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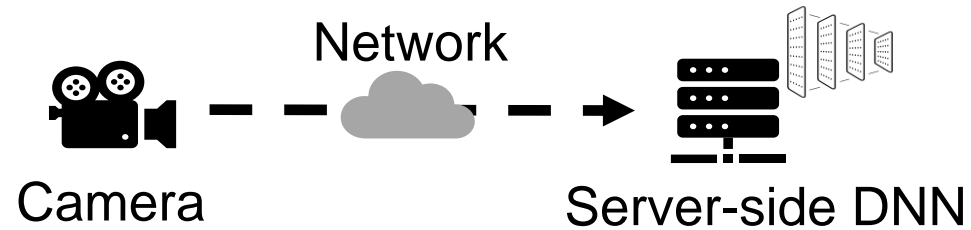
ACM Symposium  
on Cloud Computing

# Minimizing Packet Retransmission for Real-Time Video Analytics

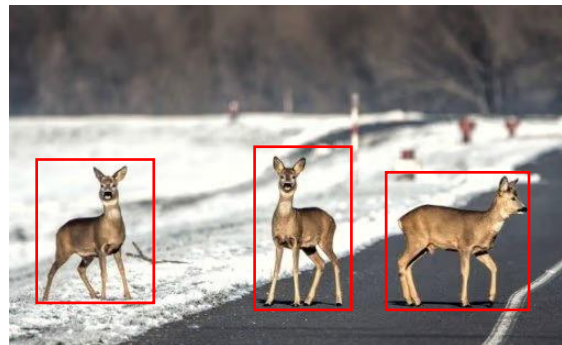
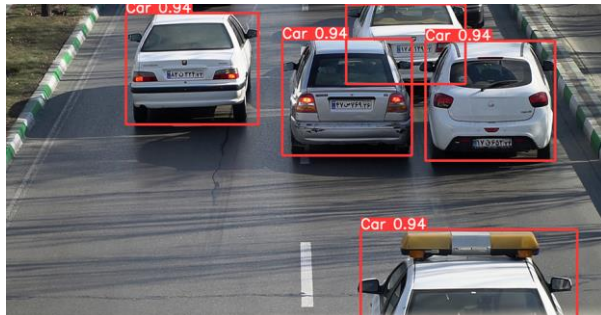
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University of Chicago

# High-quality video analytics (VA)

- In VA, videos collected by sensors are transmitted to cloud servers to run DNN-based inference

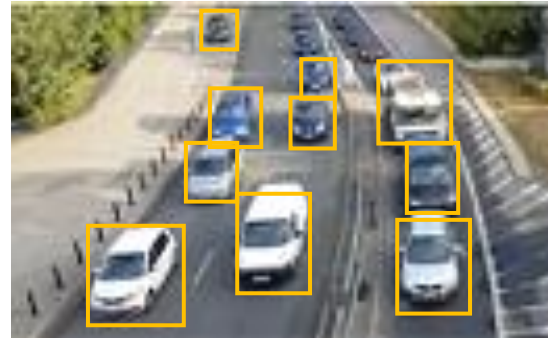


- It is used in many scenarios

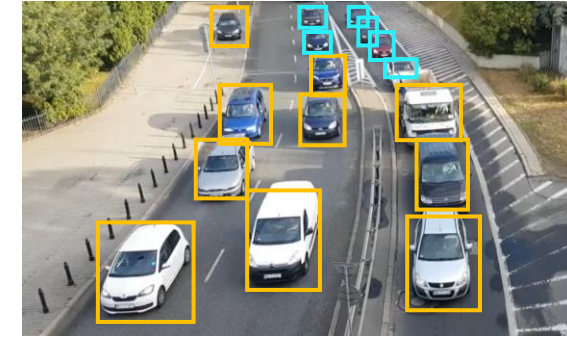


# VA requires high accuracy and low delay

- High accuracy: the analysis results are close to that using uncompressed videos
  - Example: Car detection



Using a low-quality video



Using an uncompressed video

- Low delay: we can get the results in near real time



Video conferencing: <100ms



Augmented reality: <110ms

# Our Idea

## *Application-layer designs*

To reduce packet retransmission by only sending the most relevant frames to applications determined before transmission

## *Traditional Transport-layer designs*

To reduce packet retransmission using additional information irrelevant to applications and generated before transmission

## *Our transport-layer design: T4V*

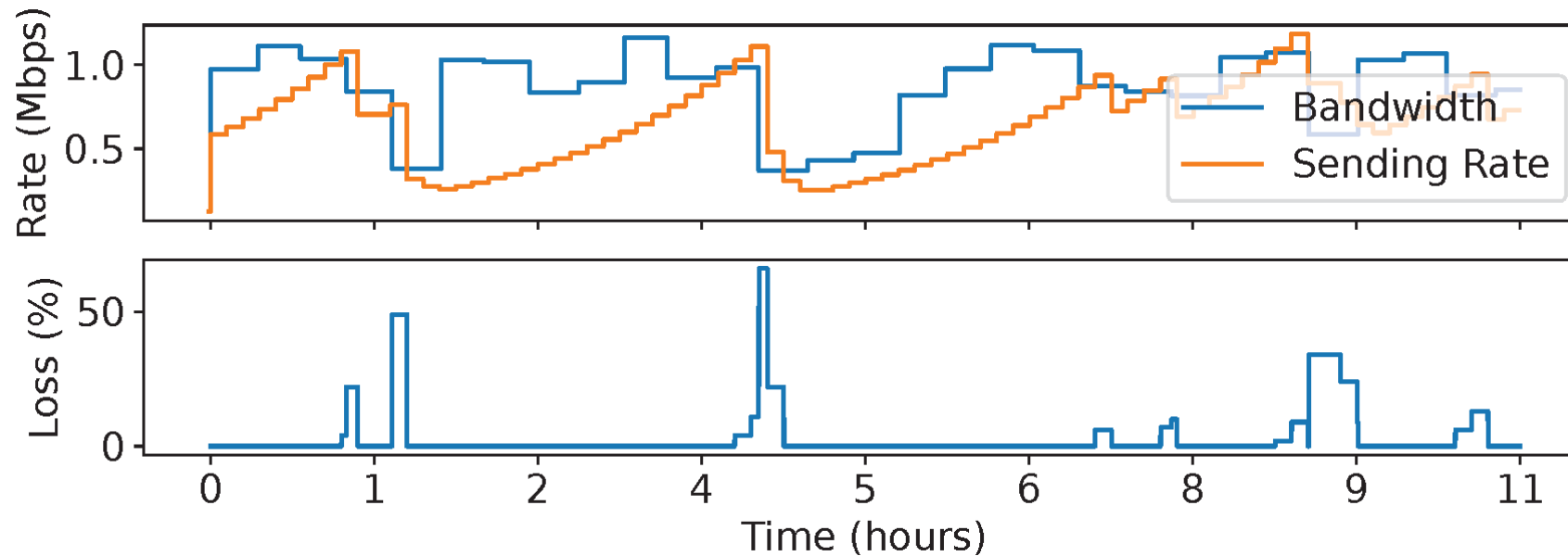
To reduce packet retransmission using application-aware additional information determined during transmission

# Application-layer designs

- Idea
  - To aggressively compress video frames, and only send the most relevant frames
- Examples
  - AStream [SIGCOMM'18]
  - DDS [SIGCOMM'20]
  - Reducto [SIGCOMM'20]
- Drawbacks
  - ***Reducing delay means sending fewer bits?***
  - ***The impact of each video frame on DNN inference can be precisely determined before transmission?***

# Sending fewer bits?

- Sending fewer bits does ***not necessarily*** reduce the delay
  - Lowering bitrate can't eliminate transient ***packet losses***.
  - WebRTC Example: Bandwidth drops can cause transient packet losses because of the hysteresis of sending rate adjustment



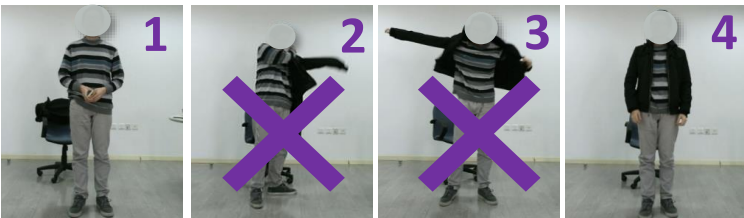
# Estimating the impact of video frames?

- The estimation of video frame impact on DNN inference ***before*** transmission is ***inaccurate***
  - Video frame impact can ***only*** be precisely obtained ***during*** transmission


# Estimating the impact of video frames? (cont.)

Example: Video frame impact can **only** be precisely obtained **during** transmission

**Case 1**  
Frame **2, 3** lost  
Inference result: standing ❌




Frame 2 has a **high** impact  
Retransmit frame 2 ✅




Inference result: putting on a coat ✅

**Case 2**  
Frame **1, 2** lost  
Inference result: putting on a coat ✅



Frame 2 has a **low** impact  
Not retransmit frame 2 ❌



Inference result: putting on a coat ✅



# Application-layer designs (revisited)

- Idea
  - To aggressively compress video frames, and only send the most relevant frames
- Examples
  - AStream [SIGCOMM'18]
  - DDS [SIGCOMM'20]
  - Reducto [SIGCOMM'20]
- Drawbacks
  - Sending fewer bits does ***not necessarily*** reduce the delay ☒
  - The estimation of video frame impact on DNN inference ***before*** transmission is ***inaccurate*** ☒

# Traditional Transport-layer designs

- Idea
  - To reduce packet retransmission by additional information
- Examples
  - Forward-error correction (FEC)
  - Bounded-loss transport
  - Selective retransmission
- Drawbacks
  - The additional information is still determined **before** transmission ☒
  - The additional information is **irrelevant to** the frame loss impact on DNN inference ☒

# Our Idea (revisited)

## *Application-layer designs*

To reduce packet retransmission by only sending the most relevant frames to applications determined before transmission

## *Traditional Transport-layer designs*

To reduce packet retransmission using additional information irrelevant to applications and generated before transmission

## *Our transport-layer design: T4V*

To reduce packet retransmission using application-aware additional information determined during transmission

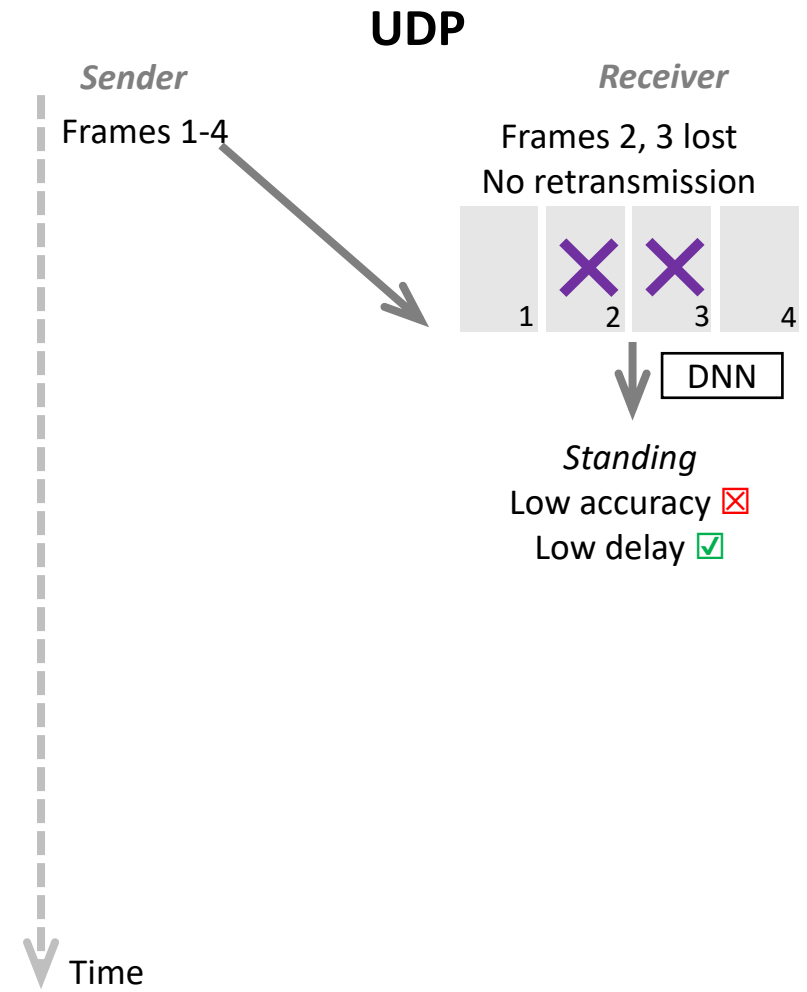
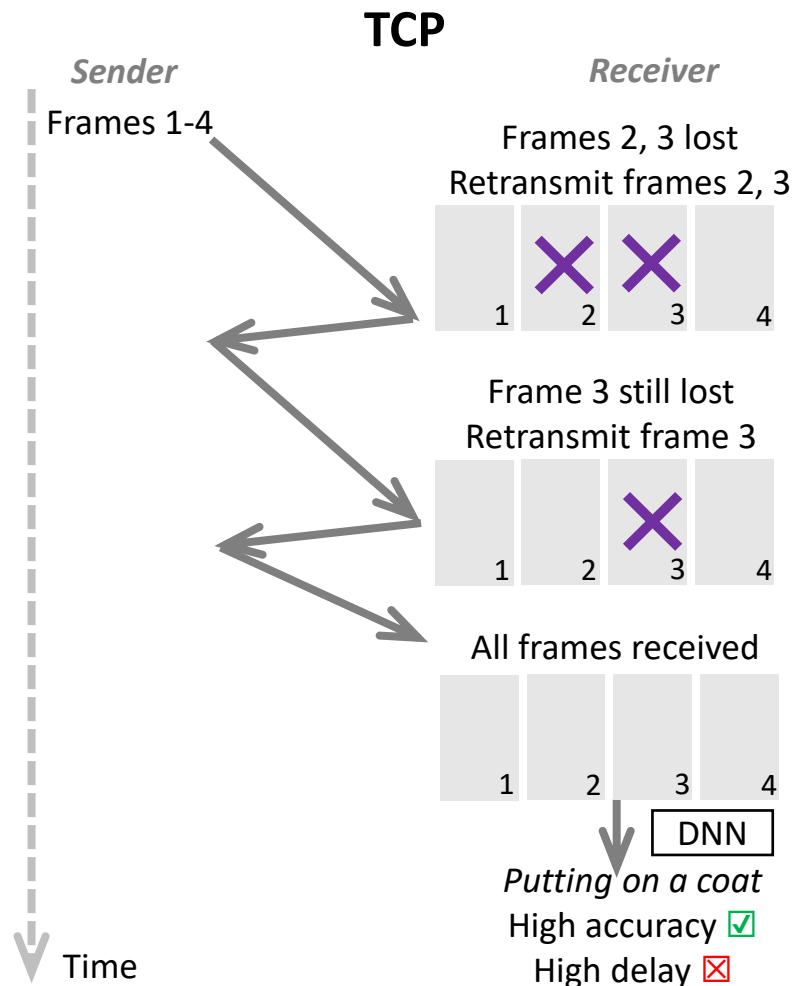
# Improvement brought by our idea



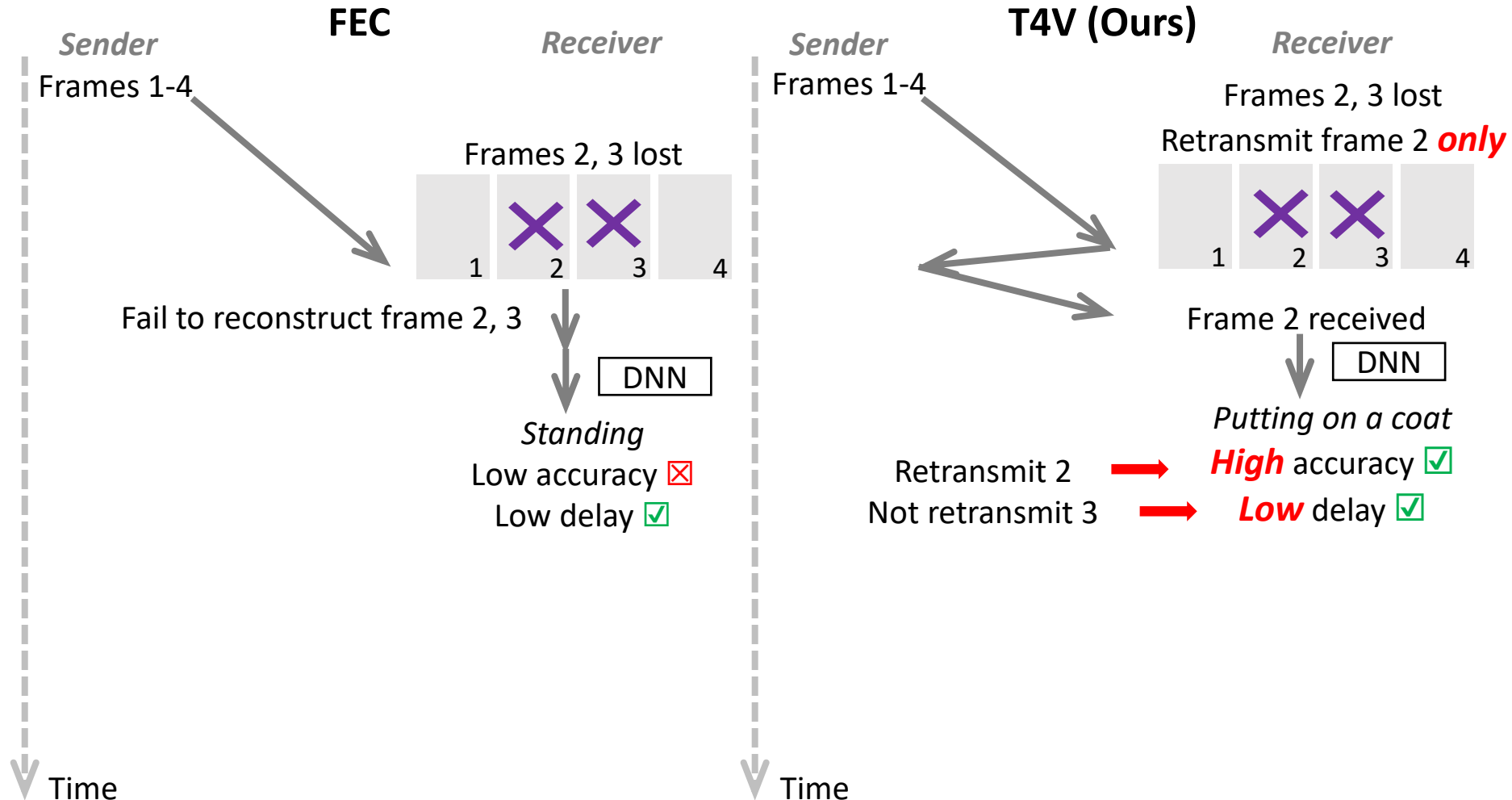
2 or 3 received:  
*putting on coat* ✓

2 and 3 lost:  
*standing* ✗

Frame loss rate: 50%



# Improvement brought by our idea (cont.)

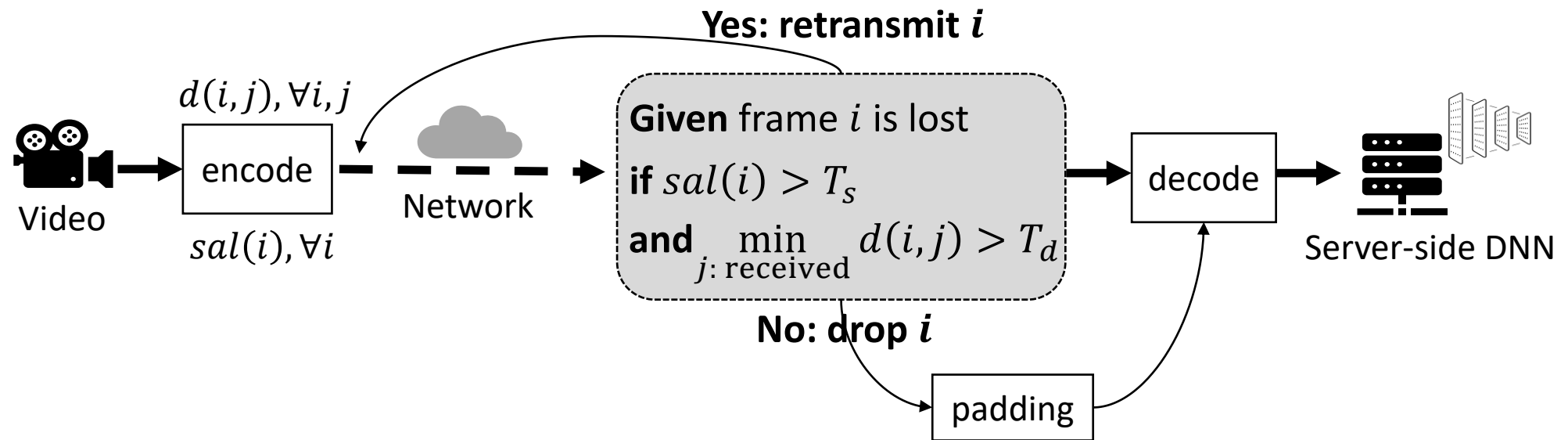


# Design of T4V

- Key Idea: ***Incremental impact*** of each frame ***conditioned on*** received frames
  - Definition: *Given the received frames*, how much obtaining or losing a frame would change the inference result
  - Components
    - Frame difference: Pixel-wise difference between frames
    - Saliency: Pixel-wise accumulation of the gradient of the inference result with respect to the frame **[Open Question 1]**

# Design of T4V (cont.)

- How to use the incremental impact
  - $d(i, j)$ : frame difference between frame  $i$  and  $j$
  - $sal(i)$ : saliency value of frame  $i$
  - $T_s, T_d$ : user-defined thresholds



[Open Question 2]

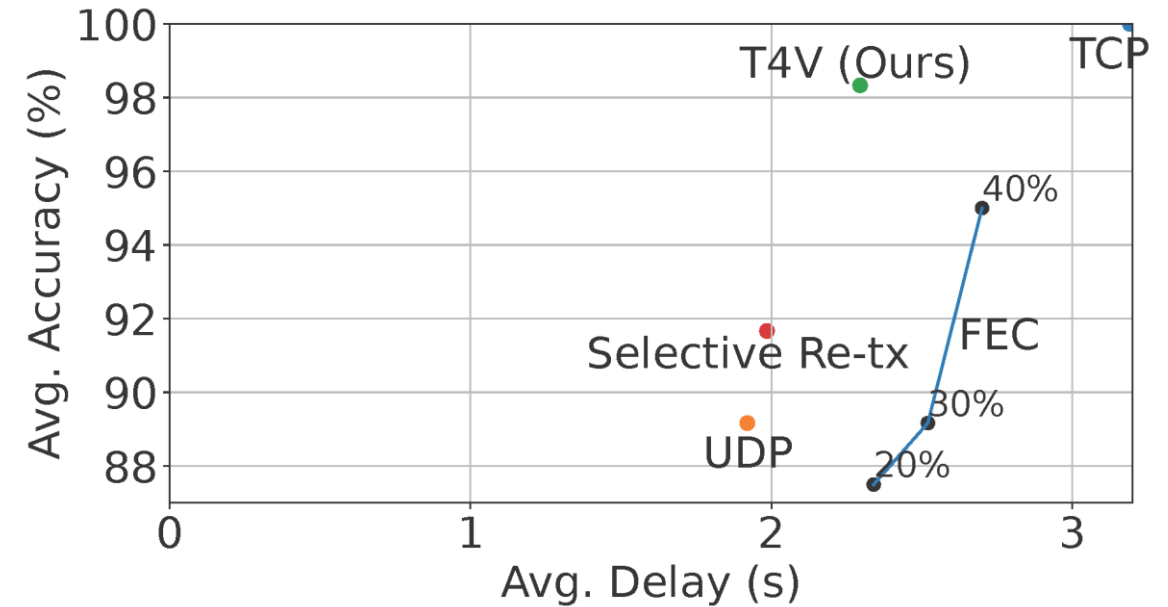
# Open Questions

- Saliency estimation
  - The **overhead** to get **accurate** saliency values is high (forward propagation and backward propagation on a large DNN)
  - Direction: saliency values can be reasonably **approximated** by training **cheap predictors**
- Faster retransmission decisions
  - Retransmission **decisions** require nontrivial **computation**
  - Direction 1: to **offload** some compute to sensors
  - Direction 2: to **pipeline** packet retransmission with DNN inference on the received frames



# Case Study

- VA application: action recognition
- DNN: I3D
- Network simulation
  - Streaming delay = size of transmitted packets / bandwidth (200Kbps)
  - Each frame is sent in one packet (1.5KB, consistent with the average frame size in a low-quality video (e.g., 360p))
  - Frame loss rate: 30%
  - $T_s = 0.01$  and  $T_d = 0.001$
  - 100 rounds of independent tests
- Data: 12 video clips from Kinetics-400
  - 32 frames per video clip
- Baselines: TCP, UDP, FEC, and H.264-based selective retransmission



- T4V vs. TCP: Similar accuracy with 30%+ less packet retransmissions
- T4V vs. UDP: inaccuracy reduced from 11% to 2% at marginal delay inflation (15%)
- T4V vs. selective retransmission: reduces inaccuracy from 8% to 2% with only 10% delay increase

# Conclusions

- We propose a transport-layer design, T4V, for real-time video analytics.
- T4V makes a case for deciding whether to retransmit a frame based on its incremental impact on inference output conditioned on received frames.
- Our contribution is a framework to make retransmission decisions based on the incremental impact per frame, and a case-study evaluation to quantify its potential benefit.