

Special Topics: Machine Learning (ML) for Networking

COL867

Holi, 2025

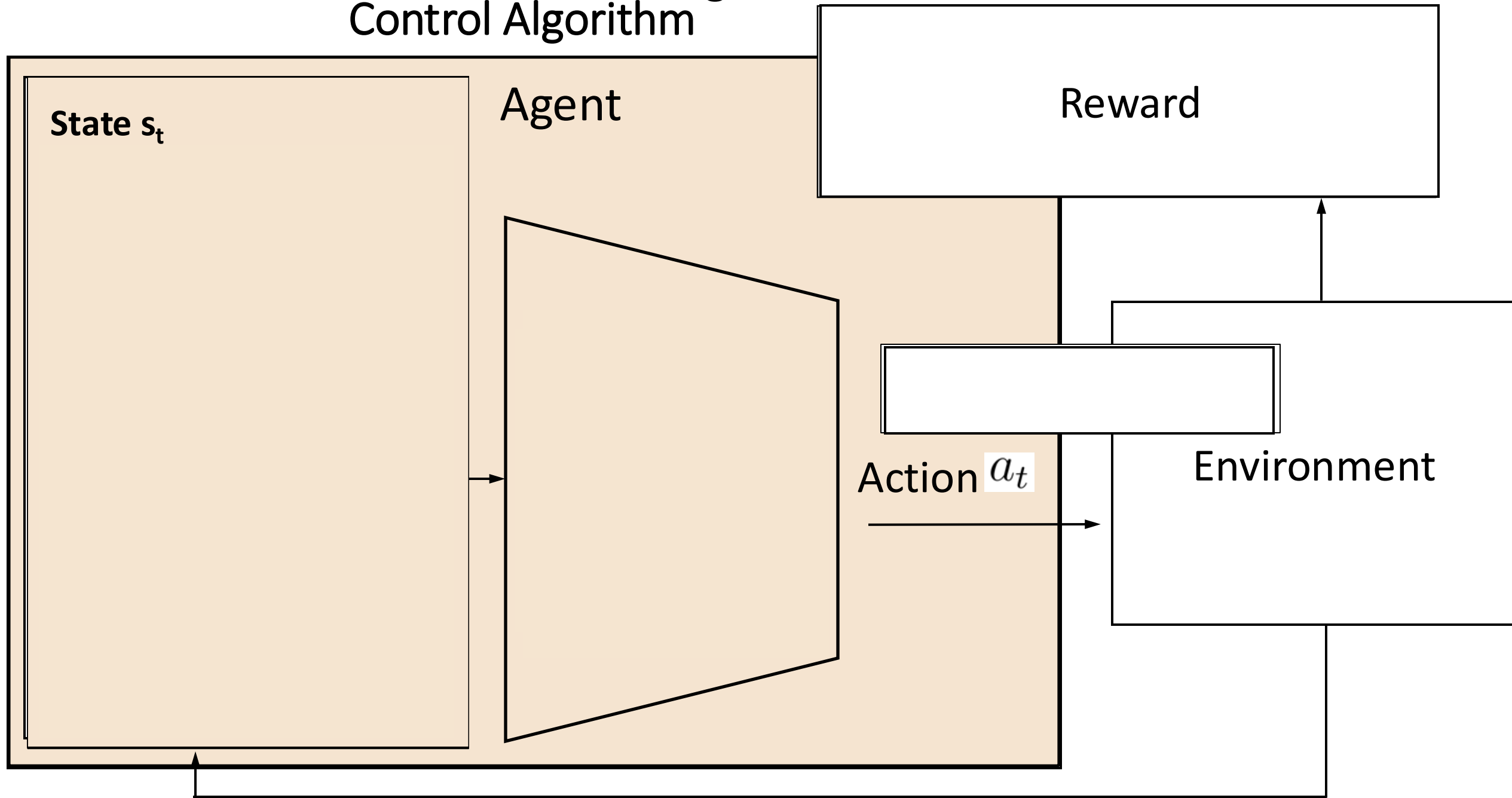
Resource Allocation

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Case Studies: ML for Specific Network Learning Tasks

- Application Classification
- Application Performance Monitoring
- Security
- **Resource Allocation**
 - Load balancing
 - Job scheduling
 - Network slicing
 - Rate control ... → Transport layer - CA

TCP Aurora: RL-based Congestion Control Algorithm

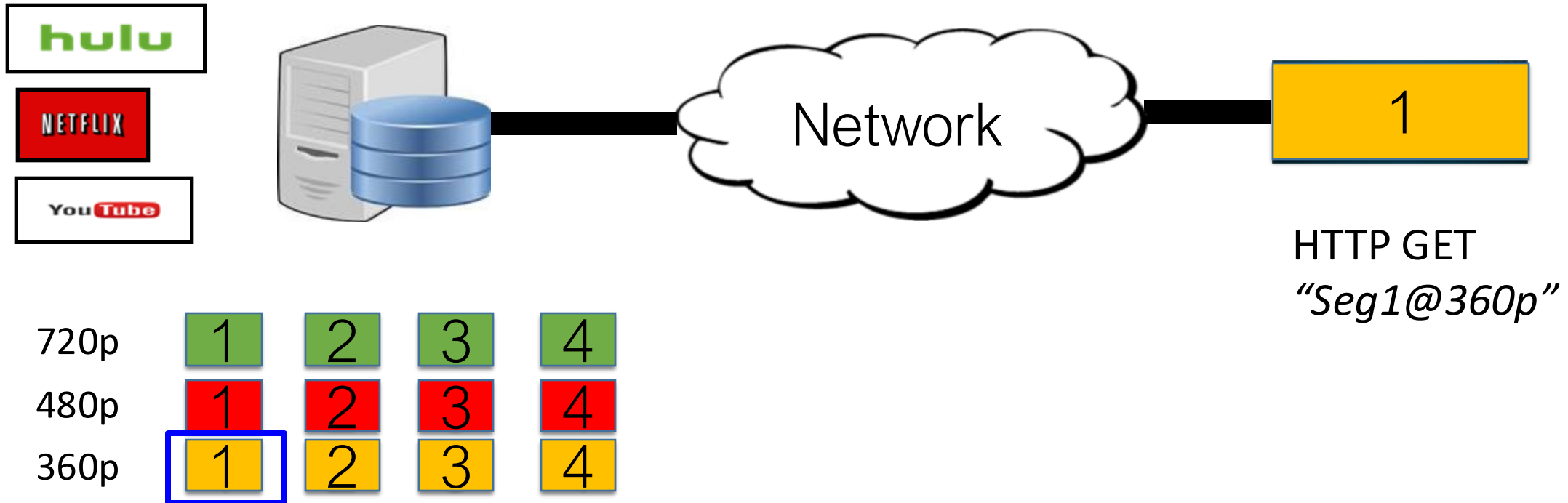


Rate Control (at End-host)

- Transport layer
- **Application layer: HTTP-based Adaptive Streaming**

Bitrate Adaptation

HTTP Adaptive Streaming (HAS) used for delivering Video over the Internet



Bitrate Adaptation

- **What:** Adapt player bitrate to Changing network conditions or changes in the viewing context (such as screen size, compute capability)
- **Why:** Support diverse clients and network conditions
- **How: ??**



low: 500 Kbps
needed \rightarrow 4 seconds | Download throughput: 1 Mbps

① Past chunk throughput: bps
② Remaining buffer

Goal of Bitrate Adaptation

- Bitrate adaptation aims to optimize the following objective metrics:

Minimize **Re-buffering**



High (2Mbps)
med (1Mbps)
low (0.5Mbps)

Maximize **average bitrate**



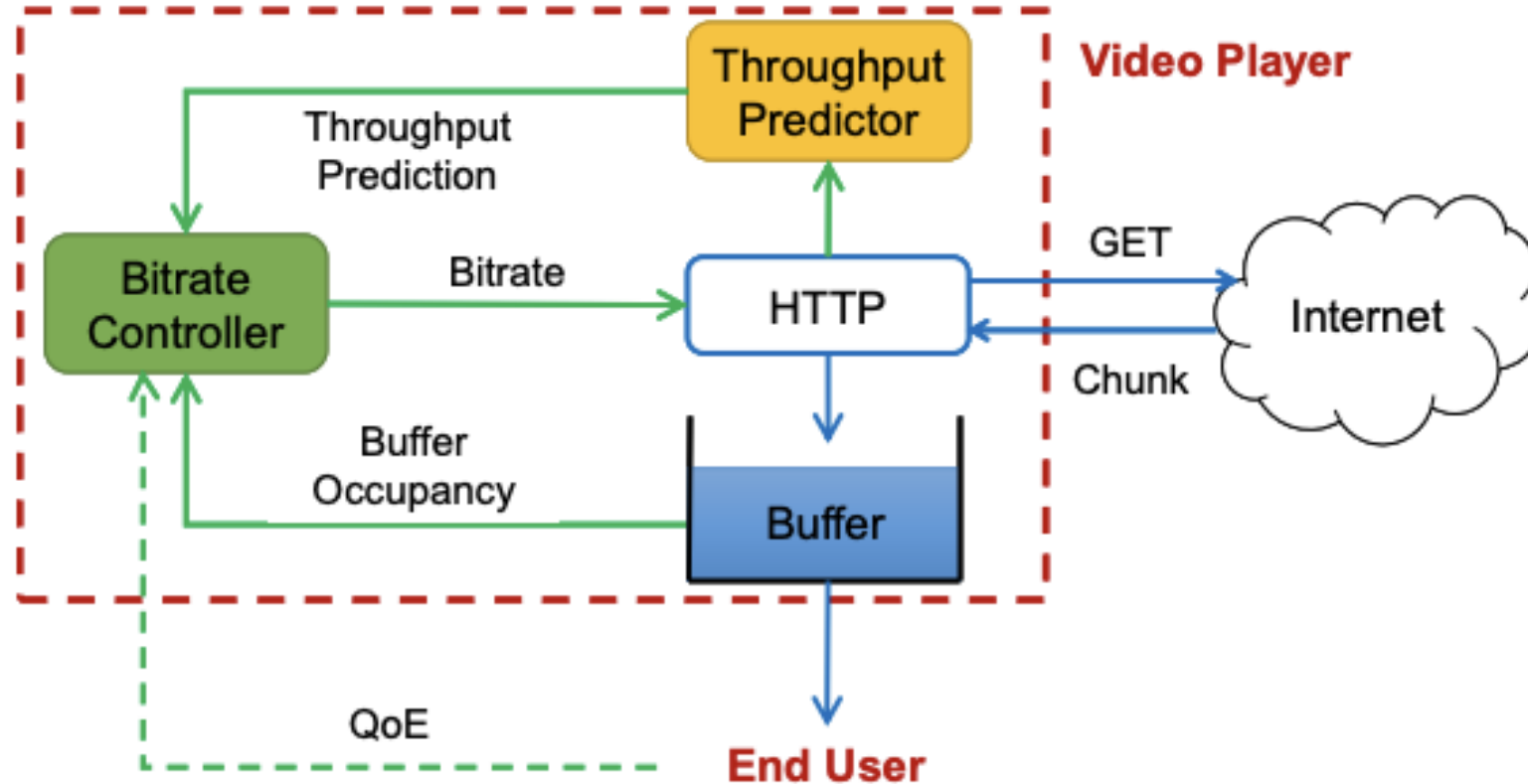
Minimize **bitrate switches**



Minimize **startup latency**



Abstract Player Model



Bitrate Adaptation

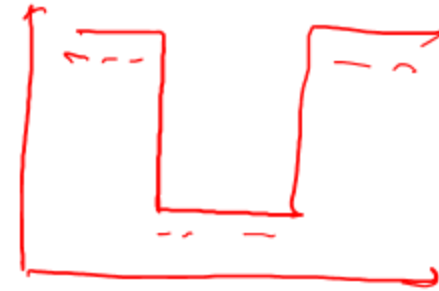
- What are the signals available to player?
 - Buffer-occupancy
 - Rate-based adaptation

Rate-based adaptation

- **Rate-based** (use past throughput)
- Assumption/Insight: Past throughput is a good indicator of future throughput
- Example heuristic:

$$\max_{r \in R} (1 + \alpha) \times r \leq T_{n-1}$$

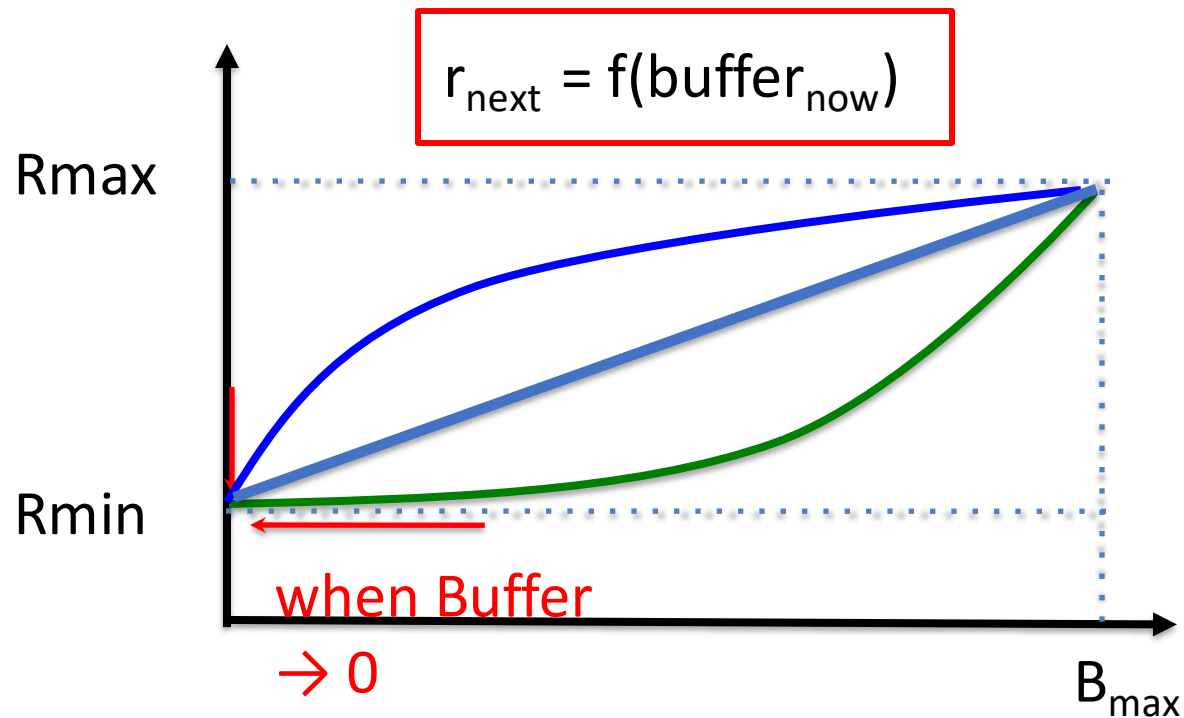
$\alpha \geq 0$



Buffer-based Adaptation

(Netflix)

- **Buffer-based** (use current buffer occupancy)





More Systematic Framework

- Model Predictive Control: a control technique that uses a model to predict the future behavior of a system, and then calculates the best actions to take
- Can we use MPC for HTTP-adaptive streaming?

① Future throughput

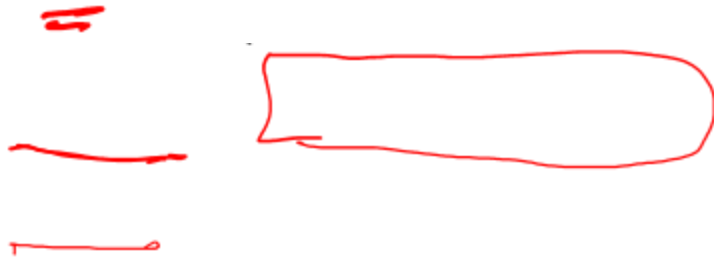
② Bitrate Select | Future throughput

$$B_K > B_{\max}$$

QoE Maximization Problem

$$\max_{R_1, \dots, R_K, T_s} QoE_1^K$$

Future throughput
is known
→



$$(\alpha) = \begin{cases} 0 & \text{if } x < 0 \\ x & \text{otherwise} \end{cases}$$

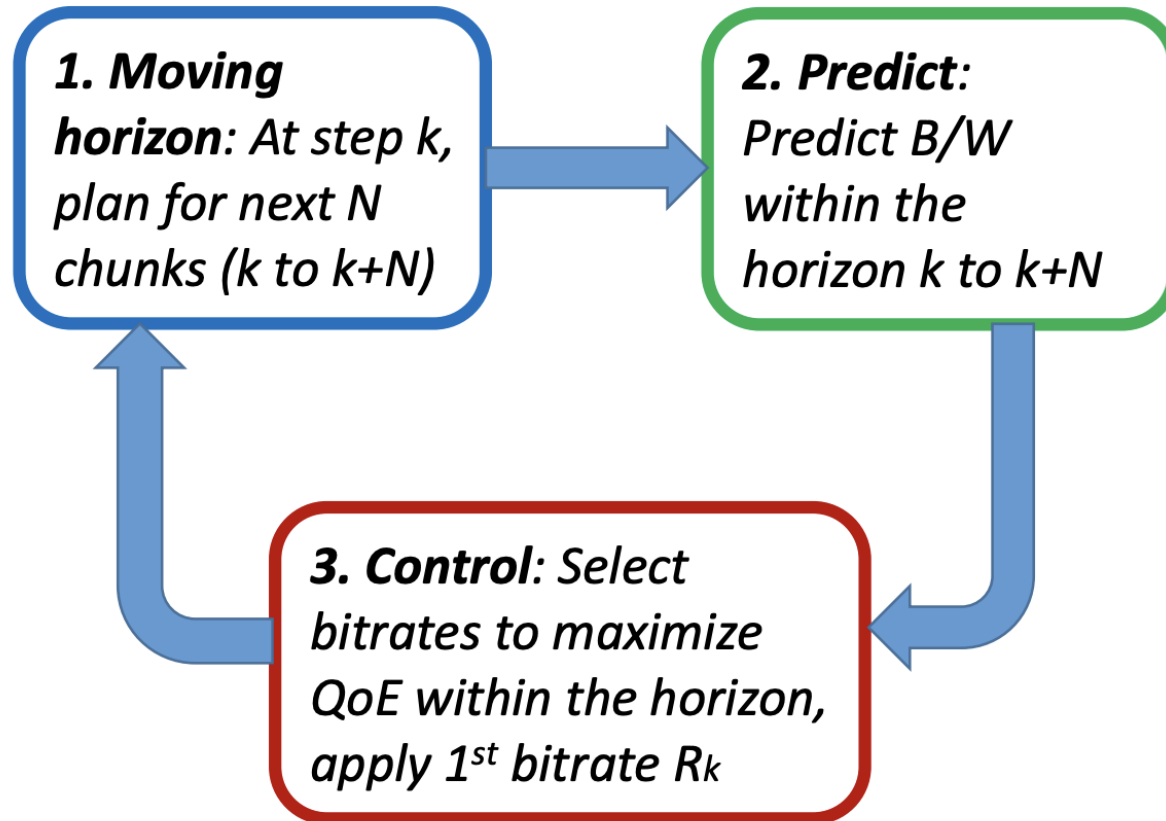
A diagram showing a vertical line with a double underline, followed by a horizontal line with a double underline, and then a vertical line with a double underline.

Buffer occupancy
before k^{th} chunk
has been done

last ~~Ad~~ chunk: 2Mbps

Model Predictive Control

Past throughput



Open question: How to predict the future bandwidth?
- How to incorporate errors in the control?

Bitrate Adaptation Heuristics

- Rate-based (use past throughput)
- Buffer-based (use current buffer occupancy)
- Assumption/Insight: Past throughput is a good indicator of future throughput

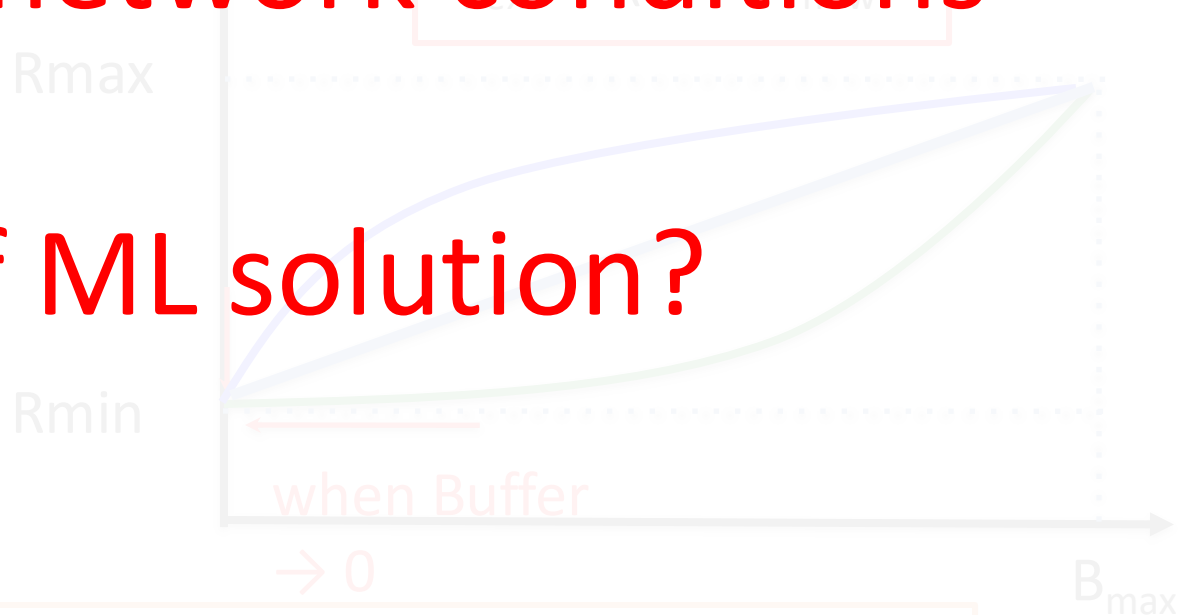
Solution: learn from video streaming sessions in actual network conditions

- Example heuristic:

$$\max_{r \in R} (1 + \alpha) \times r \leq I_{n-1}$$

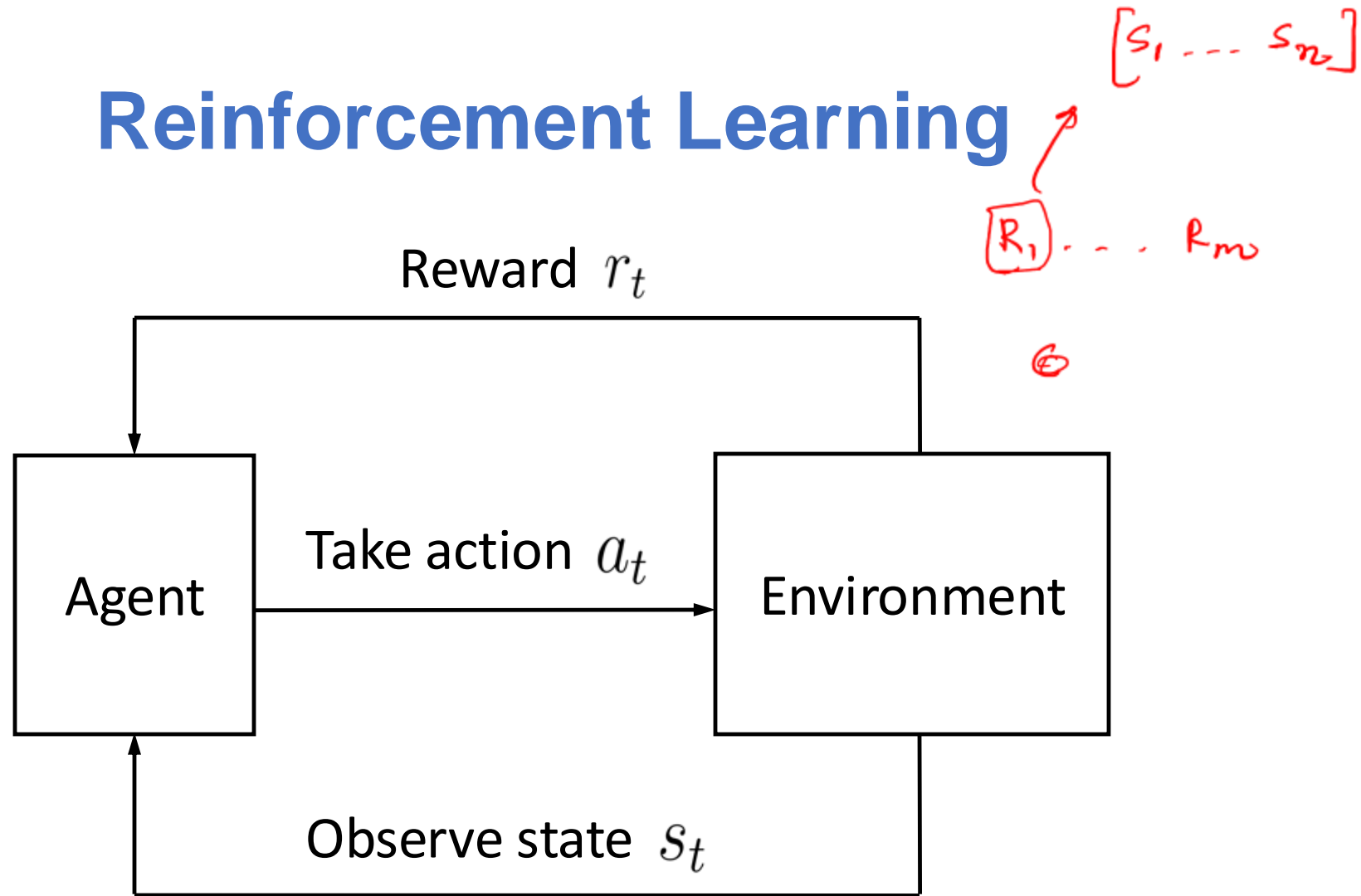
- Limitation: ?

What kind of ML solution?



Both these heuristics have limitations

Reinforcement Learning



Goal: maximize the cumulative reward $\sum_t r_t$

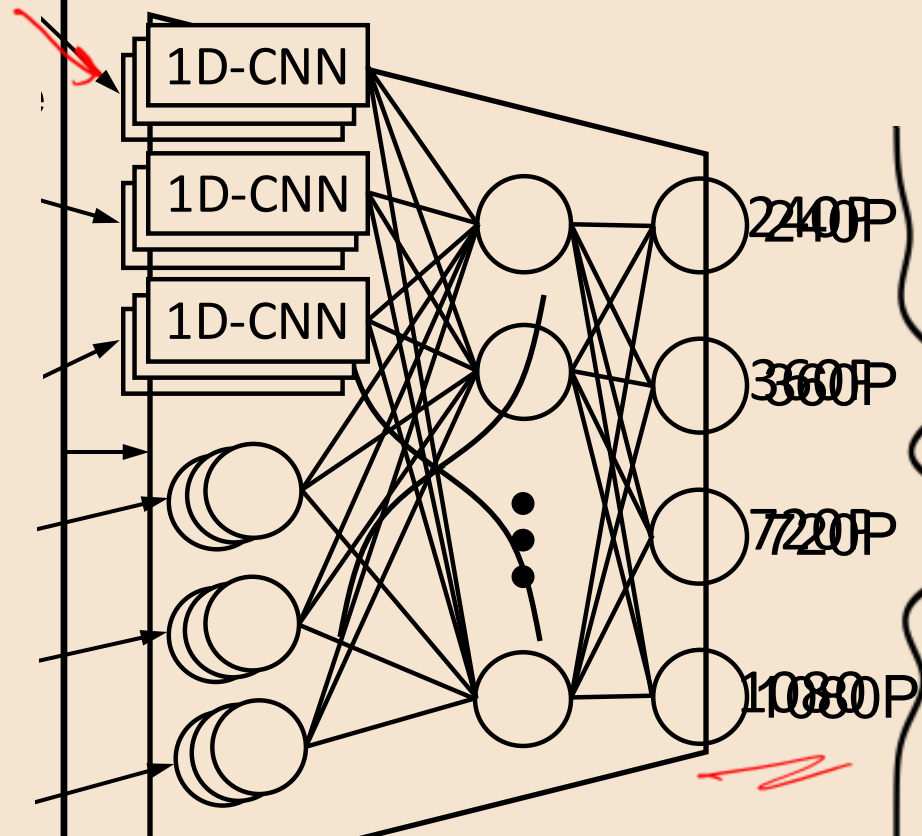
$$\frac{n_1}{L} \neq R_v$$

Pensieve Design

State s_t

Agent

Reward \hat{r}_t

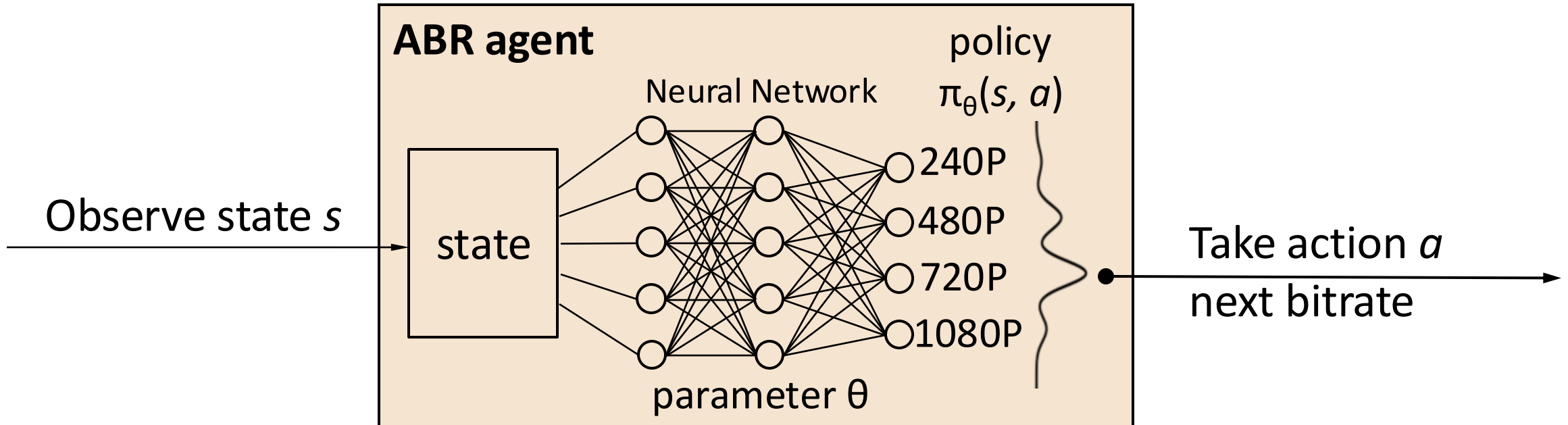


Action a_t

Environment

720P

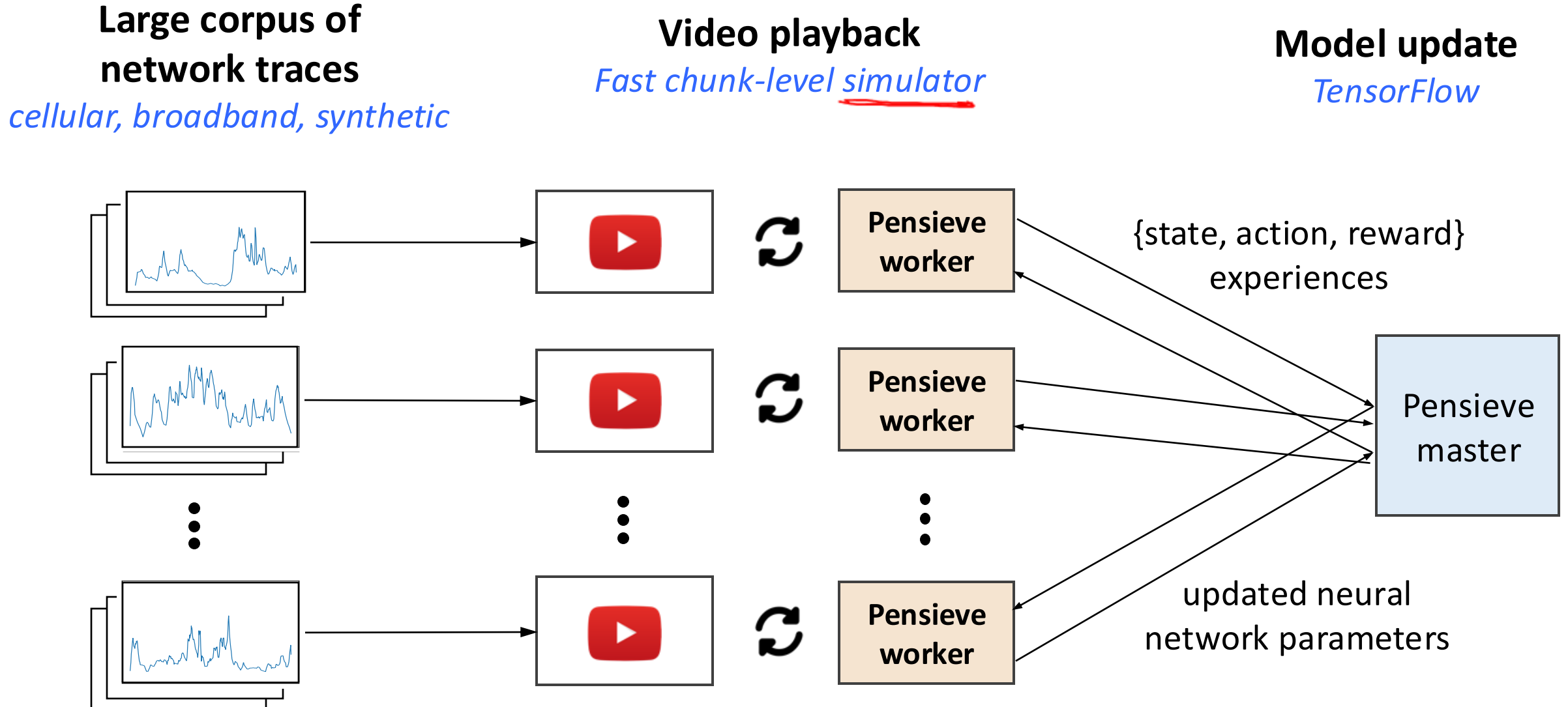
How to Train the ABR Agent



Collect experience data: trajectory of [state, action, reward]

Training: $\theta \leftarrow \theta + \alpha \nabla_{\theta} \mathbb{E}_{\pi_{\theta}} \left[\sum_t r_t \right]$ estimate from empirical data

Pensieve Training System



- ① Simulator
- ② Emulation
- ③ Real-world

QoE Breakdown



Pensieve reduces rebuffering by 10-32% over second best algorithm

Summary

- Pensieve uses Reinforcement Learning to generate ABR algorithms
- Optimizes bitrate selection by **directly learning** from different network conditions
- Uses trace-driven simulation to experience many network conditions
- Open question: Can it generalize to diverse network conditions?

Summary

- **Module 1:** Case studies of network learning problems
 - ① Security
 - ② Application perf monitor
 - ③ Resource Allocation
 - ④ Traffic classifier
- **Module 2:** Deep learning-based pipelines for network learning
 - **Generalized packet representation**
 - Foundation model
- **Next class: Generalized packet representation**
 - New Directions in Automated Traffic Analysis
- **No class on Friday. We will meet on Tuesday**