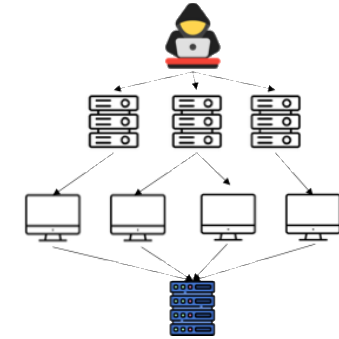
Adaptive Video Streaming:

Pensieve [SIGCOMM'17], Fugu [NSDI'19], ...
Gelato [CoNEXT'24]



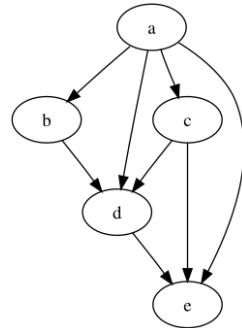
Congestion Control:

Aurora [ICML'19], Orca [SIGCOMM'20]...
Sage [SIGCOMM'22]



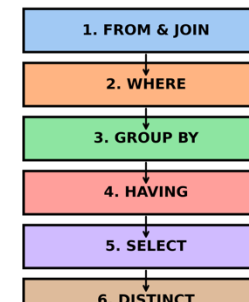
DDoS Detection:

GRU-RNN [NetSoft'18], ...
LUCID [IEEE TNSM'20]



Cluster scheduling:

Decima [HotNets'18], Gavel [SIGCOMM'20]...
Sinan [SIGCOMM'22]



Query Optimization:

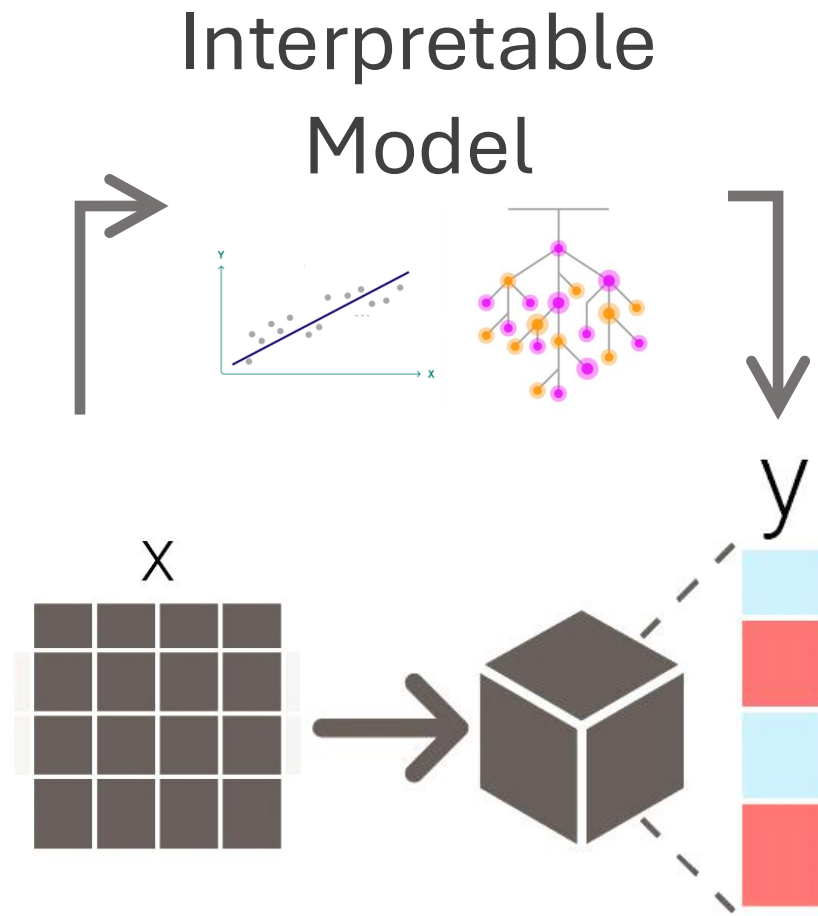
Neo [VLDB'18], Bao [SIGMOD'21]...
Lero [VLDB'23]

The Need for Explainability

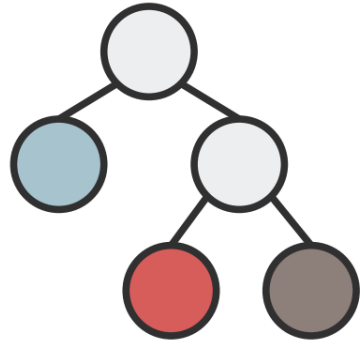
- Operators are hesitant to deploy ML solutions despite high performance in lab
- Difficulty:
 - Understanding
 - Debugging
 - Trusting



Surrogate Functions as Explanations

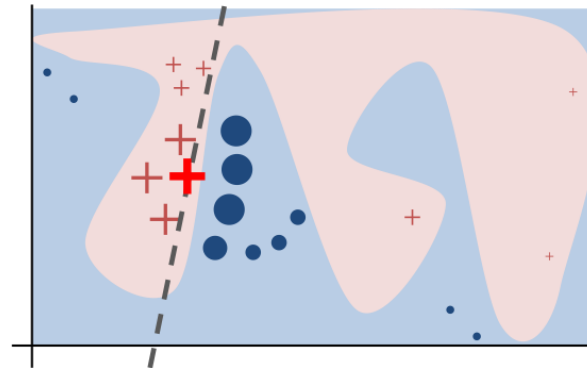


Feature-level Explainers



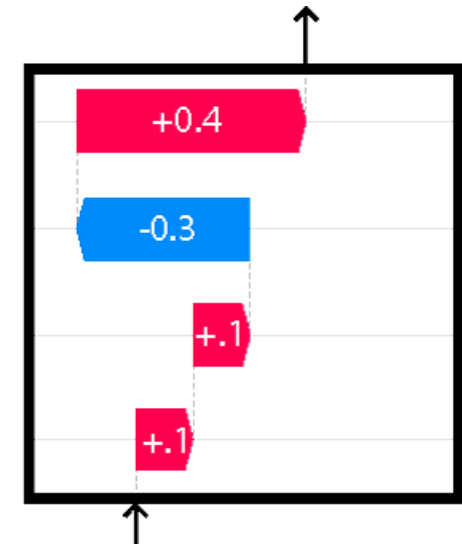
Decision Trees

Metis [SIGCOMM'20]
Trustee [CCS'22]



Local Linear Regression

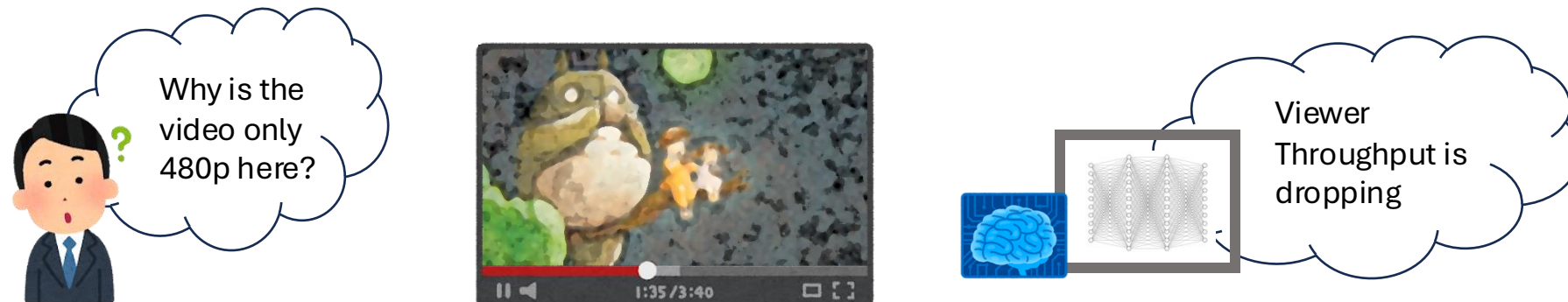
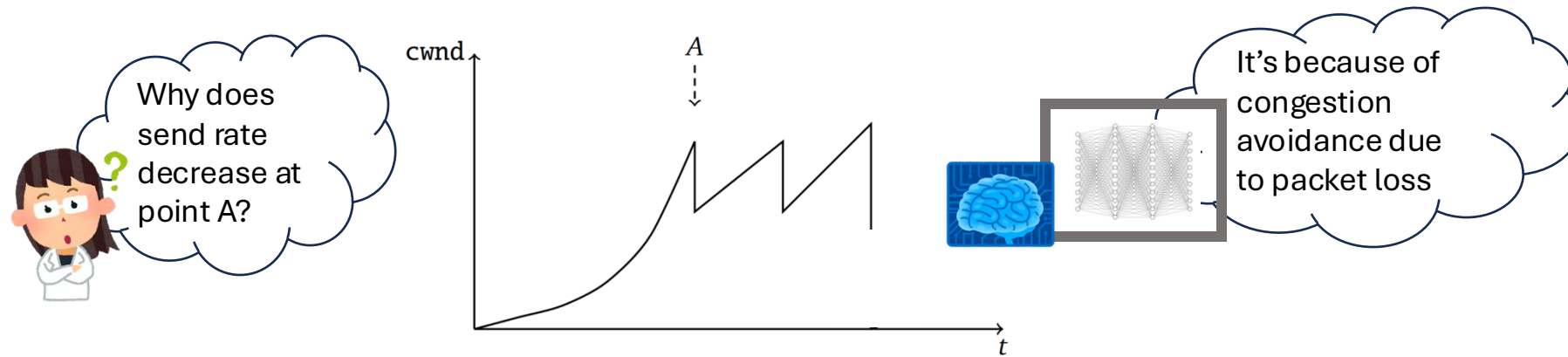
LIME [SIGKDD'16]



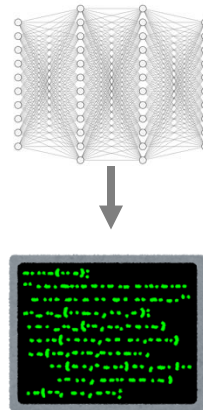
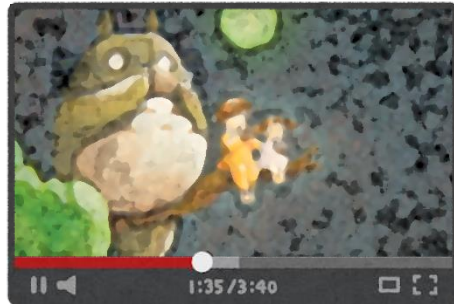
Game-theoretic

SHAP [KAIS'14]

The Vision: Intuitive Understanding

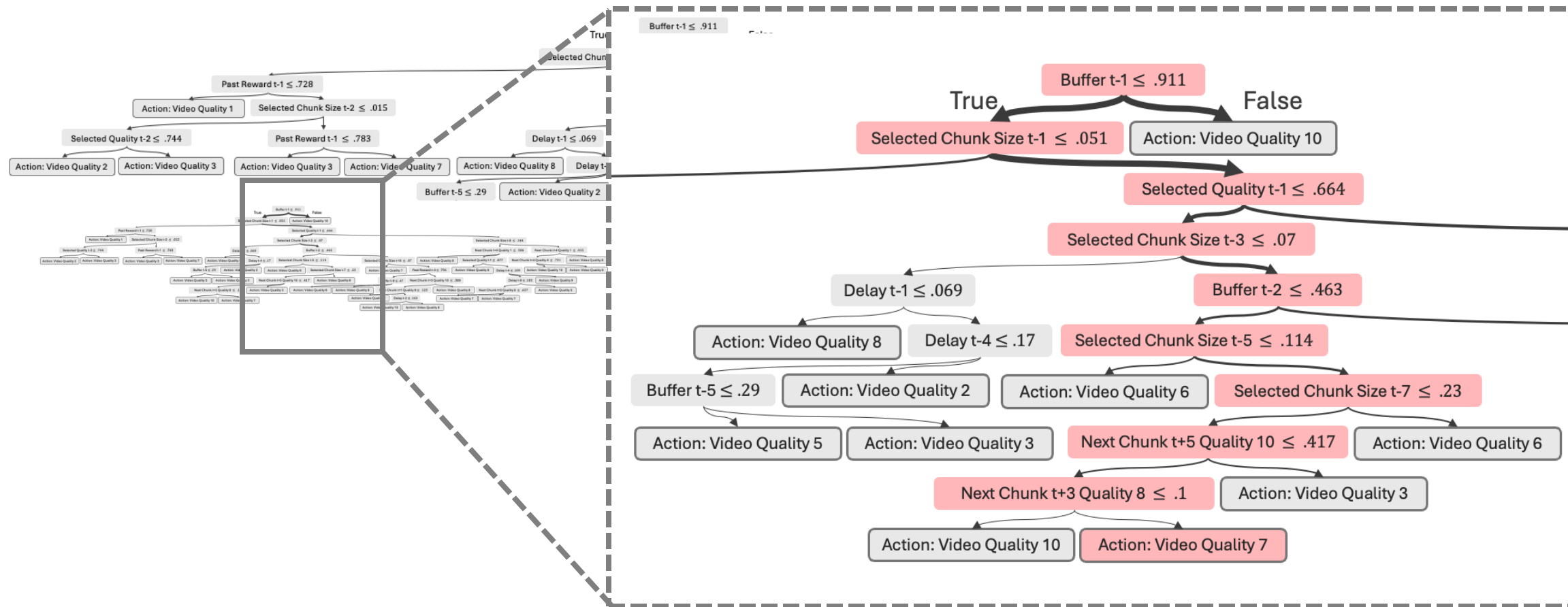


Reality of Current State-of-the-Art



Buffer $t-1 \leq .911$; Chunksize $t-1 \leq .051$; Quality $t-1 \leq .664$;
Chunksize $t-3 \leq .070$; Buffer $t-2 \leq .463$; Chunksize $t-5 \leq .114$;
Chunksize $t-7 \leq .230$; Next
Chunk $t+5$ Quality $10 \leq .417$;
...

Reality of Current State-of-the-Art

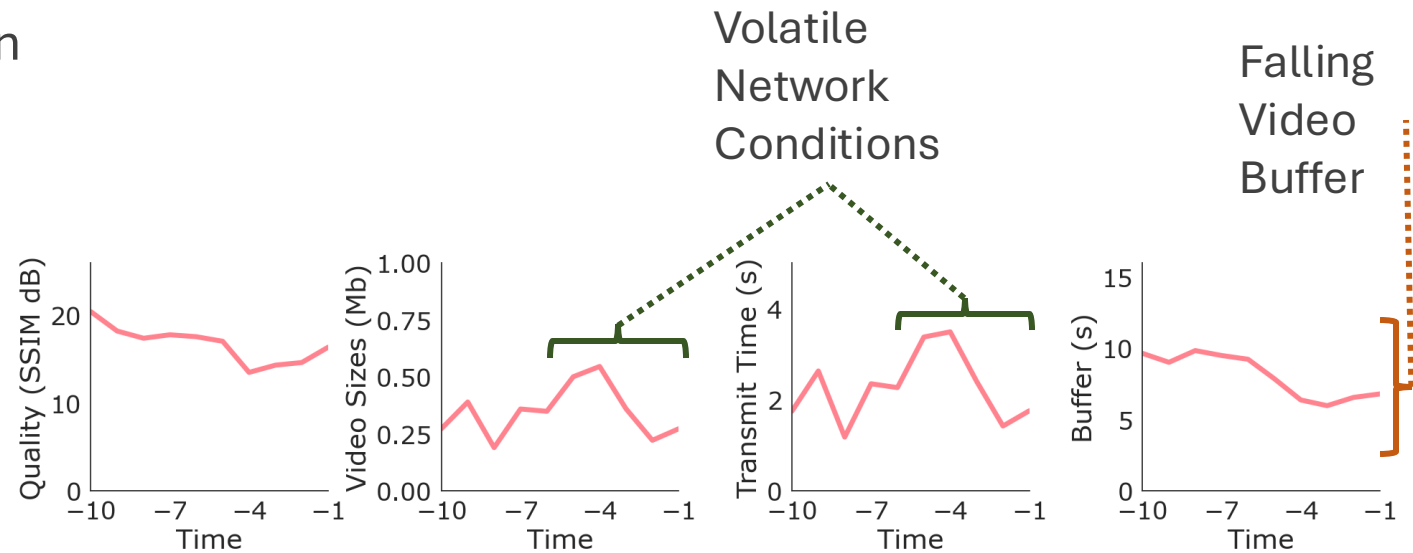


Limitations of Feature-level Explainers in Systems

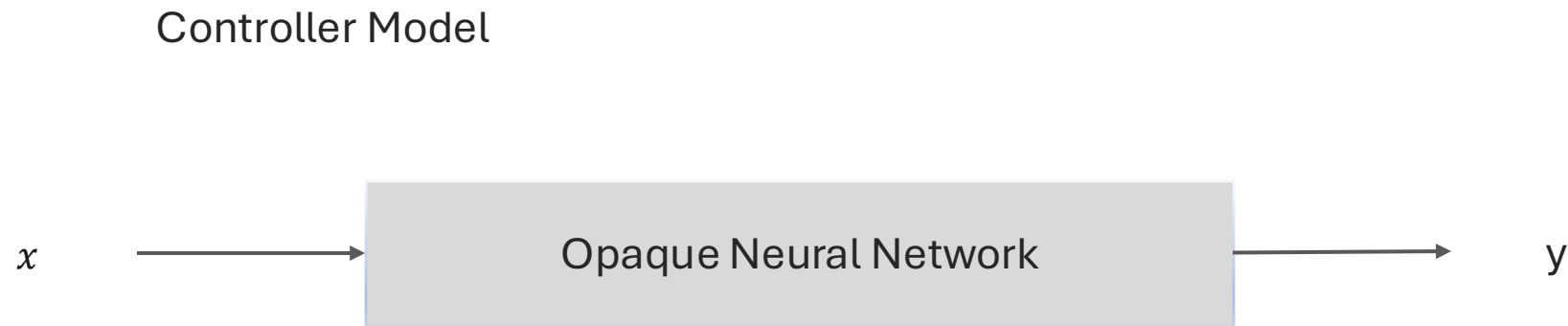
- Salient features as explanations
 - Can involve low-level system features
 - Miss trends and inter-feature interactions
 - Difficult to decode

Key Idea: Concepts as Units of Explanation

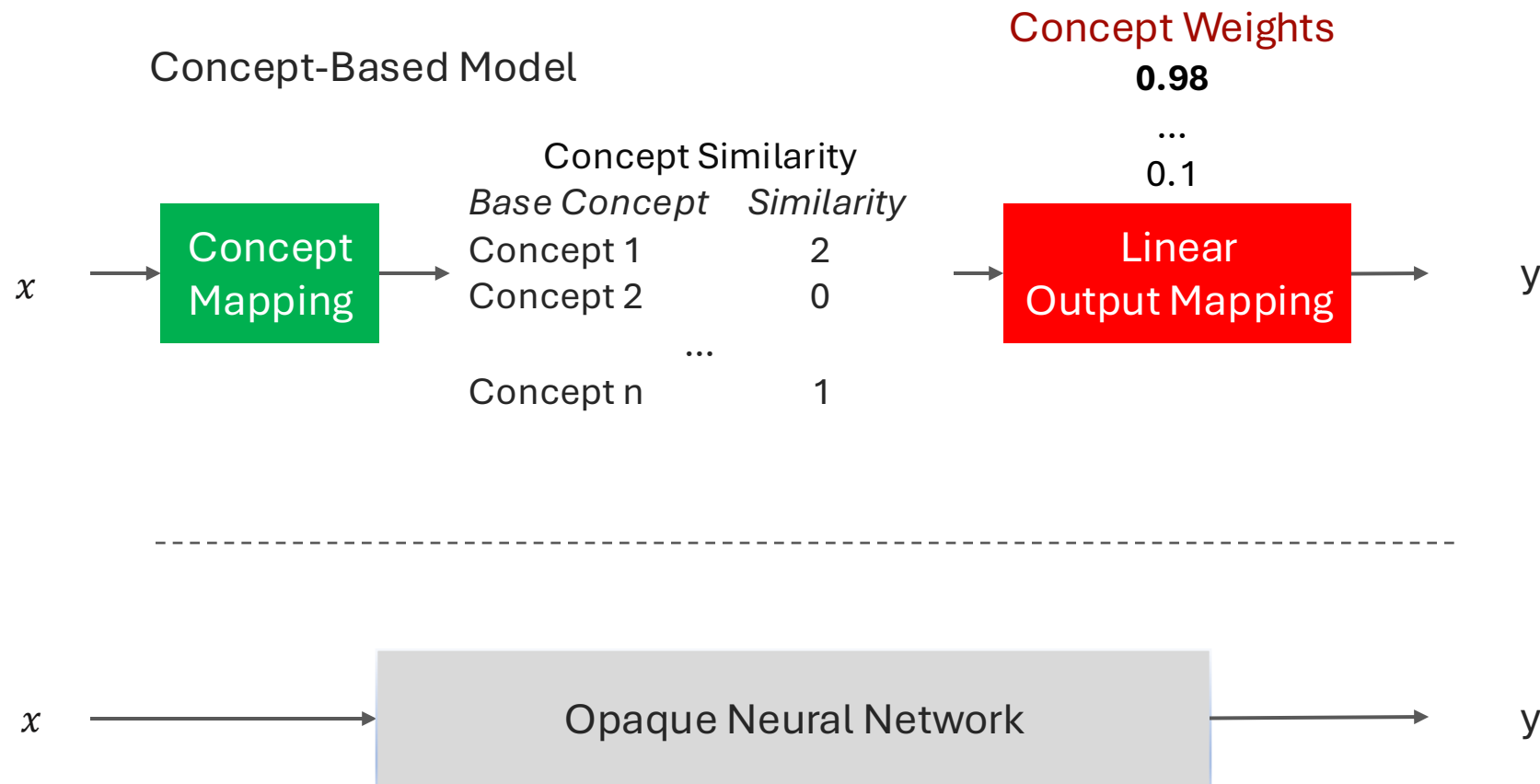
- Concepts
 - Human-understandable attributes that capture controller and environment characteristics
 - Can capture intricate patterns, trends, and behaviors in systems
 - Align with human intuition



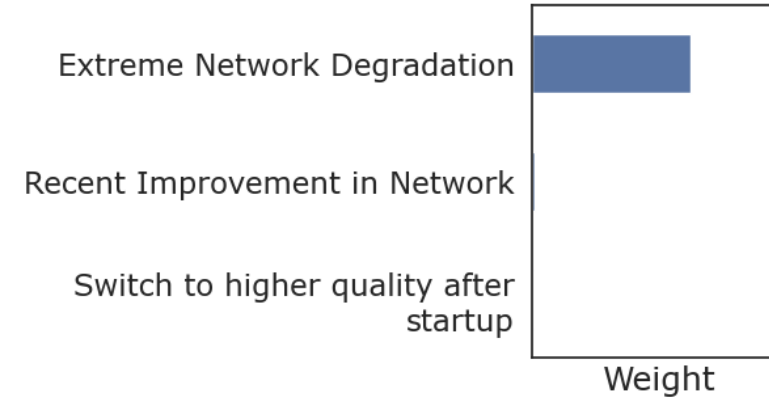
Key Idea: Surrogate Concept-Based Model



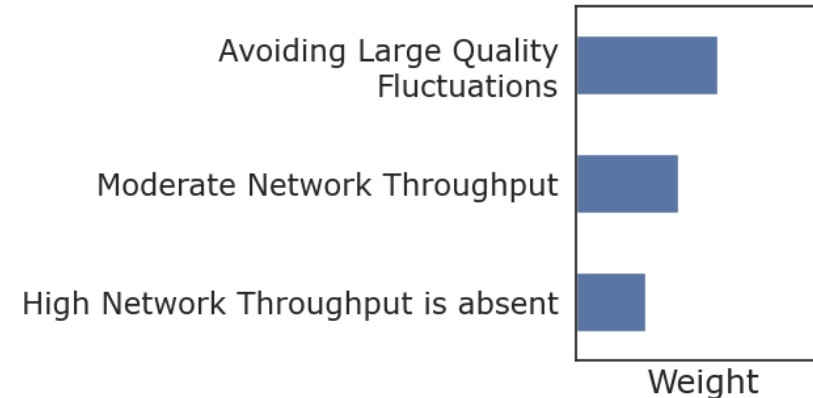
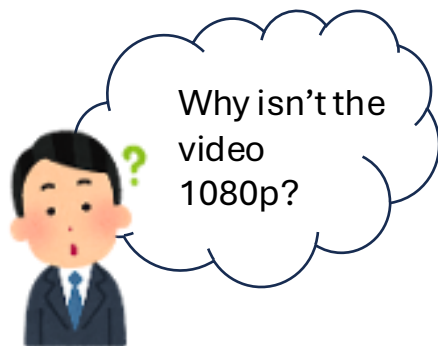
Key Idea: Surrogate Concept-Based Model



Concept-Based Explanation

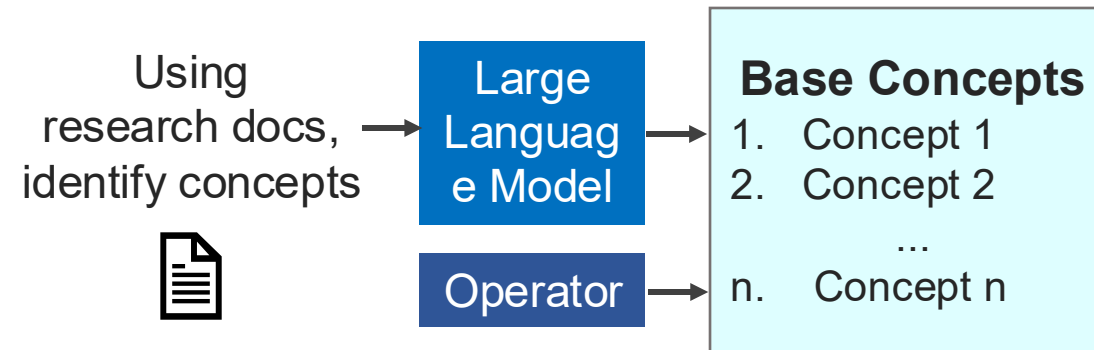


Factual Explanation



Counterfactual Explanation

Training Pipeline: 1. Base Concept Generation



Adaptive Bitrate Selection: A Survey

Yusuf Sani, Andreas Mauthe, and Christopher Edwards

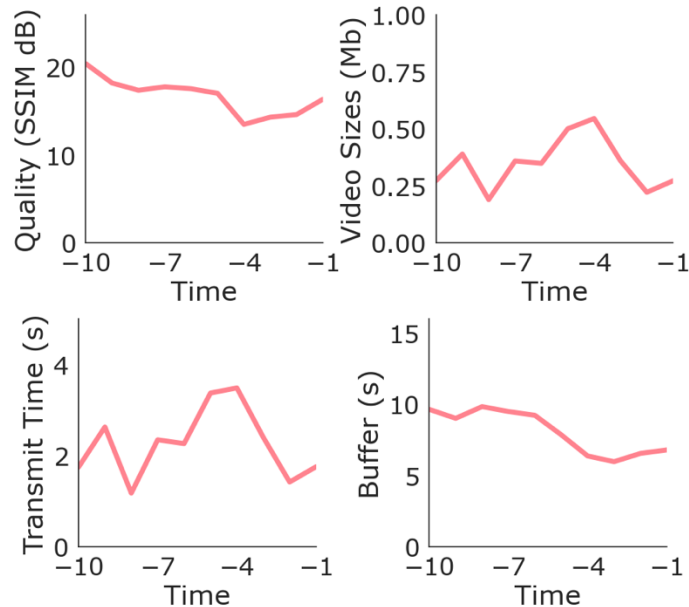
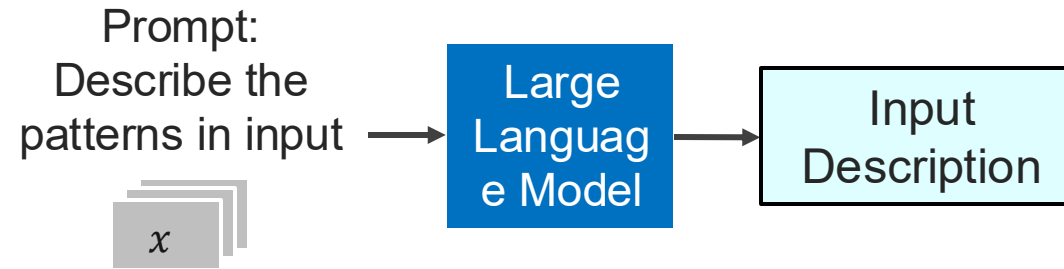
A comprehensive survey on machine learning for networking: evolution, applications and research opportunities

Raouf Boutaba^{1*}, Mohammad A. Salahuddin¹, Noura Limam¹, Sara Ayoubi¹, Nashid Shahriar¹, Felipe Estrada-Solano^{1,2} and Oscar M. Caicedo²

Base Concepts

-
- 1. Volatile Network Conditions
 - 2. Falling Buffer
 - ...
 - n. Medium Throughput

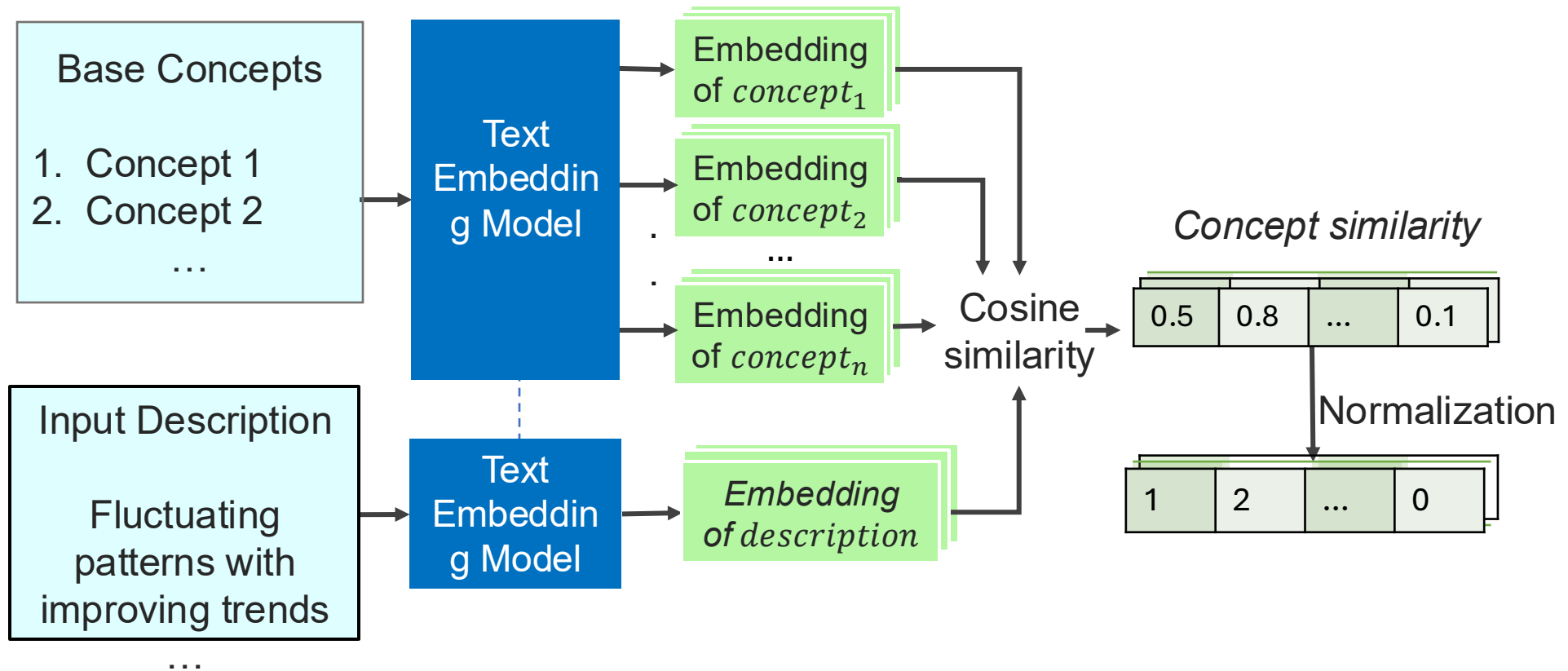
Training Pipeline: 2. Input Description Generation



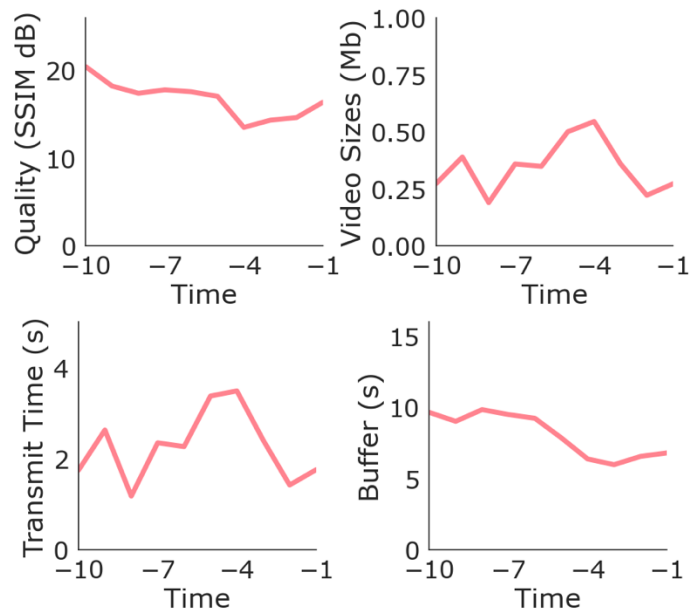
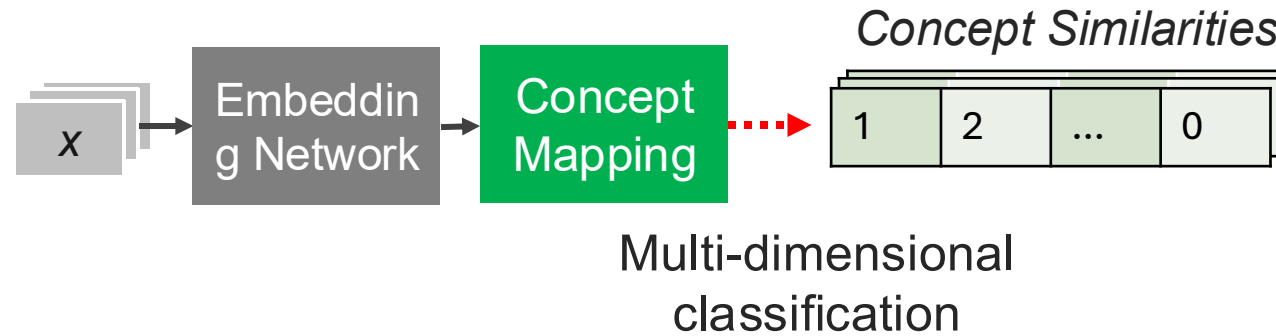
Network conditions:

- Initially starts off with a fluctuating pattern, as observed from the features "Transmission Time of Chunk."
- In the middle...
- Overall, the trend is improving, indicating the presence of stable network conditions.

Training Pipeline: 3. Target Concept Similarity Generation

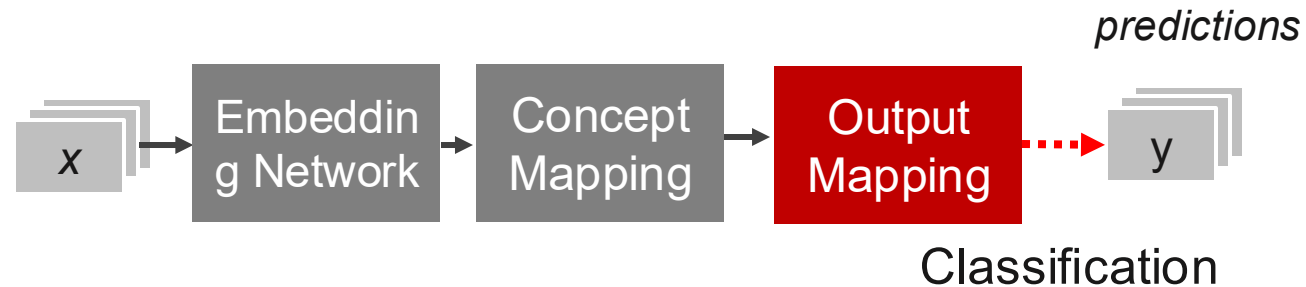


Training Pipeline: 4. Training Concept Mapping

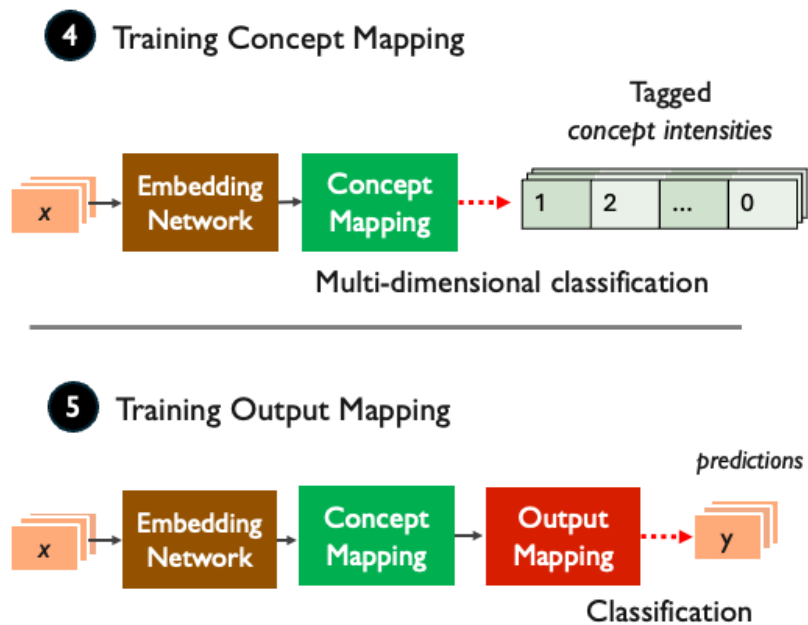
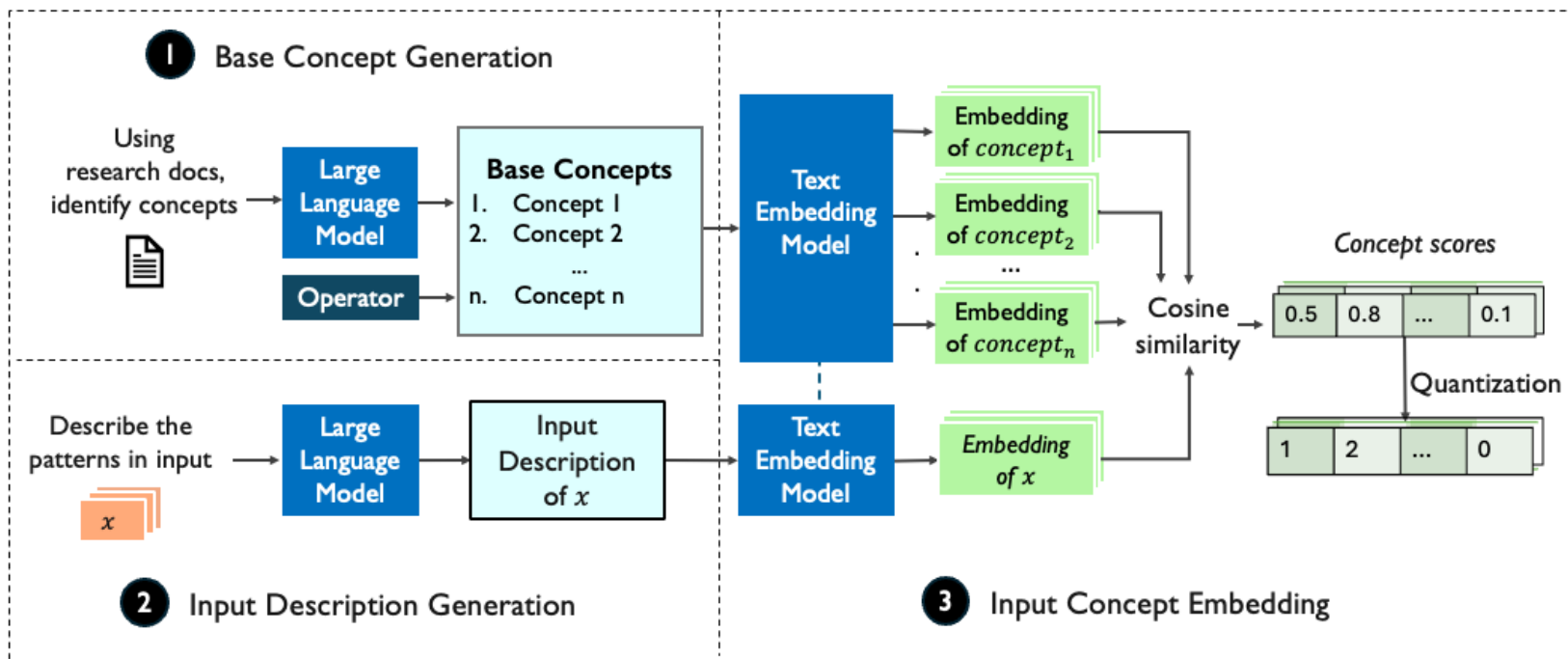


Concepts Similarity	
Base Concept	Similarity
1. Volatile Network Conditions	0
2. Falling Buffer	2
...	
n. Medium Throughput	1

Training Pipeline: 5. Training Linear Output Mapping



Training Pipeline



Experimental Settings

- LLMs evaluated:
 - Closed-source: OpenAI GPT 4o
 - Open-source: Llama 3.3 70B
- Baseline:
 - Trustee [CCS'22]

Application	Controller	Learning Paradigm
Adaptive Bitrate Streaming	Gelato [CoNEXT'24]	Reinforcement Learning
Congestion Control	Aurora [ICML'19]	Reinforcement Learning
DDoS Detection	LUCID [IEEE TNSM'20]	Supervised Learning

Fidelity

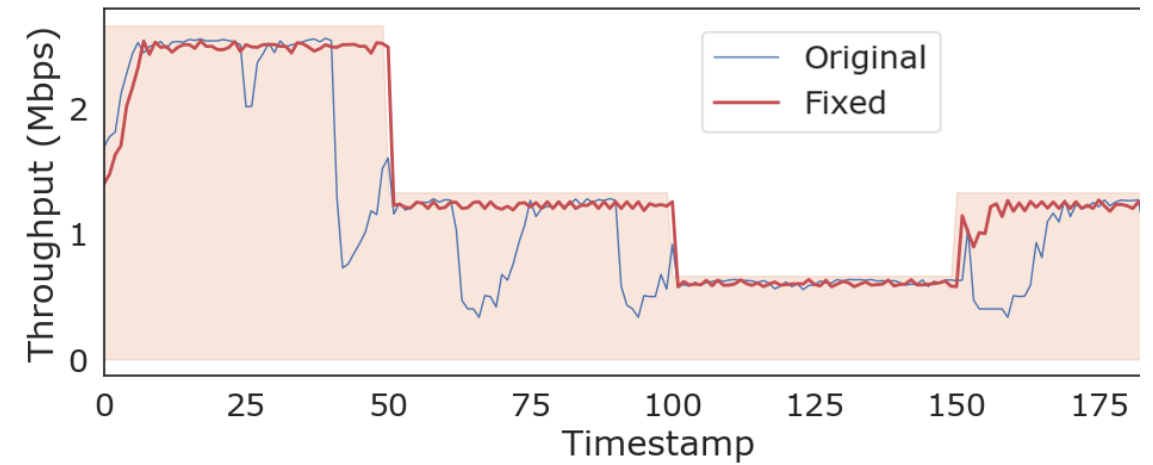
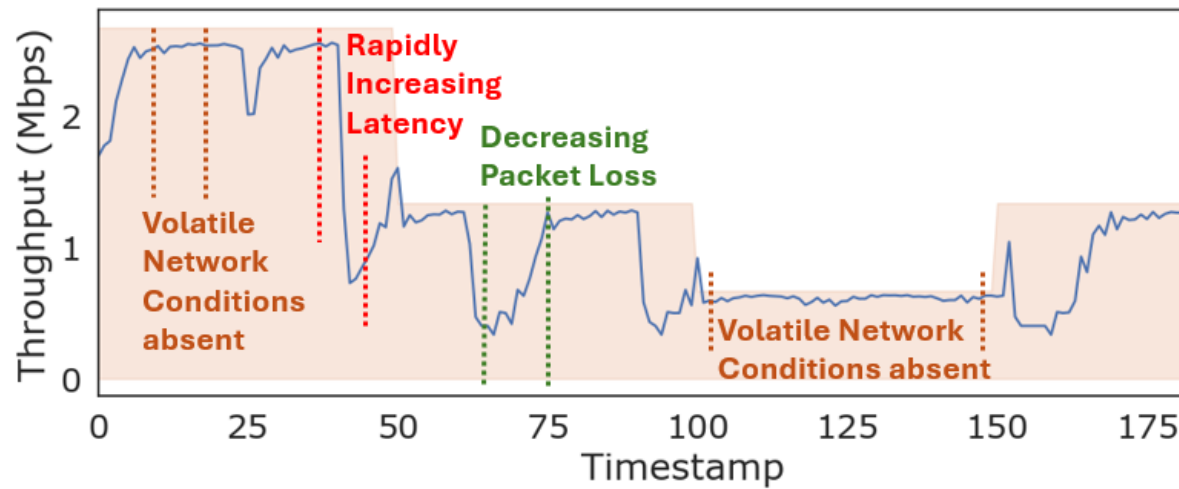
Application	Trustee		Agua	
	Full	Pruned	Llama 3.3 70B	GPT 4o
Adaptive Bitrate Streaming	0.946	0.949	0.982	0.983
Congestion Control	0.215	0.235	0.932	0.936
DDoS Detection	0.991	0.977	0.996	1.000

Use Cases

1. Debugging
2. Data shift detection
3. Concept-level retraining
4. Dataset expansion

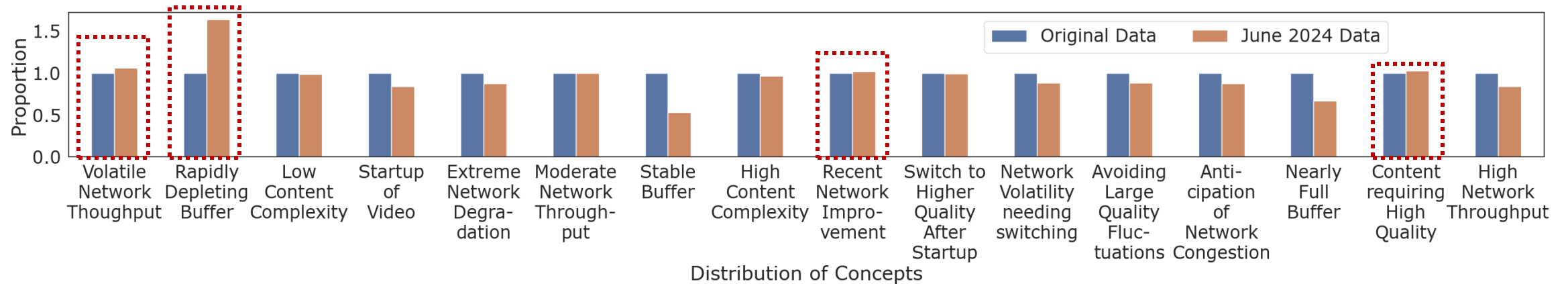
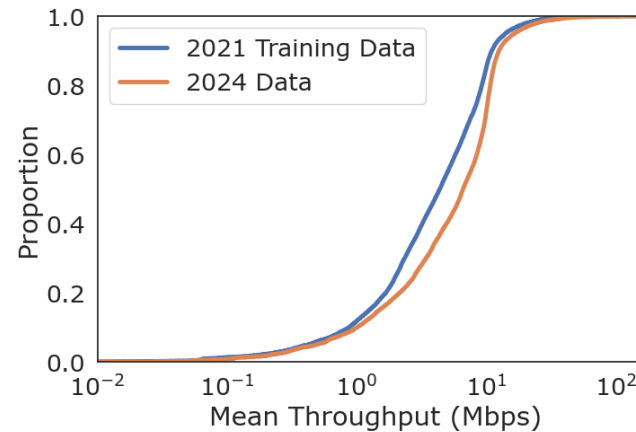
Use Cases: Debugging

Aurora Throughput



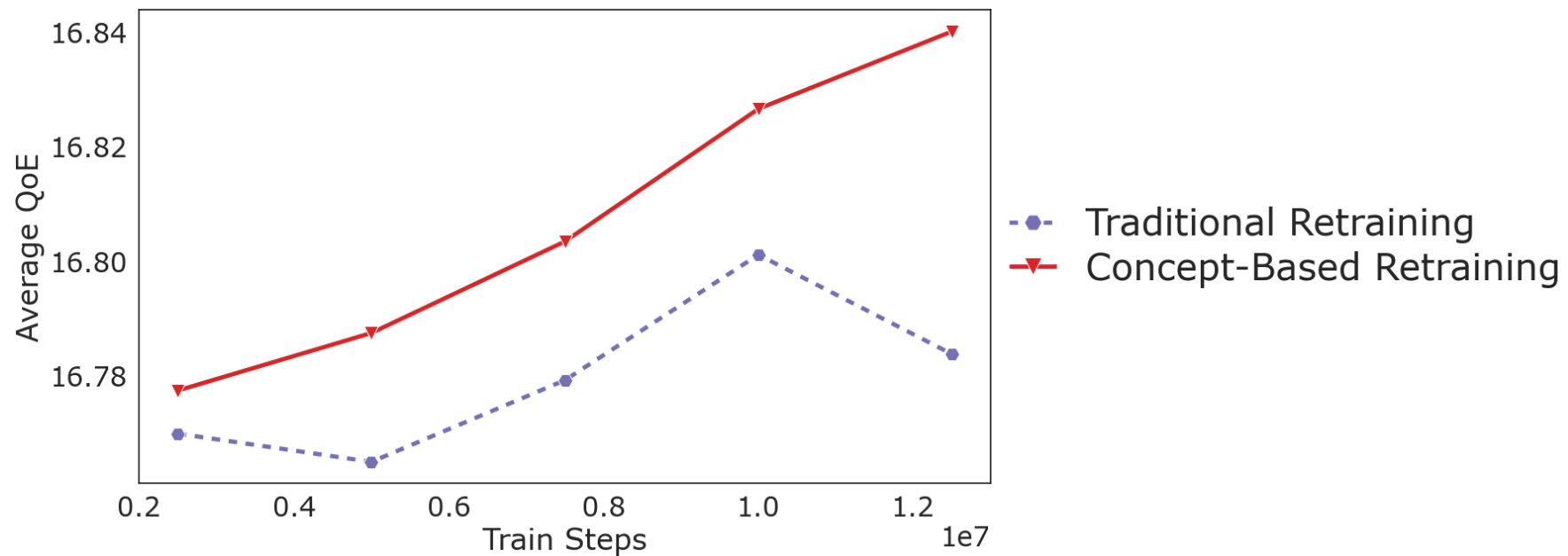
Use Cases: Distribution Shift Analysis

Gelato Datasets



Use Cases: Concept-Aware Retraining

Adaptive Bitrate Streaming Performance



Summary

- With Agua, we propose to move beyond feature-level explainability and towards an operator-aligned concept-level view
- Agua builds a surrogate concept-based model, attaining high fidelity and interpretability across
 - Congestion control, DDoS detection, adaptive video streaming
- Agua works with neural network controllers using supervised, unsupervised, and reinforcement learning
- We demonstrate Agua's ability to intuitively enable
 - Debugging, data shift detection, retraining, dataset expansion
- Code: <https://github.com/NetSAIL-UCI/agua>