

Optimizing Live Game Streaming Platforms Using Segment-of- Interests

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Outline

- Introduction & Motivation
- System Overview
- Sol Detector
- Resource Allocator
- Evaluation
- Conclusion

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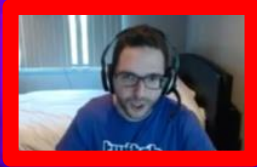
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Twitch



Pokemon HG Shiny Badge Quest | Double RE's fo...
GalacticElliot playing Pokémon Omega Ruby/Alpha Sapphire on National Po...

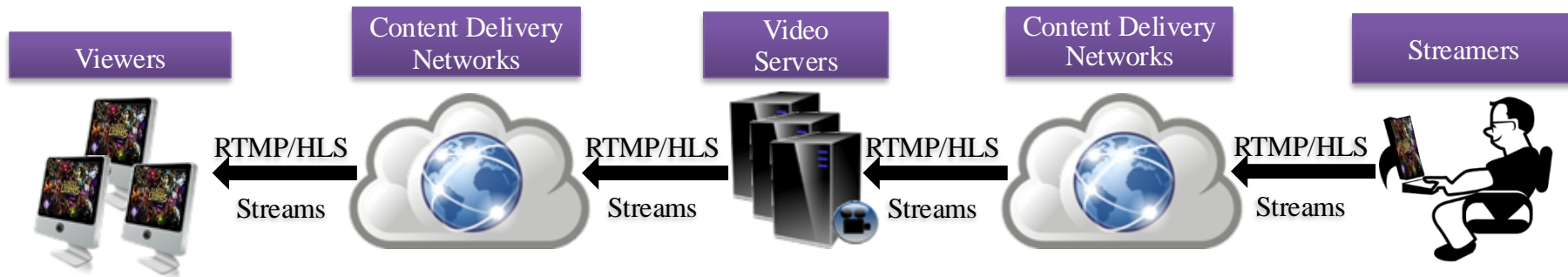
Gameplay



Webcam
image

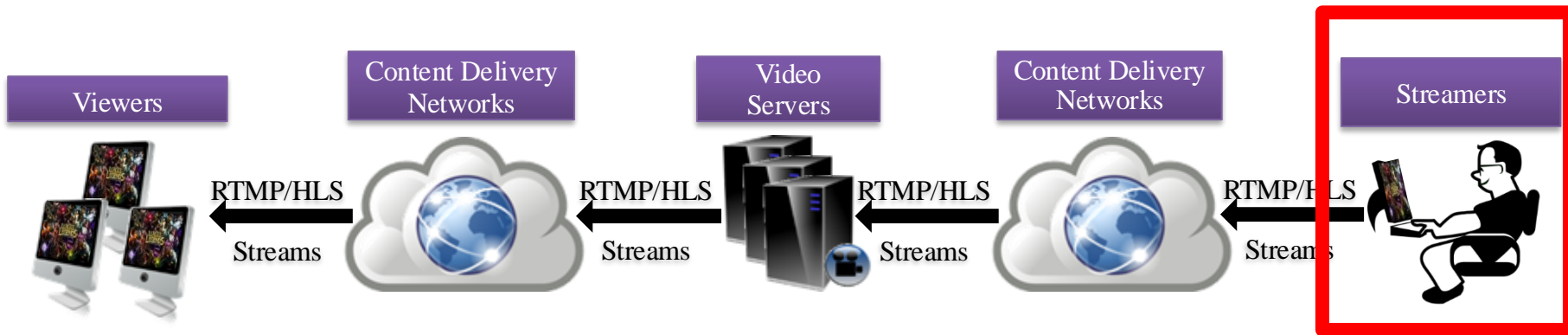
Chat

How Does Live Game Streaming Work?



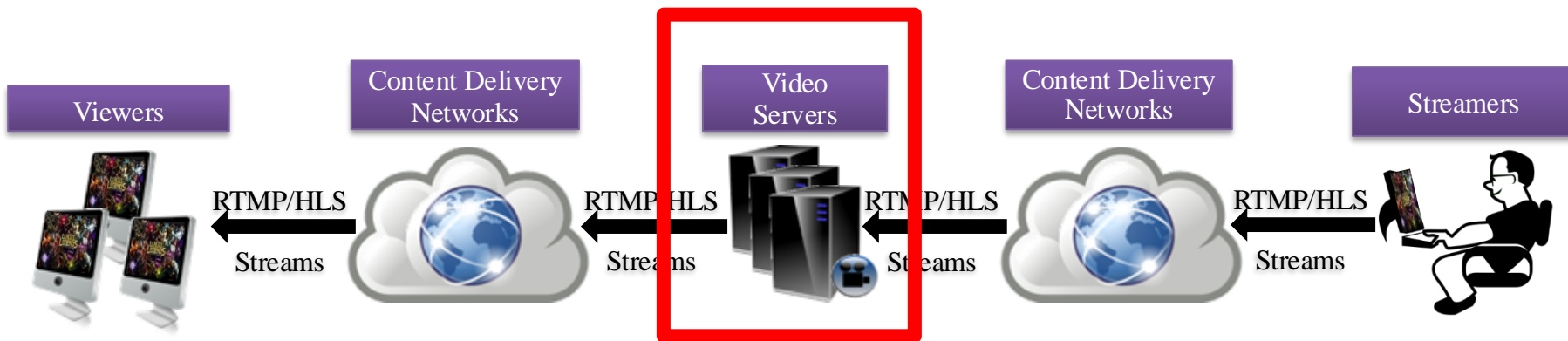
Streamer

- Streamer: send the game stream to video servers
 - Often with streaming software such as OBS
 - Using protocols such as RTMP, HTTP Live Streaming



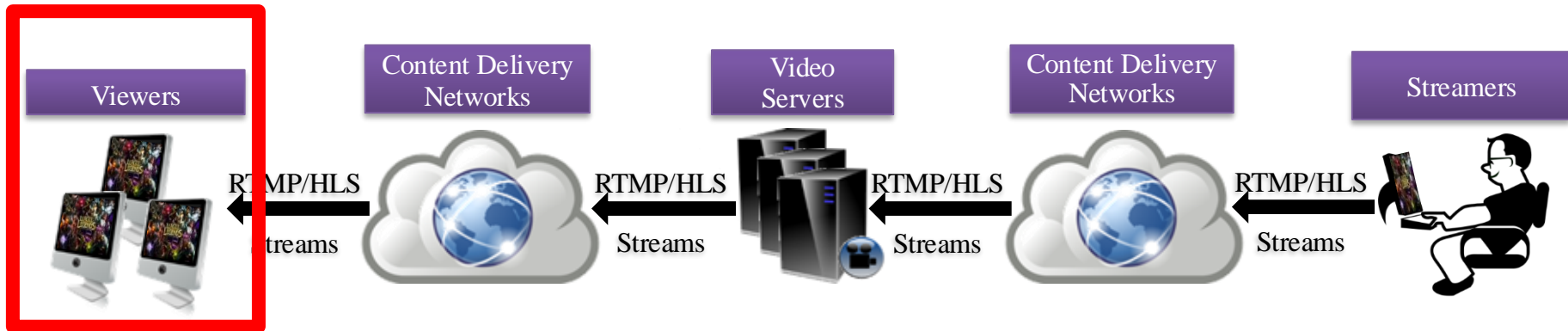
Video Servers

- Relay the stream to viewers
 - Transcode them if necessary



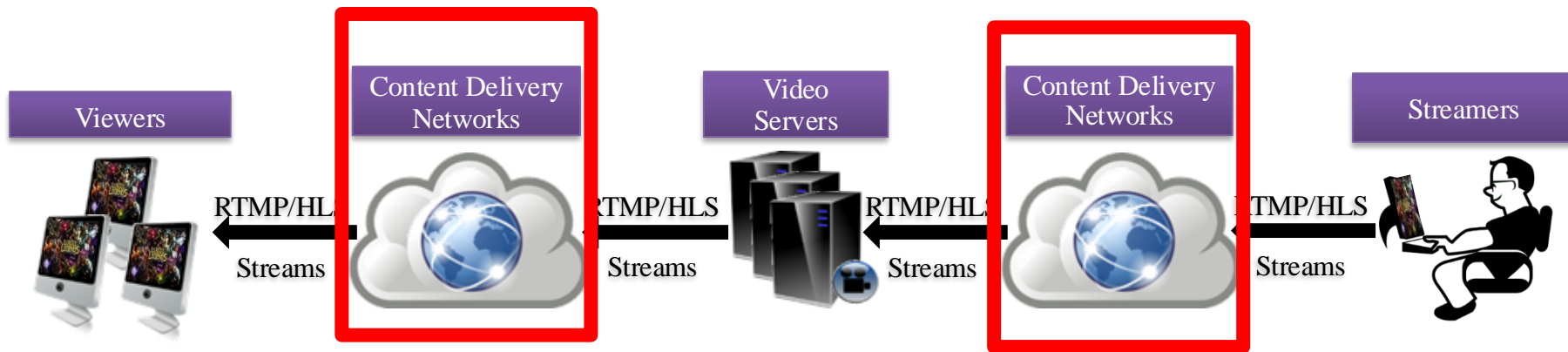
Viewers

- Watch the stream with video players



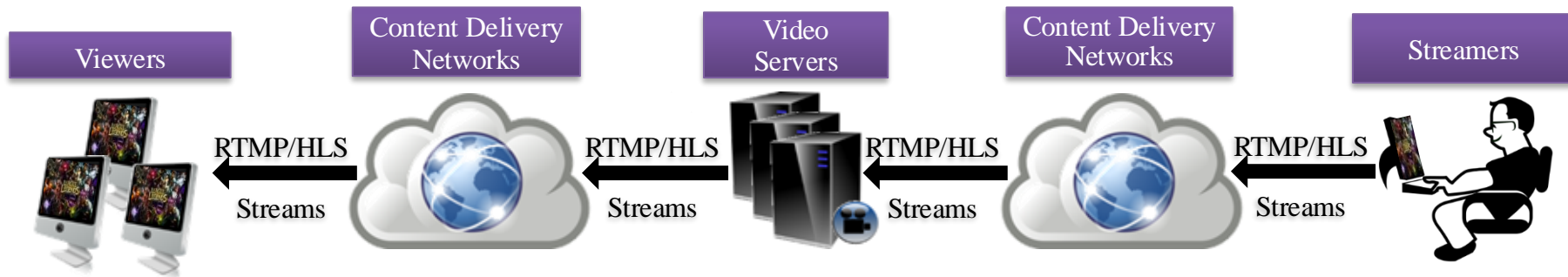
Content Delivery Networks

- Streams often go through CDNs
 - Companies such as Twitch need to pay for the bandwidth (to CDN companies such as Akamai)



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Who are running these services?

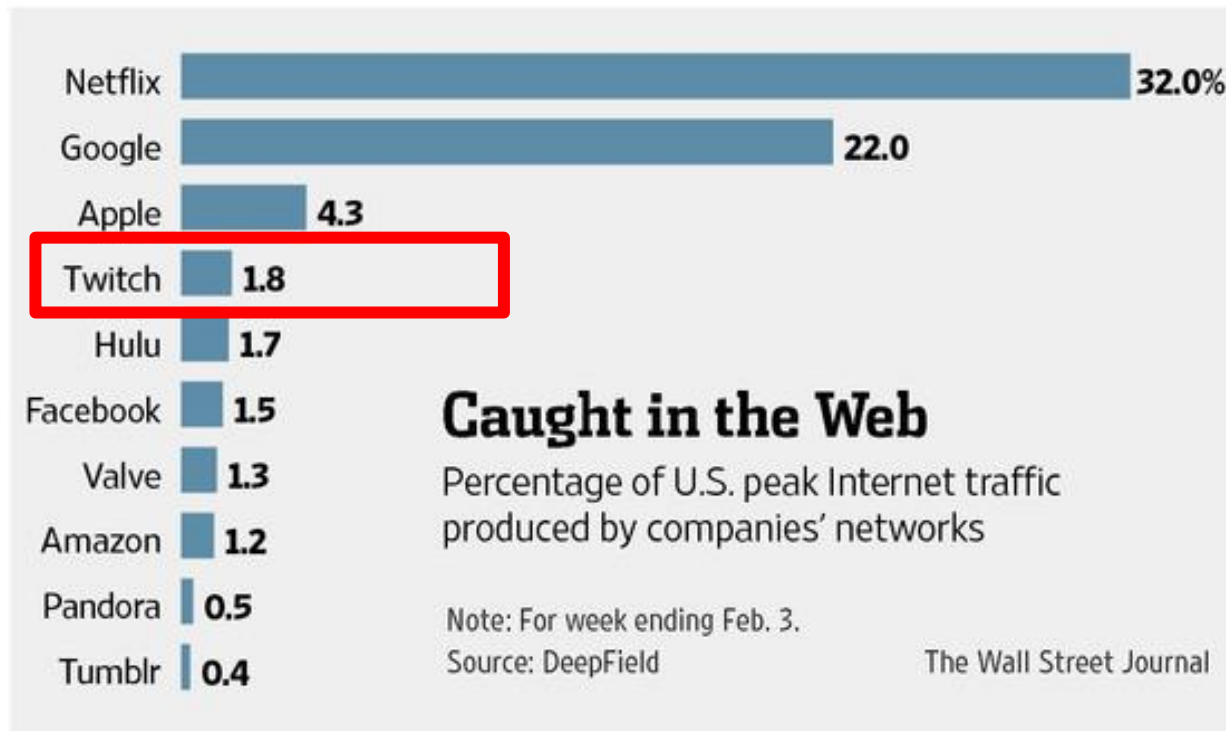
Who Are Running These Services?

- Twitch
 - Spin-off from Justin.tv
 - The broadcast platform for most E-sport event
- Ustream
 - The streaming service used by Sony PS4
- YouTube Gaming
 - Launched in Aug, 2015



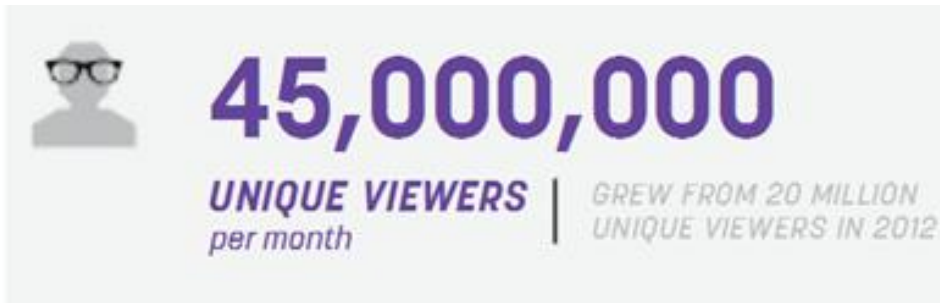
Twitch Is Hot

- 4th largest in peak US Internet traffic

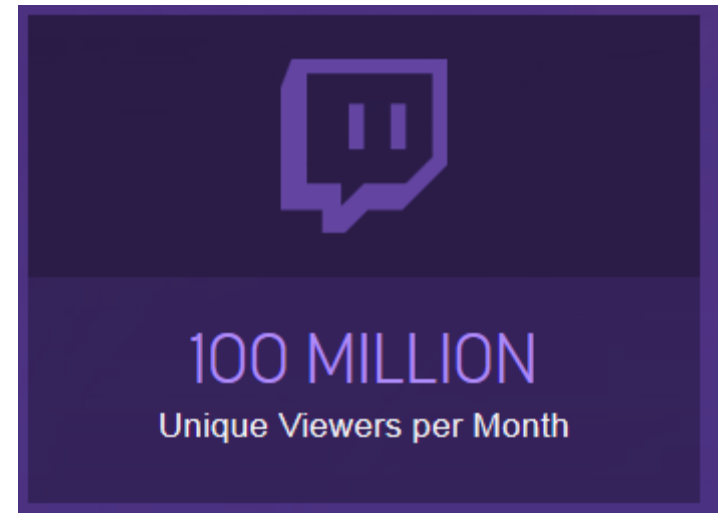


Growing Demands

- Number of viewers doubled in 2014
 - More than twice in 2013



2013



2014

Acquired by Amazon.com for \$970 million

Everything looks great

WHAT'S THE PROBLEM?

Problem: Lag

- Lag is annoying when watching live streams
 - Rebuffering
 - Other visual artifacts
 - Skipped video segment
- Bad for user experience

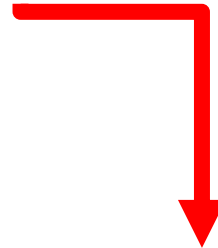


Possible Solutions

- Put in more resource
 - Cost money
- Lower the video quality
 - Bad for user experience

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- Put in more resource
 - Cost money
- Lower the video quality
 - Bad for user experience



How to do this without degrading the user experience?



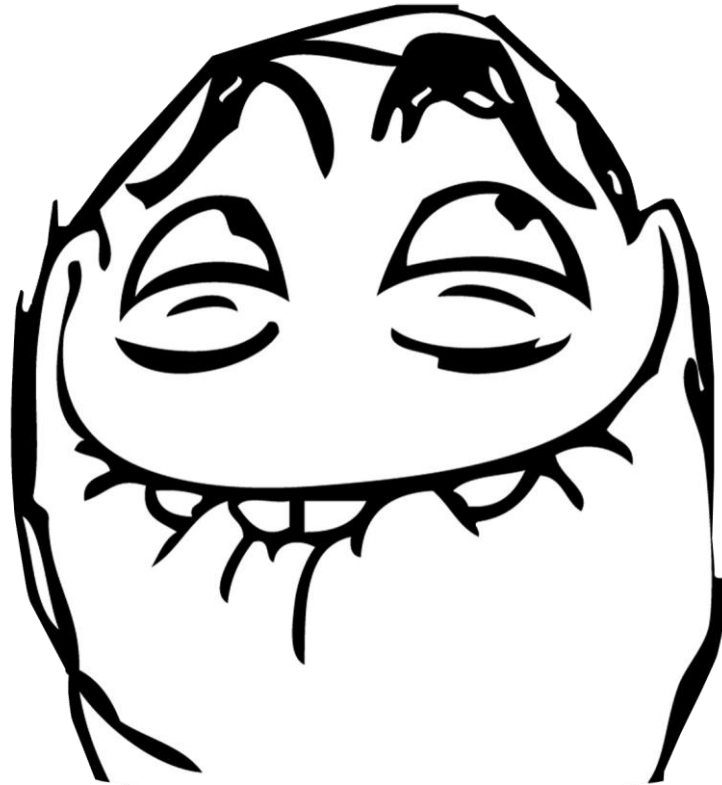
LAG AT DIFFERENT TIME CAUSE DIFFERENT LEVEL OF DEGRADATION IN USER EXPERIENCE



At Clutch Time



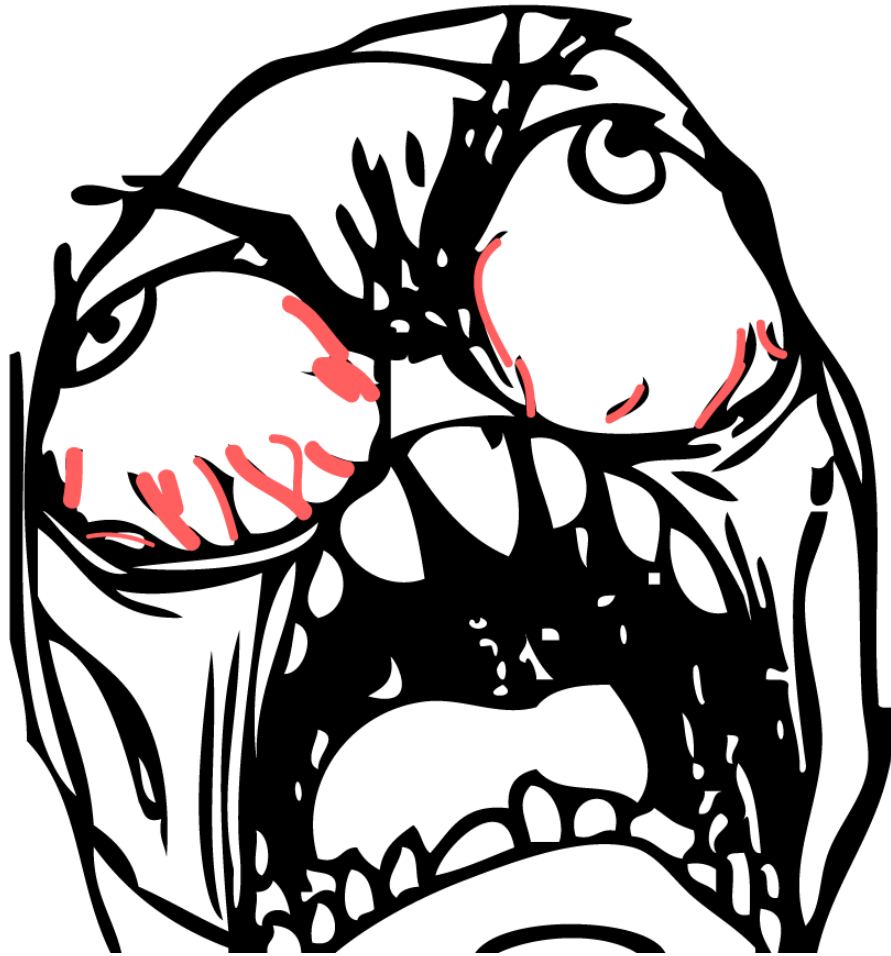
At Clutch Time



At Clutch Time



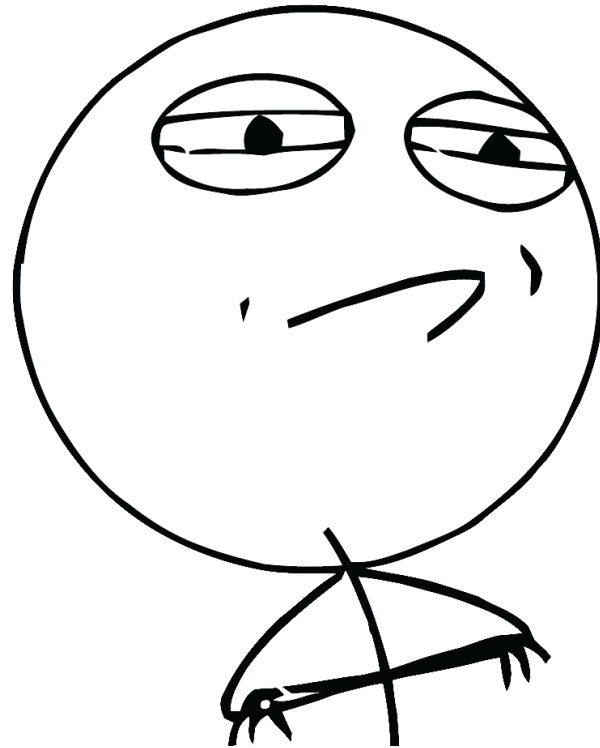
At Clutch Time



At Normal Time



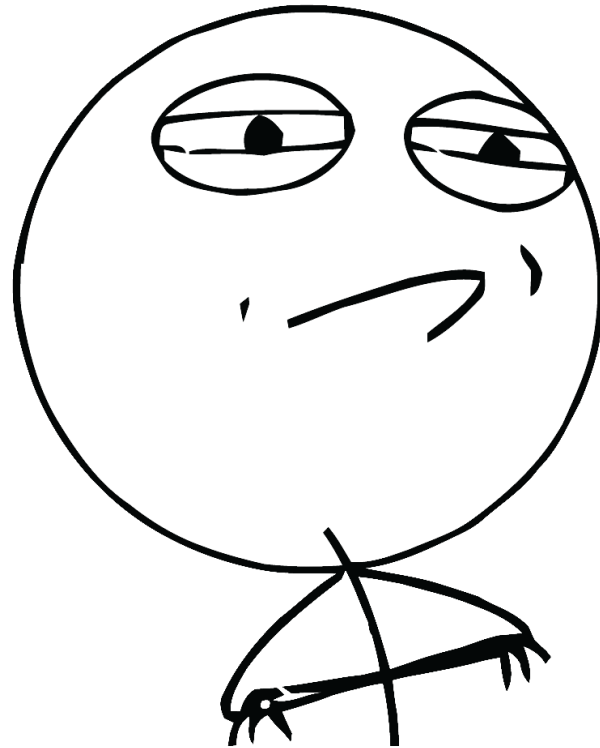
At Normal Time



At Normal Time



At Normal Time



Not that bad

Concept of Segment of Interest (Sol)

- Different segments have different importance
 - We refer to the important one as **Segment of Interest (Sol)**
- We should put in more resource to deliver Sol
 - Sacrifice non-Sol if we have to



Contributions

- Develop Sol-driven streaming platform
- Answer two questions
 - How to detect Sol efficiently?
 - How to allocate resources among channels?

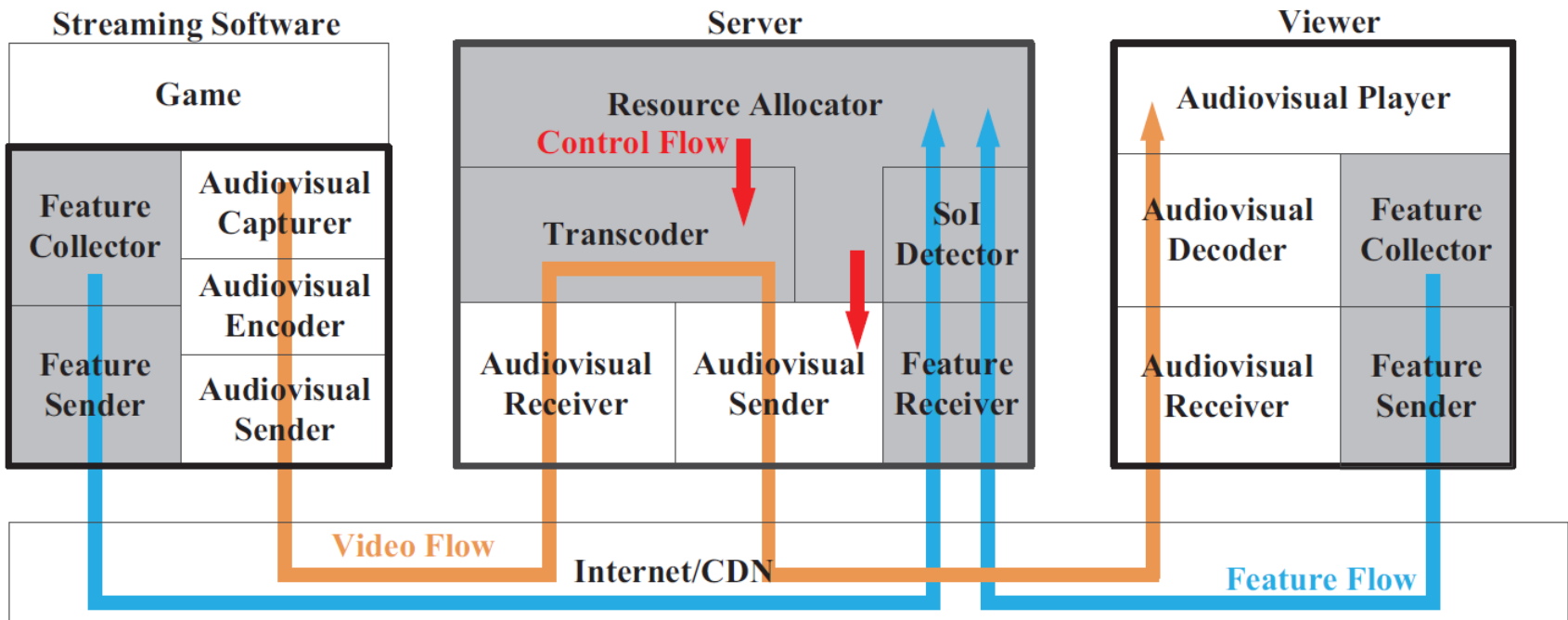
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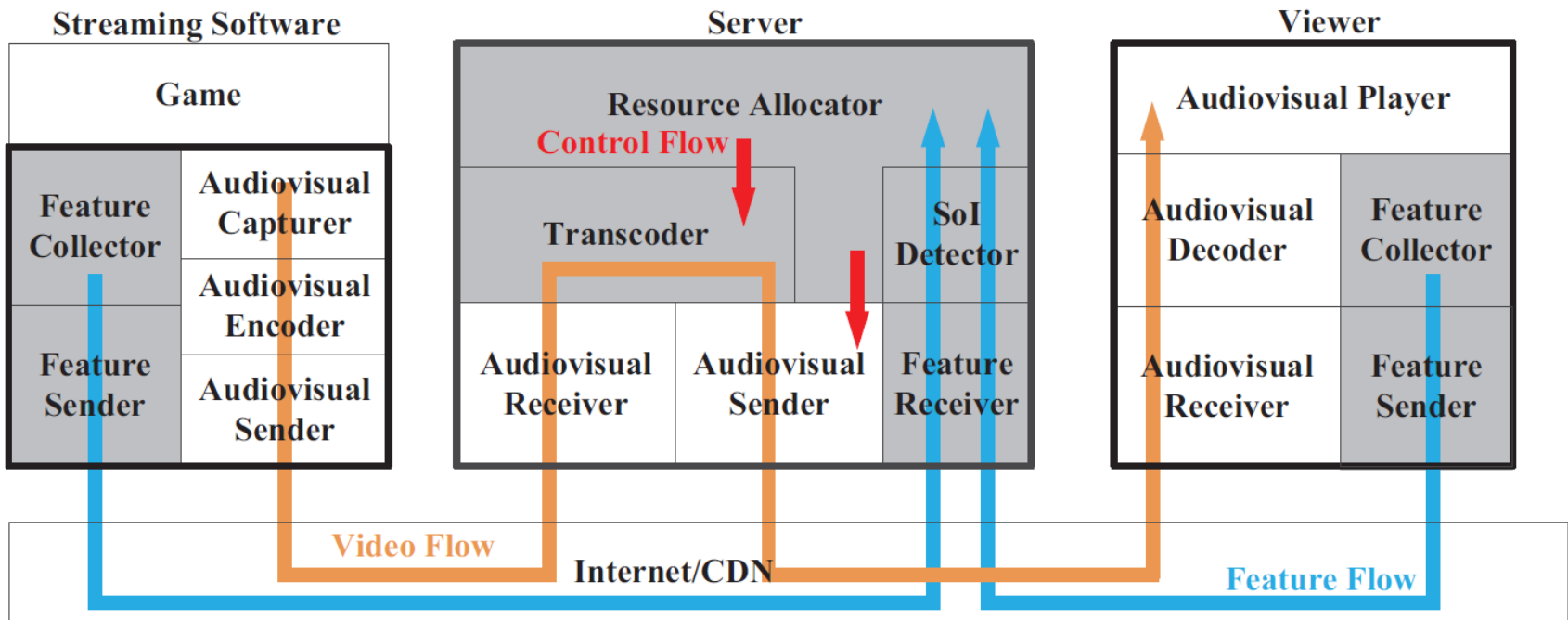
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System Architecture



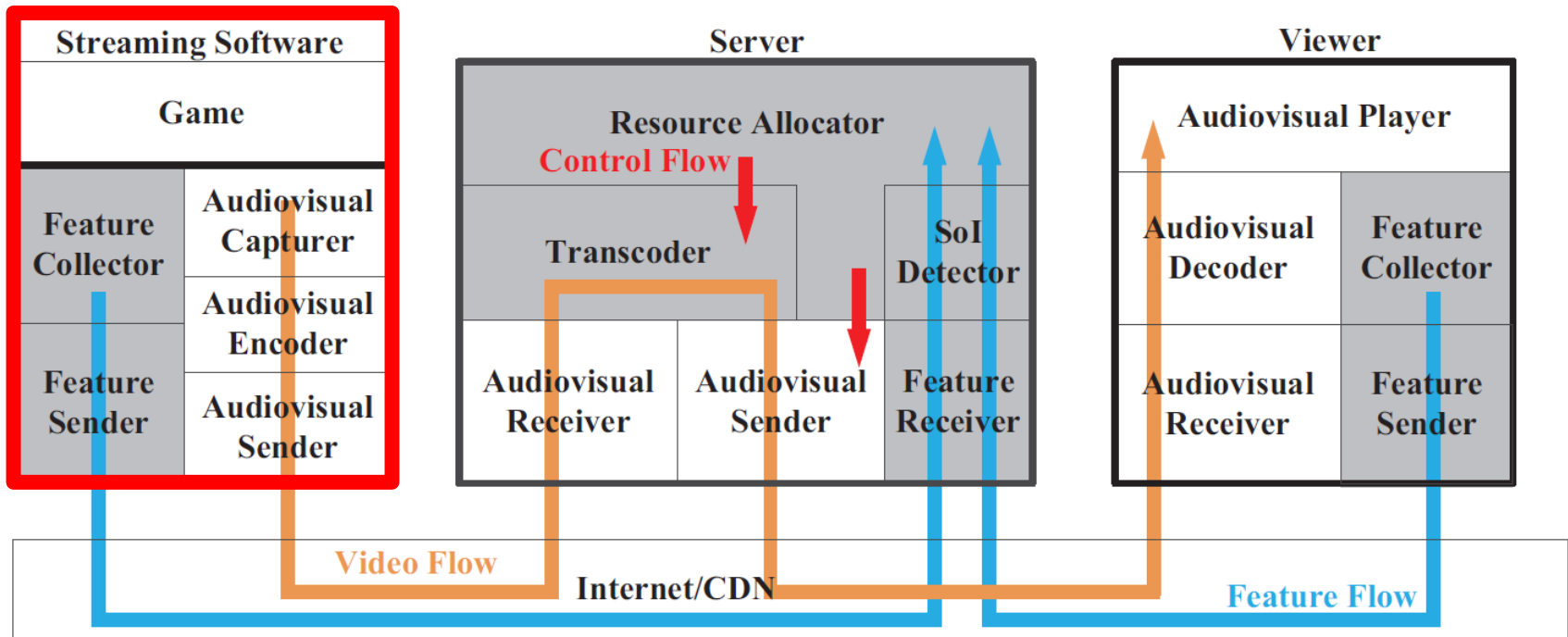
Streaming software

- Stream the gameplay to server



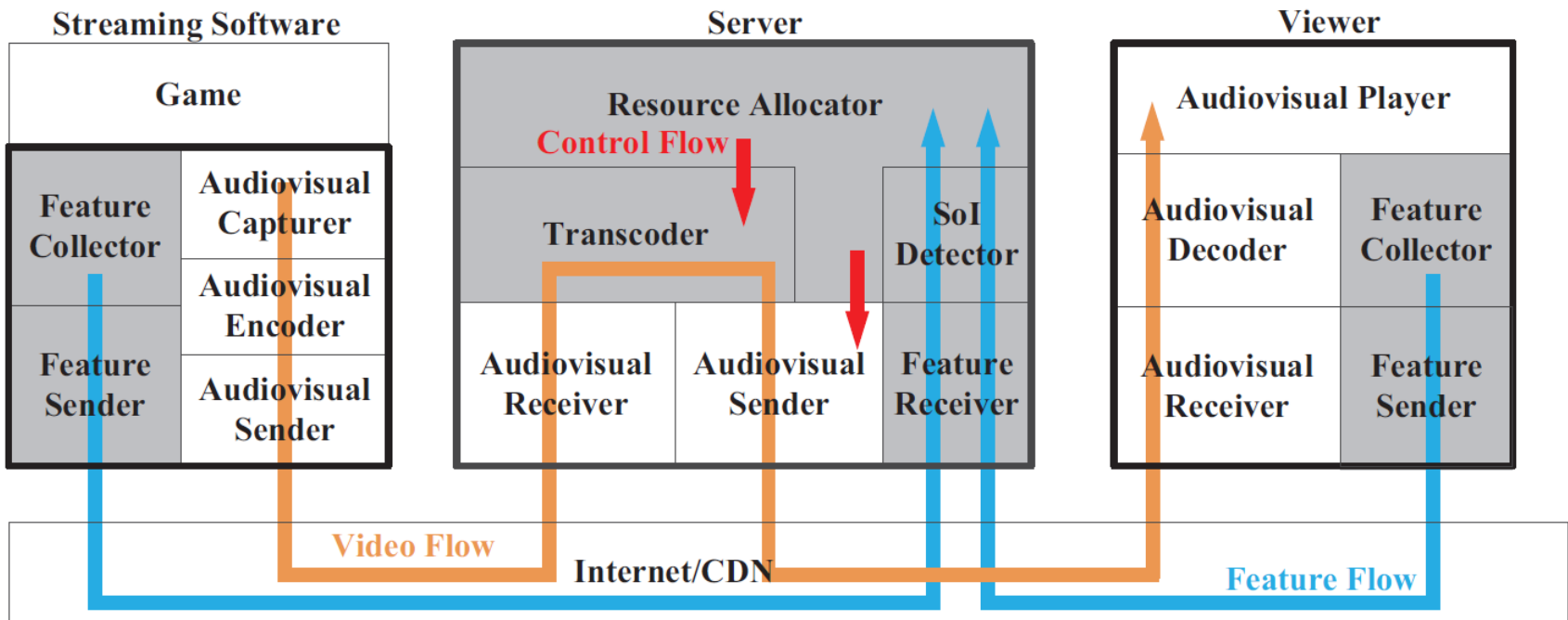
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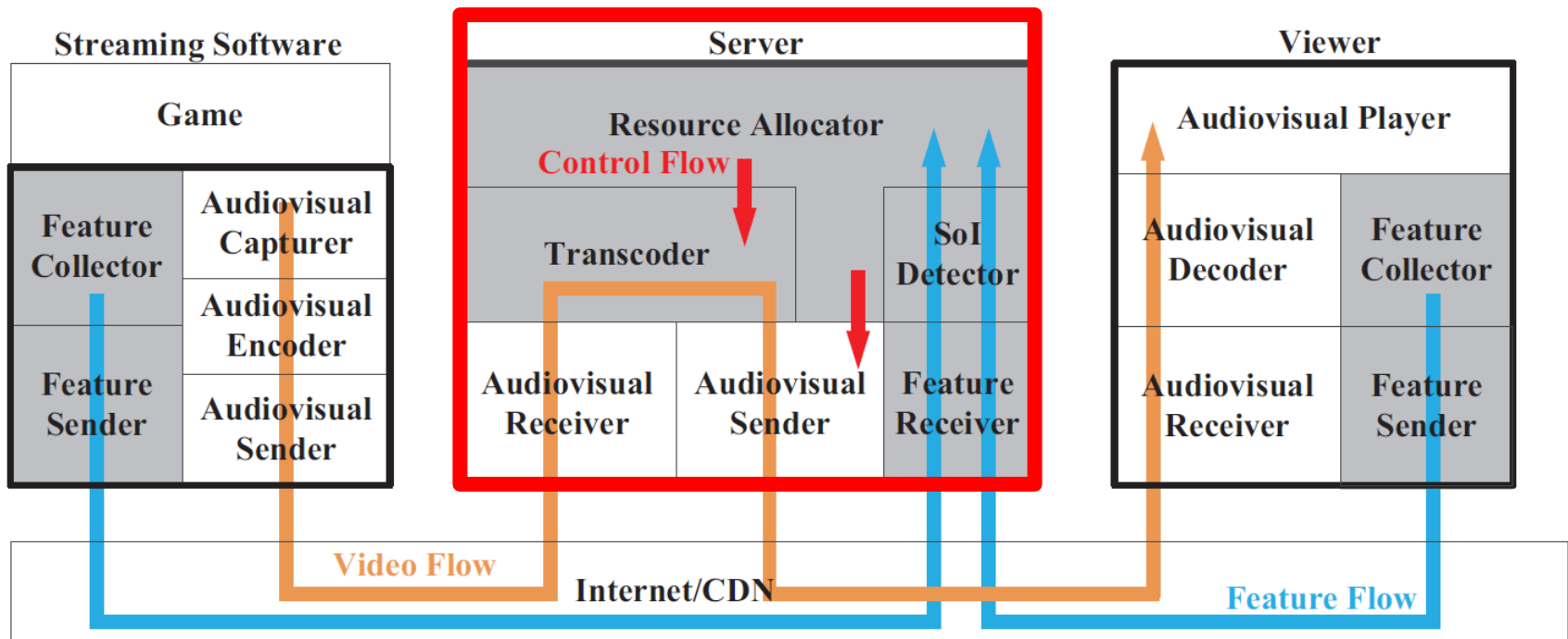
Streaming Server

- Relay the stream to viewers
- Transcode if necessary



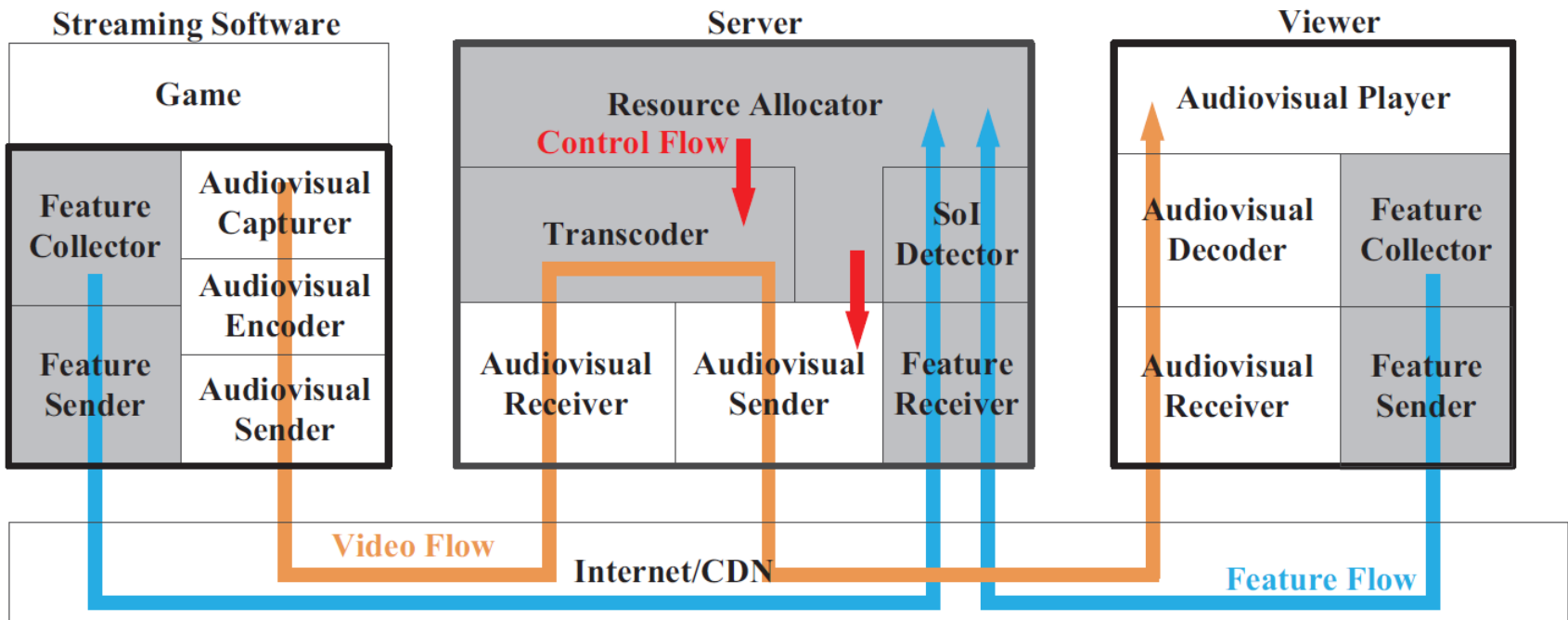
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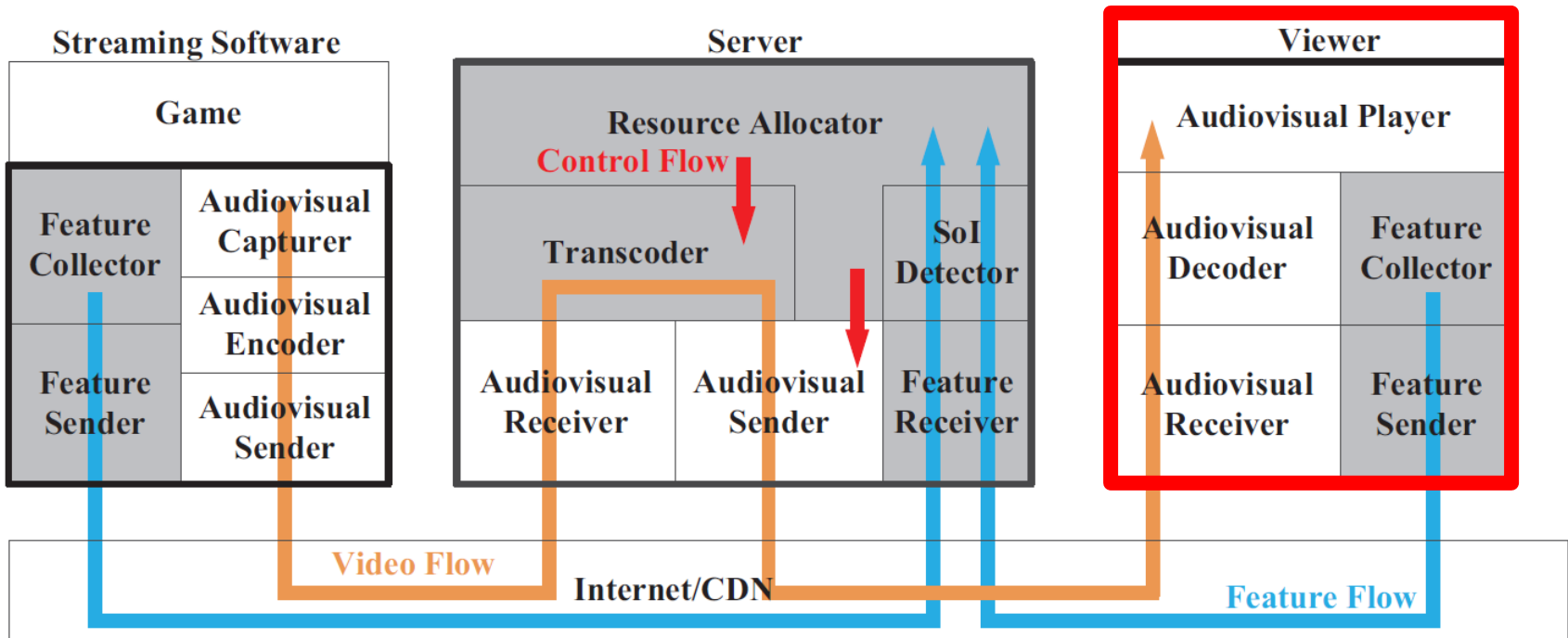
Viewer

- Replay the gameplay stream



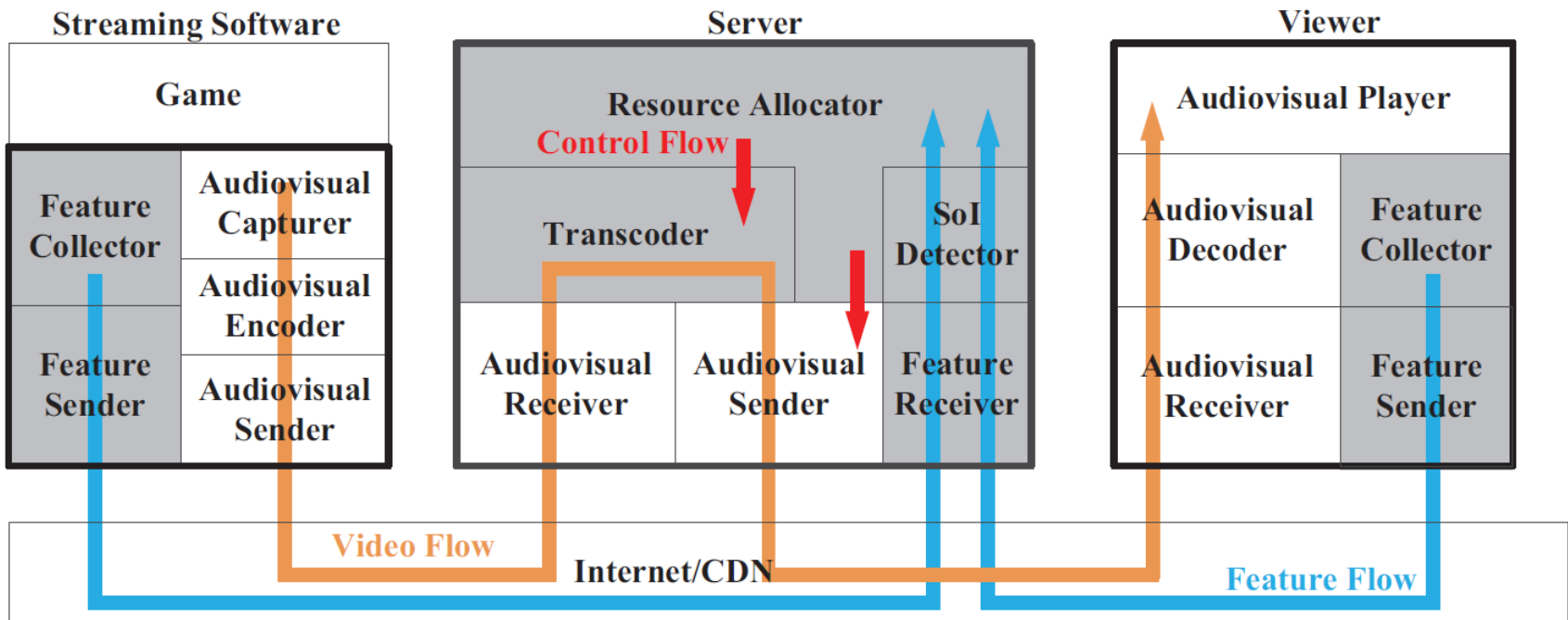
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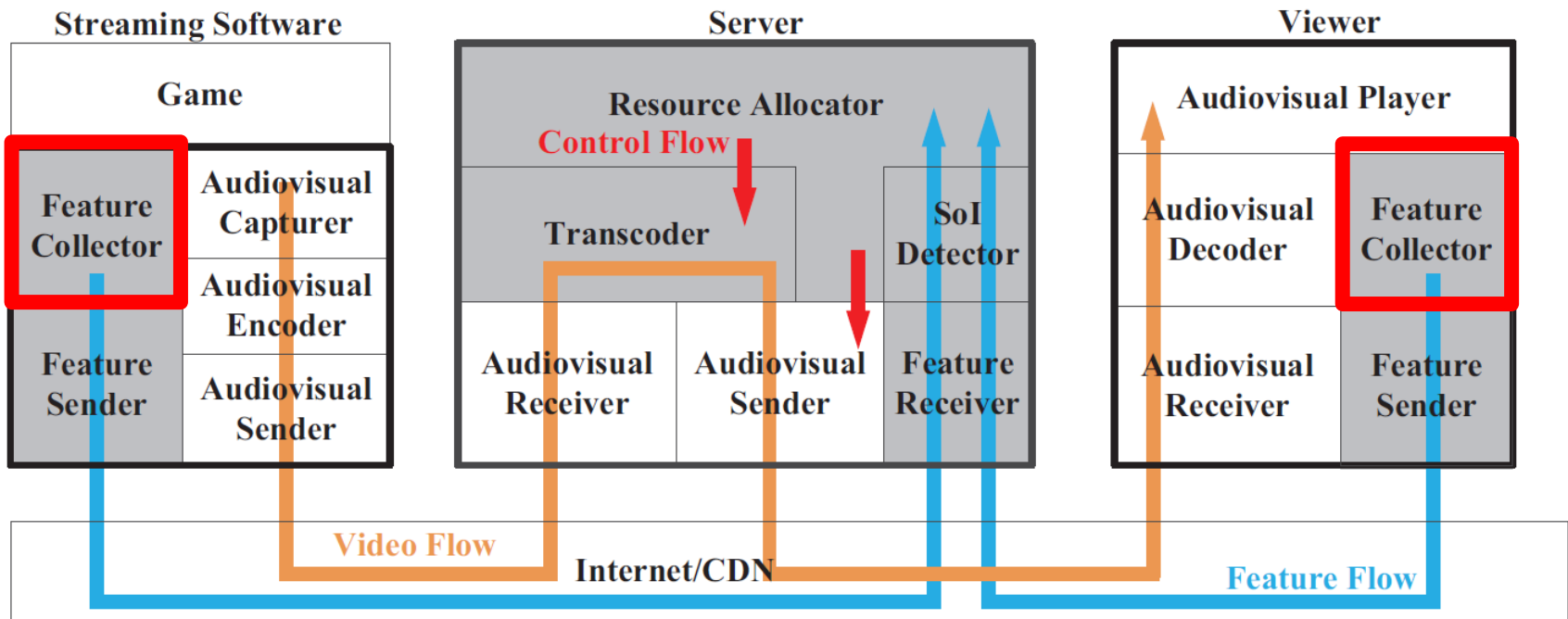
Feature Collector

- Collect the features from streamers
 - Such as CPU usage, foreground window name...
- Also from viewers



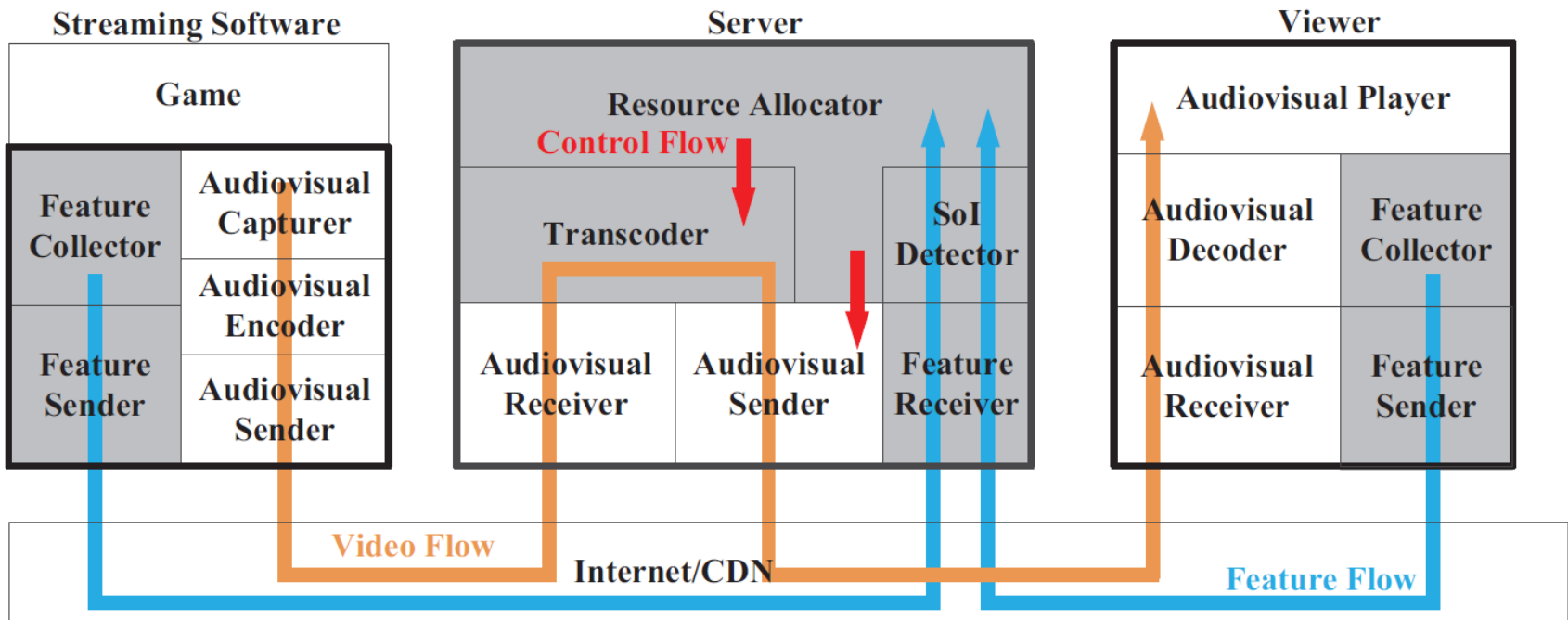
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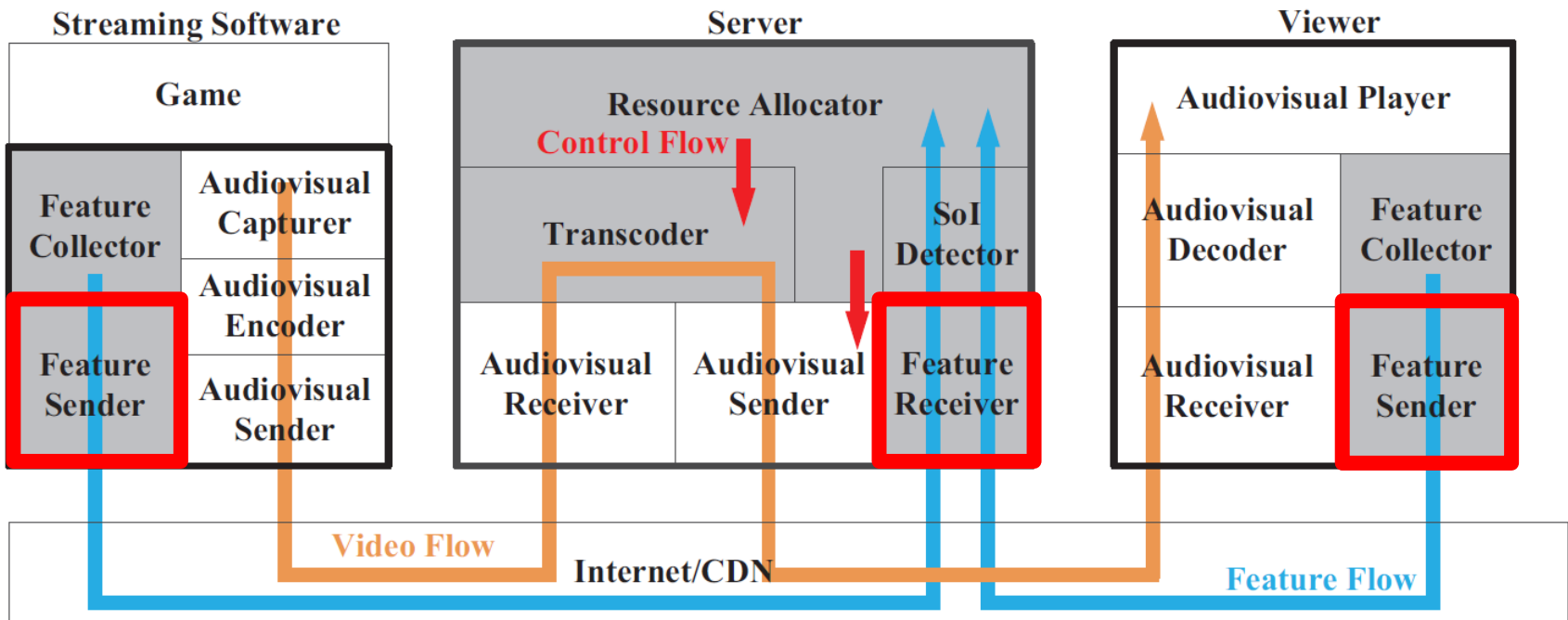
Feature Sender/Receiver

- Send/receive the collected features



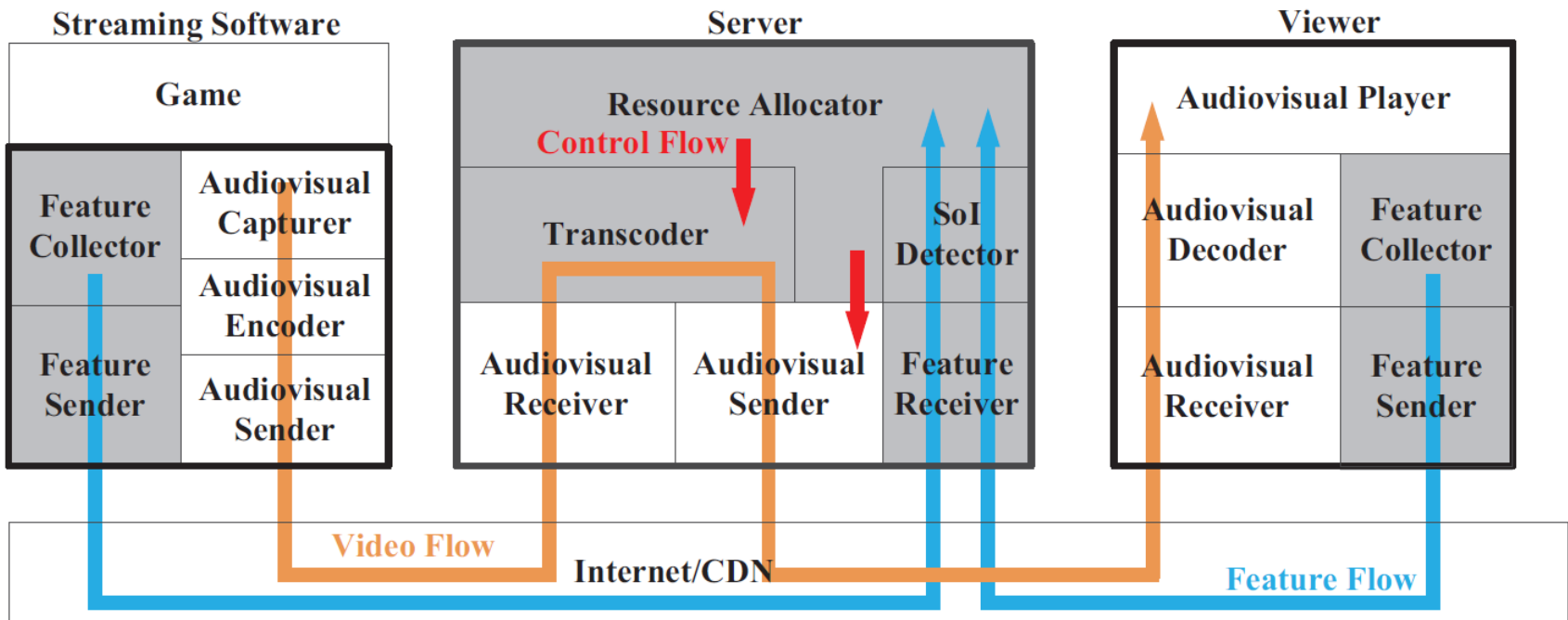
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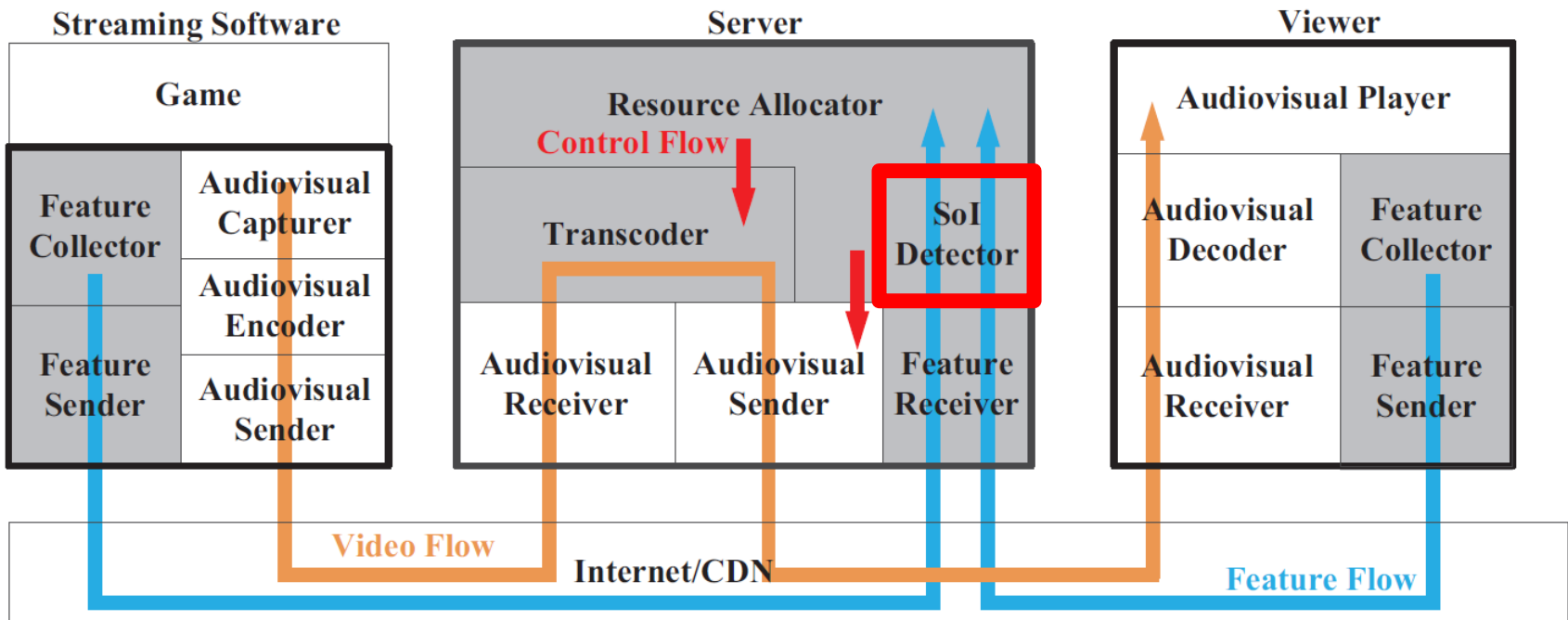
Sol Detector

- Detect Sol from the collected features



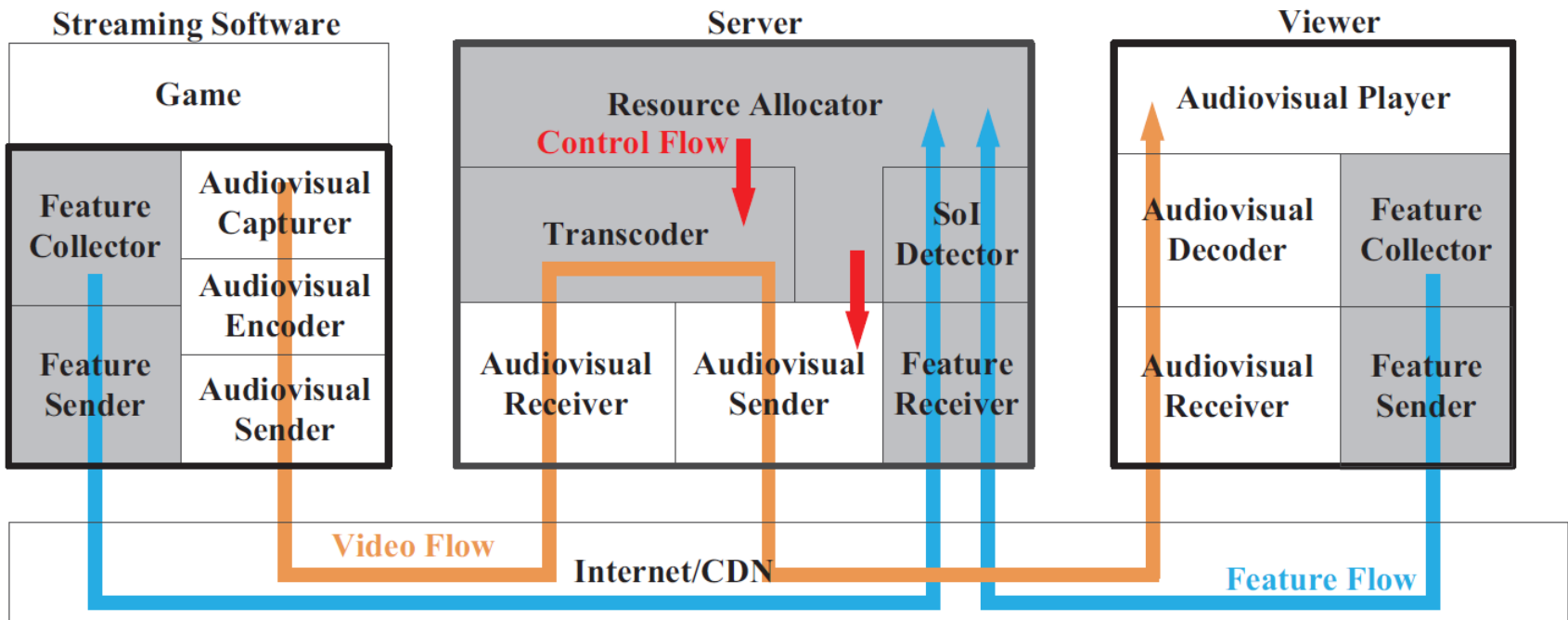
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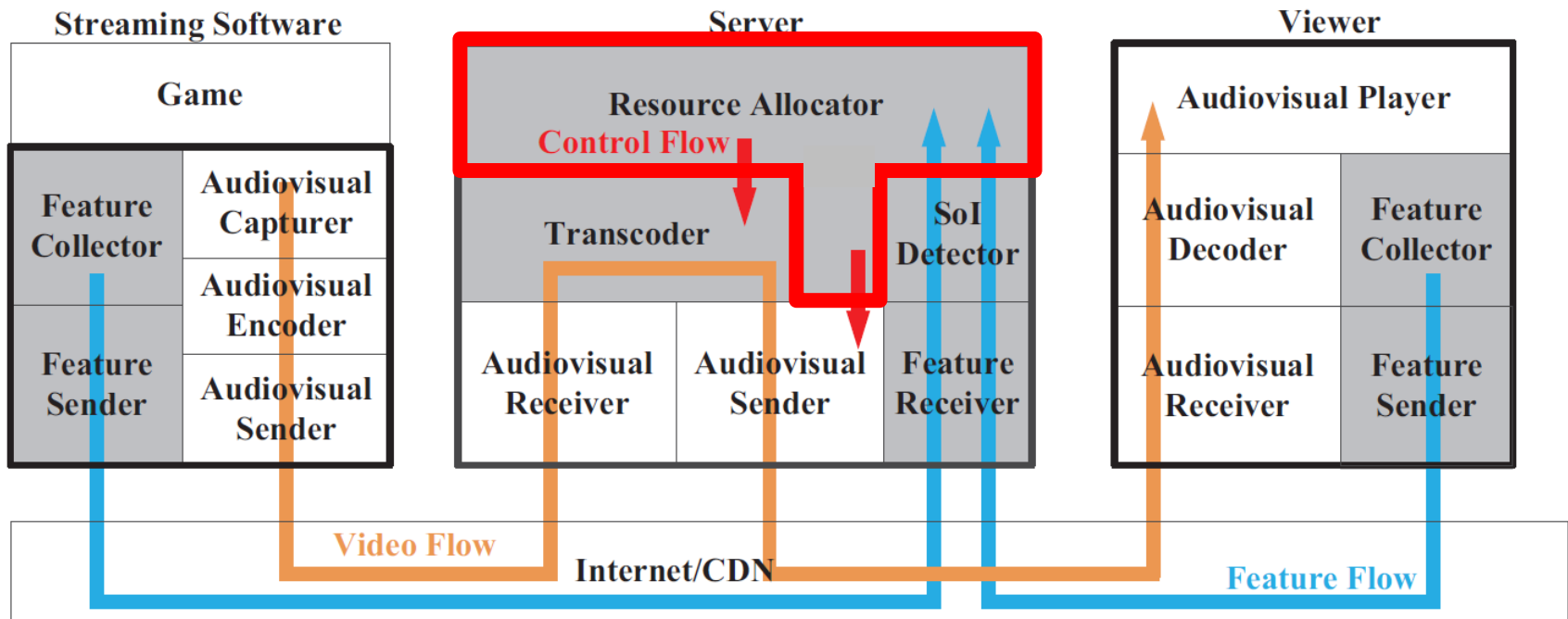
Resource Allocator

- Allocate resources between different streams in the system



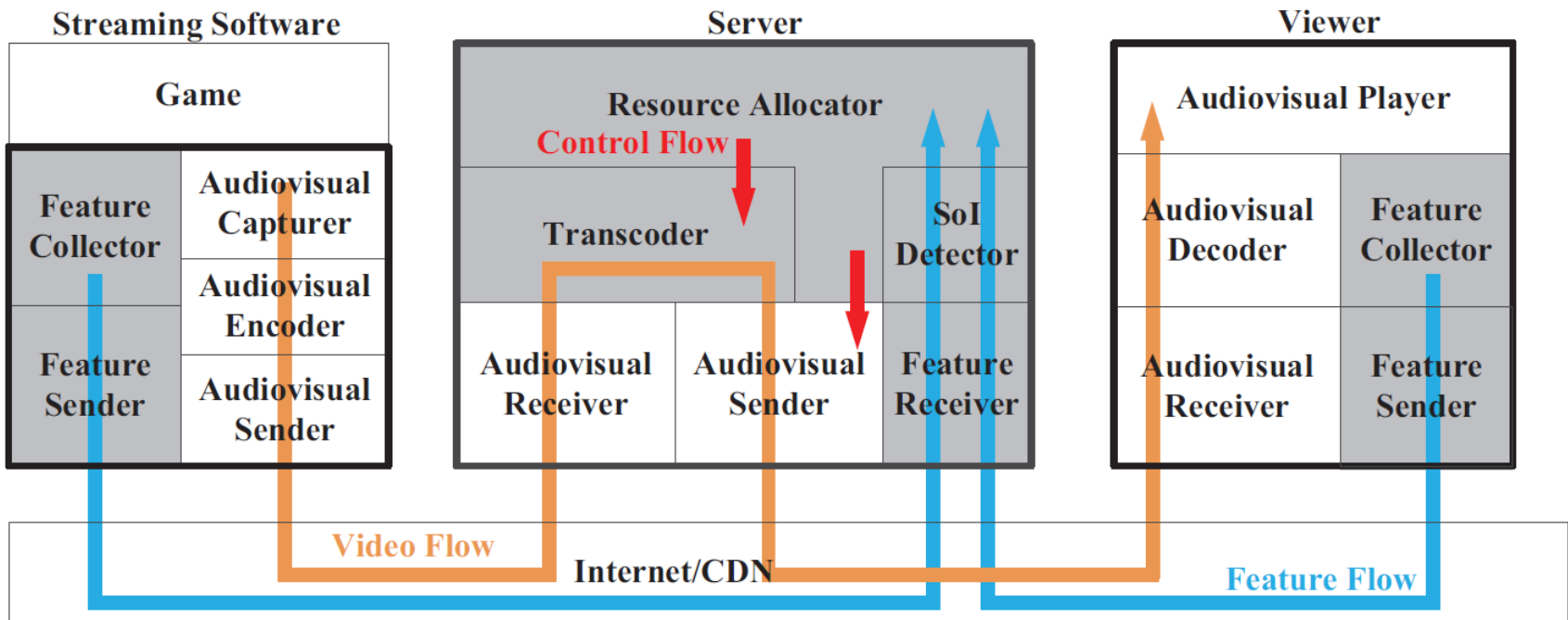
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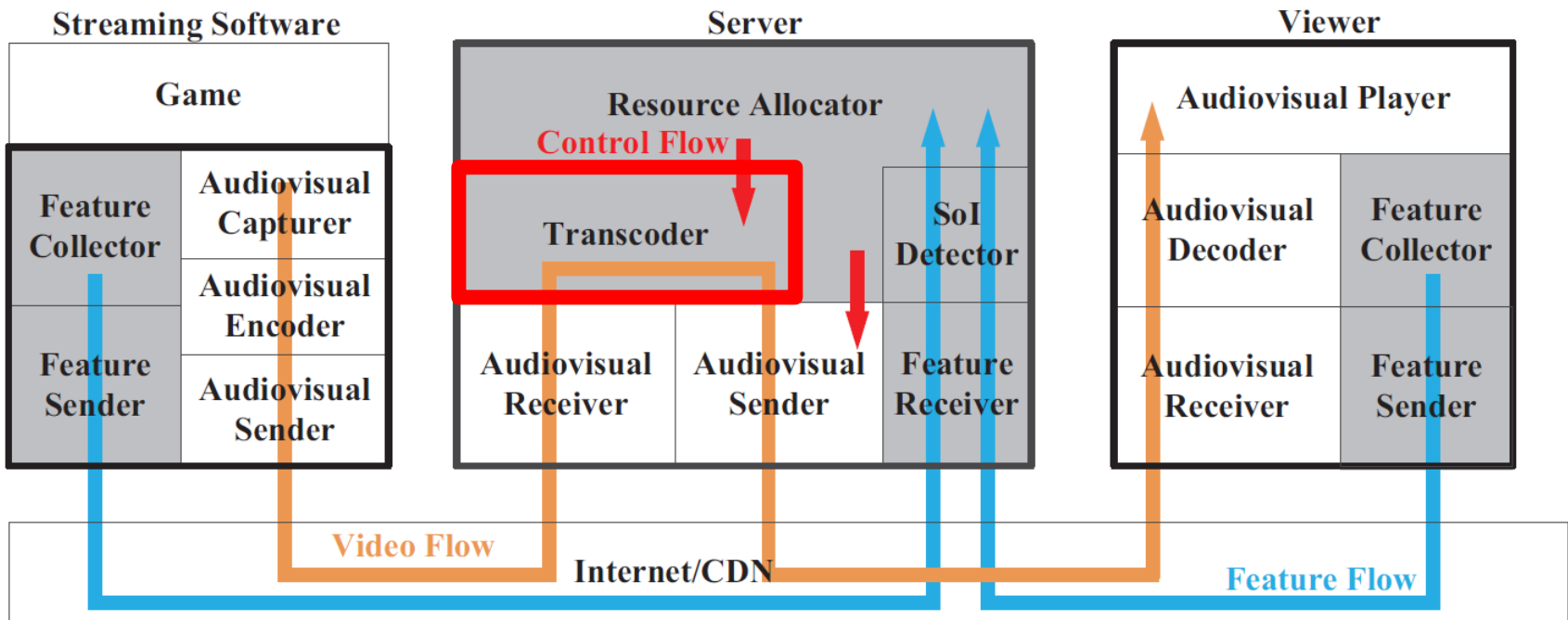
Transcoder

- Transcode the stream



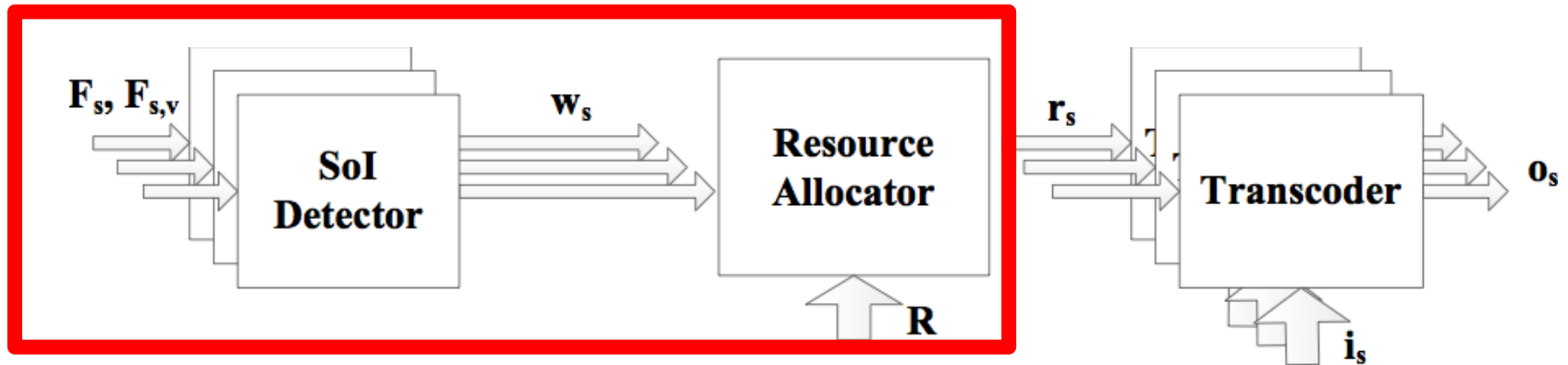
Transcoder

- Transcode the stream



Core Components

- Sol Detector
- Resource Allocator



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Sol Detection

- How to detect Sol?
 - How to detect if viewers are interested?
- From viewers?
 - Ask viewers to manually tell us? Too annoying
 - Automatically? Require mass deploy, privacy issue



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Detect from streamers



How To Detect From Streamer?

- Detect from content (video) [1]?
 - Computationally intensive, not real-time friendly
 - “The segment at 3 hours ago is valuable, to the viewers who’ve watched it 3 hours ago”
- Detect from other features **Our work**
 - Features that can be obtained in real-time

Sol Detecting Problem

- How to determine Sol weight using features
 - From streamers
 - From viewers (if presented)
- Model the problem as
 - Classification, estimate 0/1
 - Regression, estimate $[0, 1]$



Features Collected

- CPU usage
- GPU usage
- Context switch
- Streaming bitrate
- Keyboard/mouse input event
- Number of face in webcam
- System sound magnitude
- Microphone sound magnitude
- Name of foreground window

All can be obtained in real-time

But identifying the relations is hard...

Machine Learning

- Use machine learning algorithms
 - Identify the relations between features and Sol
- Which machine learning model to use?[1]
 - Deep Learning?
 - Support Vector Machine?
 - Random Forest?
 - Gradient Boosting Tree?

Selected Model

- We choose
 - Random Forest
 - Gradient Boosting Tree
- Both are ensemble models of decision trees
- Efficient in training
 - Unlike SVM and Deep Learning
- Popular
 - Applied in multiple fields

Hyperparameter

- RF-based
 - N , Number of trees
 - X , Maximum number of feature in splitting
 - H , Maximum Depth
- GBT-based
 - N , Number of trees
 - E , Shrinkage (learning rate)
 - H , Maximum Depth
 - M , Subsample

Solution Approach

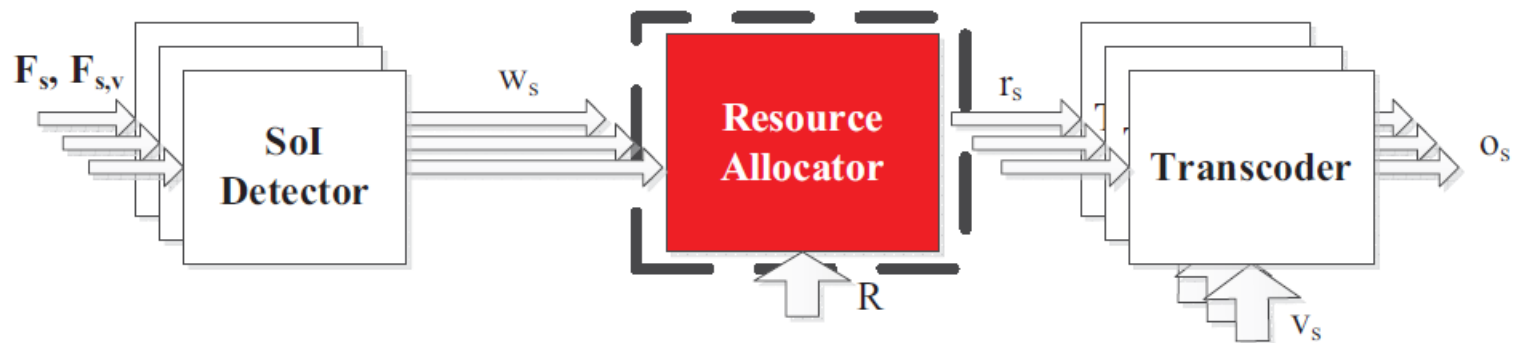
- Implemented in Python
 - Scikit-learn for Random Forest
 - XGBoost for Gradient Boosting Tree
- Using both regressor and classifier variant of them
 - RFR-based & GBTR-based algorithm
 - RFC-based & GBTC-based algorithm
- We conduct supervised learning
 - Details are given in evaluation

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Resource Allocation Problem

- How to allocate resources among streams to maximize viewing quality q_s ?
 - Take r_s as the decision variable
- Leverage Sol weight w_s



Connect r_s With q_s

- We adopt a rate-distortion model [1]

$$d_s = D_{0,s} + \frac{\theta_s}{r_s - R_{0,s}}$$

- d_s : distortion in Mean Square Error (MSE)
- $D_{0,s}, \theta_s, R_{0,s}$: model parameters obtained by non-linear regression

Quantifying Viewing Quality

- We use Peak-Signal-to-Noise Ratio (PSNR)

$$-q_s = 10 \log_{10} \frac{255^2}{d_s}$$

- The proposed algorithm can utilize other quality metrics with monotonically increasing property

Problem Formulation

$$\text{maximize } \sum_{s=1}^S q_s w_s = \sum_{s=1}^S 10 \log_{10} \frac{255^2}{D_{0,s} + \frac{\theta_s}{r_s - R_{0,s}}} w_s$$

$$\text{s.t. } \sum_{s=1}^S r_s w_s \leq R;$$

$$r_s \in R^+, \forall s = 1, 2, \dots, S.$$

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Objective: Maximize the viewing quality of all the viewers in Sol

Problem Formulation

$$\text{maximize } \sum_{s=1}^S q_s w_s = \sum_{s=1}^S 10 \log_{10} \frac{255^2}{D_{0,s} + \frac{\theta_s}{r_s - R_{0,s}}} w_s$$

$$\text{s.t. } \sum_{s=1}^S r_s w_s \leq R;$$

Under the constraint that the total bandwidth usage is under R

$$r_s \in R^+, \forall s = 1, 2, \dots, S.$$

Derive Closed Form Formula

- Leverage Lagrangian Multiplier method
 - Added λ
- We get Lagrangian function L

$$\sum_{s=1}^S \left(10 \log_{10} \frac{255^2}{D_{0,s} + \frac{\theta_s}{r_s - R_{0,s}}} w_s \right) + \lambda \left(\left(\sum_{s=1}^S r_s w_s \right) - R \right)$$

- Take partial derivative of L

$$\frac{\partial \mathcal{L}}{\partial \lambda} = \left(\sum_{s=1}^S r_s w_s \right) - R;$$

In order to get the extreme value of L

$$\frac{\partial \mathcal{L}}{\partial r_s} = \lambda + \frac{10\theta_s}{\log 10 (r_s - R_{0,s})^2 \left(D_{0,s} + \frac{\theta_s}{r_s - R_{0,s}} \right)}$$

S of them

Solving the Formula

- Let $\frac{\partial \mathcal{L}}{\partial r_s} = 0$

- After derivation

$$r_s = R_{0,s} + \frac{-(\lambda \log 10\theta_s) - \sqrt{(\lambda \log 10\theta_s)^2 - 40\lambda \log 10\theta_s^2}}{2(\lambda \log 10D_{0,s})}$$

- Combine the results above with

$$\frac{\partial \mathcal{L}}{\partial \lambda} = \left(\sum_{s=1}^S r_s w_s \right) - R = 0$$

The result of the other partial derivative

- We can obtain the optimal λ

- Then derive the optimal bitrates for streaming to each viewer \mathbf{v} of streamer s

- **Slow to solve** **There are $S + 1$ formulas**

A Real-Time Approximation Algorithm

- Solve the problem efficiently
 - With a controllable error of ϵ
- Called **SoI-based R-D Optimized Algorithm (SoI RDO)**

SoI RDO Algorithm

Algorithm 1 Efficient SoI RDO Algorithm

```
1:  $U \leftarrow 0, L \leftarrow -1$  // Upper/lower bounds
2:  $\epsilon \leftarrow 10^{-7}$  // Default error
3: while  $L + \epsilon < U$  do
4:    $\lambda = (U + L)/2$  Binary search on proper  $\lambda$  value
5:   for  $\forall s \in S$  do Derive all  $r_s$  &  $q_s$  using  $\lambda$  this round
6:     Derive all the  $r_s$  using  $\lambda$  and Eq. (5)
7:   if  $\exists r_s$ , such that  $q_s \in \mathbb{C}$  or  $r_s < R_L$  or  $r_s > R_U$  then
8:     Adjust  $U$  or  $L$  accordingly Check if all  $r_s$  and  $q_s$  are valid
9:   else if  $\sum_{s \in S} r_s w_s > R$  then Check if use too much bandwidth
10:     $U = \lambda$ 
11:   else
12:     $obj = \sum_{s \in S} q_s w_s$  Calculate objective function
13:     $L = \lambda$ 
14: if  $obj$  is undefined, return no answer, o.w. return  $\lambda$  and  $r_s$ .
```

Lemma 1

Lemma 1 (Approximation Gap). *Sol RDO algorithm always finds λ within a gap of ε to the optimal λ^* under the given bandwidth constraint R .*

- Reason why the quality metrics need to have monotonically increasing property.

Lemma 2

Lemma 2 (Complexity). *Algorithm 1 runs in $O(S \log \epsilon^{-1})$*

- $\log \epsilon^{-1}$ round of search in worst case
- Each round need to do
 - Derive r_s, q_s for every streamer s $O(S)$
 - Check if r_s is valid for every streamer s $O(S)$
 - Calculate the total consumed bandwidth $O(S)$
 - Calculate the objective value $O(S)$
- $O(\log \epsilon^{-1}) \times (O(S)) = O(S \log \epsilon^{-1})$

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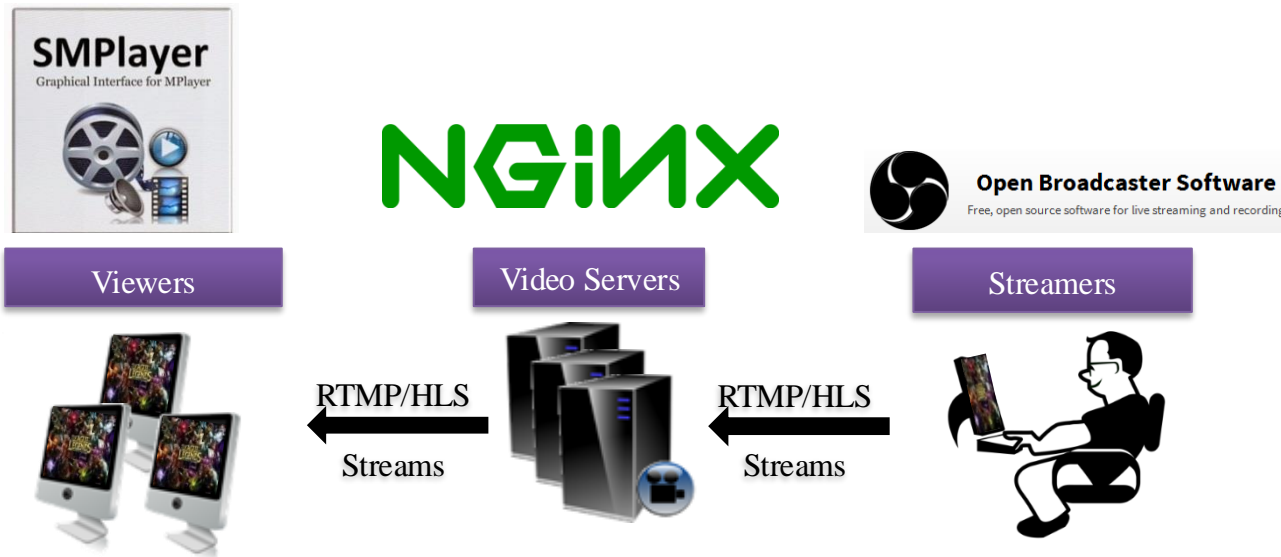
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Open Source Testbed

- Streaming software: OBS [1]
- Streaming server: NGINX [2] with RTMP plug-in
- Viewer: SMPlayer [3]



[1] <https://obsproject.com/>

[2] <http://nginx.org>

[3] <http://smplayer.sourceforge.net/>

SOI DETECTOR EXPERIMENT SETUP

Collecting the Dataset

- Collected at a League of Legend tournament
- 50 participants, 10 matches
 - 100 video & trace
 - 207004 seconds in total
- 81 valid video & trace
 - 162842 samples



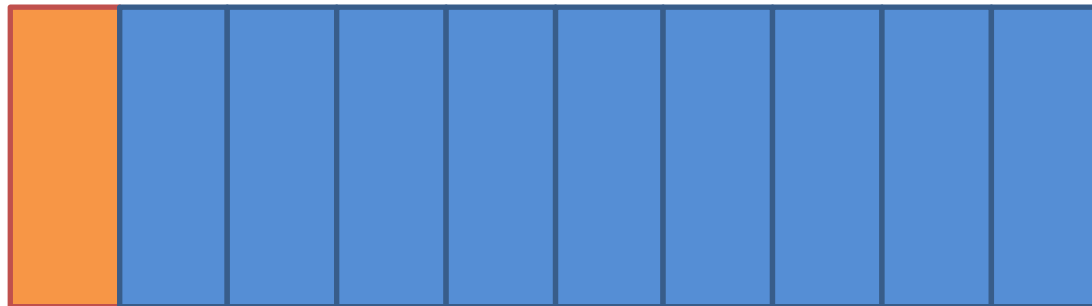
Marking Sol Ground Truth

- We recruited 17 viewers to mark Sol manually
 - 10 videos
 - With a modified video player
- Collected 74 logs
 - Cover 27010 samples



Partition the Dataset

- 90% as training dataset
 - Used in hyperparameter tuning
 - 18677 samples
- The rest 10% as evaluation dataset
 - 2076 samples



Performance Metrics

- Classification

- F-measure = $2 \cdot \frac{\textit{precision} \cdot \textit{recall}}{\textit{precision} + \textit{recall}} \in [0, 1]$

- Training Time

- Regression

- R-squared score $\in [1, -\infty]$

- Training Time

Hyperparameter Tuning

- 10-fold cross validation
- Grid search
- RF-based
 - ***N***: [30 ,60, 120, 240, 480]
 - ***X***: [5, 10, 20, 40]
 - ***H***: [10, 20, 40, 80, 160]
- GBT-based
 - ***N***: [5, 10, 20, 40, 80]
 - ***E***: [0.01, 0.05, 0.1, 0.2, 0.4]
 - ***H***: [5, 10, 20, 40, 80]
 - ***M***: [0.5, 0.6, 0.7, 0.8, 0.9]

The Optimal Hyperparameter

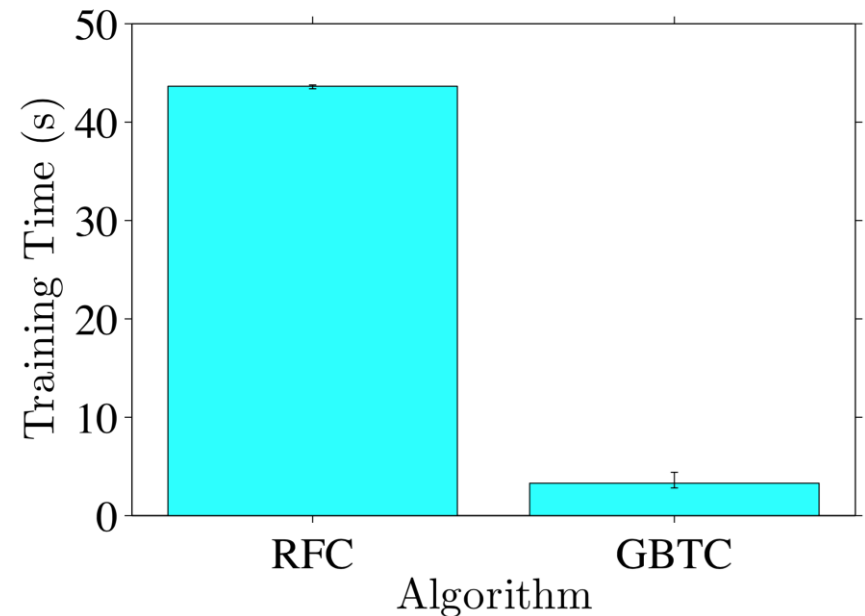
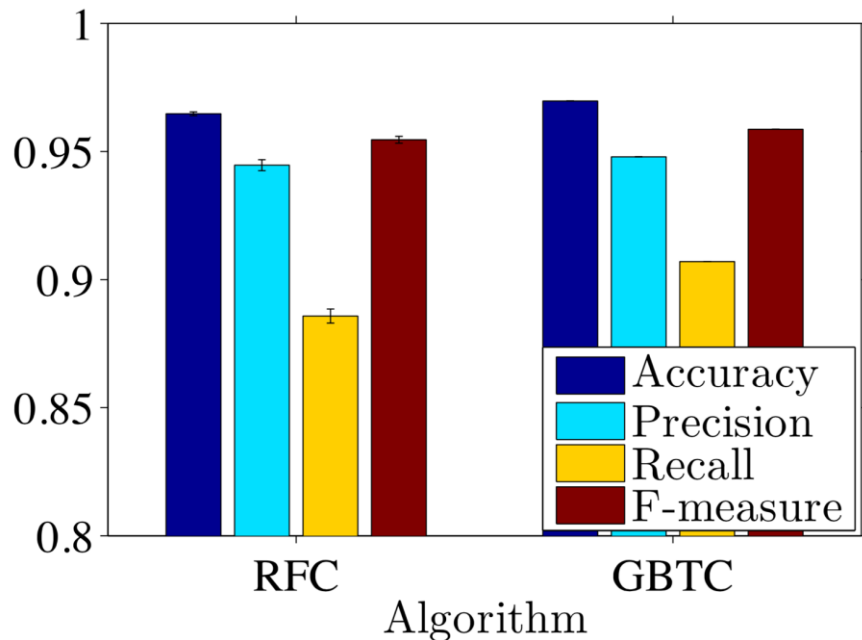
- RFR-based algorithm
 - $N: 240, X: 5, H: 40$
- GBTR-based algorithm
 - $N: 80, E: 0.2, H: 10, M: 0.9$
- RFC-based algorithm
 - $N: 480, X: 5, H: 80$
- GBTC-based algorithm
 - $N: 80, E: 0.2, H: 80, M: 0.7$

Use them in evaluation

SOI DETECTOR EXPERIMENT RESULT

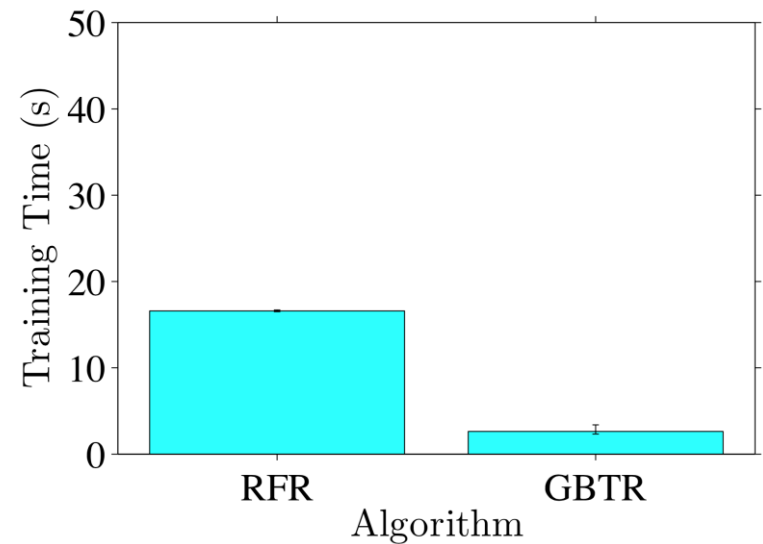
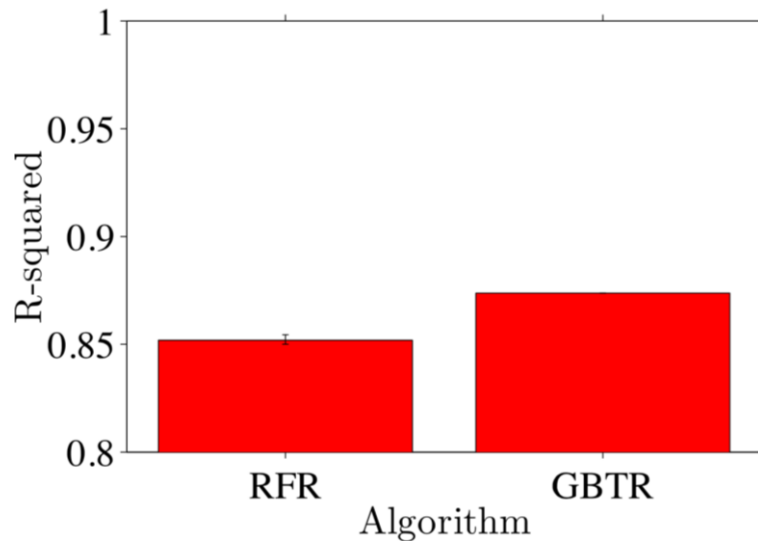
RFC-based & GBTC-based Algorithm

- Over 0.95 F-measure score
- Both terminated under 42 seconds
 - Less than 5 seconds for GBTC



RFR-based & GBTR-based algorithm

- Over 0.85 in R-squared
- Both algorithm terminated under 20 seconds



Observation

- All algorithms provide good result
 - GBT-based algorithm generally gives better result
- GBT-based algorithm have shorter training time
 - Not what we expect
 - XGBoost leverage up to 62 cores in our experiment
 - Scikit-learn's RF implementation only use single core

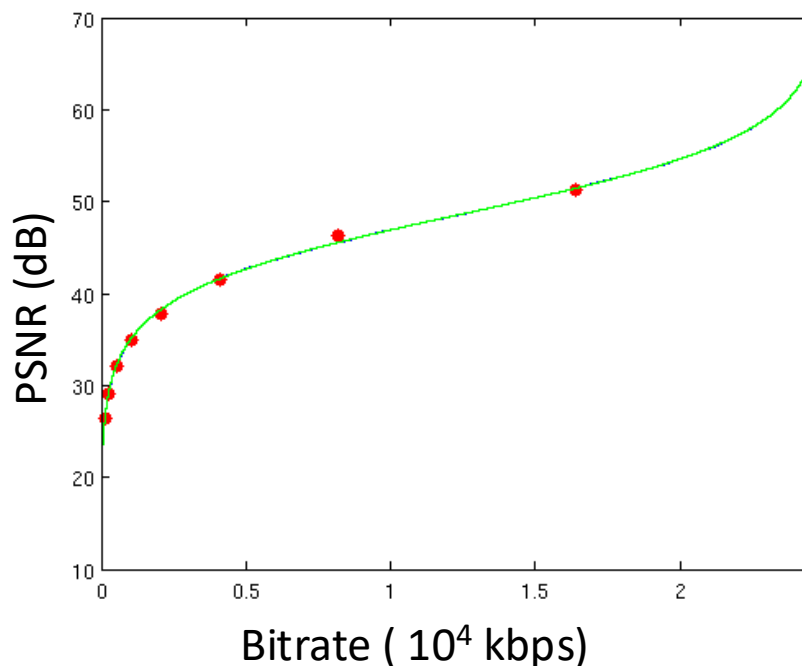
RESOURCE ALLOCATOR EXPERIMENT SETUP

Allocator Simulation Setup

- We captured 8 game streamed on our testbed
 - Age of Empires II
 - Spellweaver
 - Hearthstone: Heros of Warcraft
 - Minecraft
 - Starcraft II
- Resolution between 720p and 1080p

Allocator Simulation Setup (cont.)

- Transcode them into different bitrates
 - Obtain the PSNR value
 - Perform non-linear regression to get the R-D model parameters



Allocator Simulation Setup (cont.)

- We recruited several viewers to watch the recorded live stream session
 - Mark Sol
 - 14 logs in total



Simulator & Baseline

- Simulator is implemented in Python
- Sol RDO algorithm is implemented in C
- Two baseline solution
 - **Equal Share (ES)**: equally divides bandwidth between all viewers **State-Of-The Art live game streaming solution**
 - **R-D Optimized (RDO)**: optimized in R-D sense, but does not take Sol into consideration
Quantify the benefit of Sol driven allocation
- Also an implementation of optimal algorithm
 - In Matlab, as a benchmark of Sol RDO

Performance Metrics

- Expected quality
 - The objective value reported by algorithm
- Actual quality
 - The objective value reported by simulator
- Running time
 - Runtime of solving the resource allocation problem
- Expected consumed bandwidth
 - Reported by algorithm
- Actual consumed bandwidth
 - Reported by simulator
- Network overhead
 - Network traffic generated by feature senders

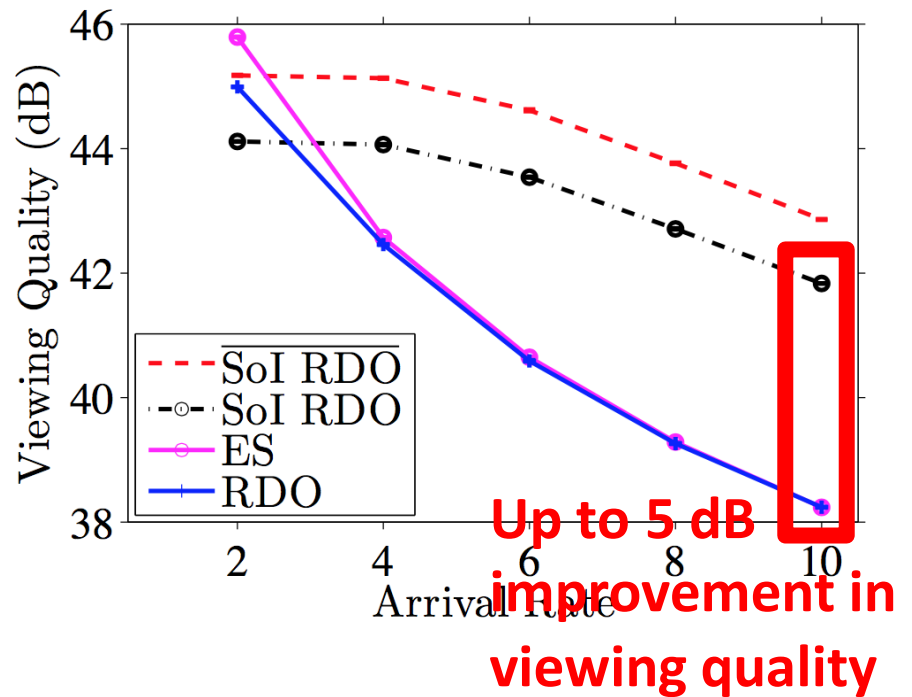
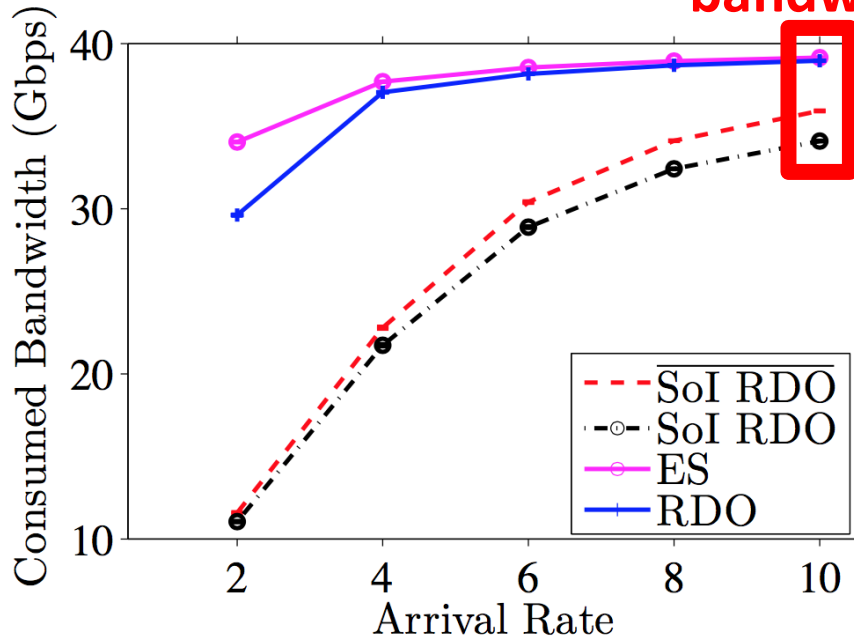
Experiment Parameters

- 6 hour long simulations
- Poisson arrival rate of viewers in each stream
 - {2, 4, 6, 8, 10} per minutes
- Number of stream
 - {16, 32, 64, 128, 256}
- Total outbound bandwidth
 - {10, 20, 40, 80, 160} Gbps
- Interval of invoking resource allocator
 - {1, 2, 5, 10, 60} seconds

RESOURCE ALLOCATOR RESULTS

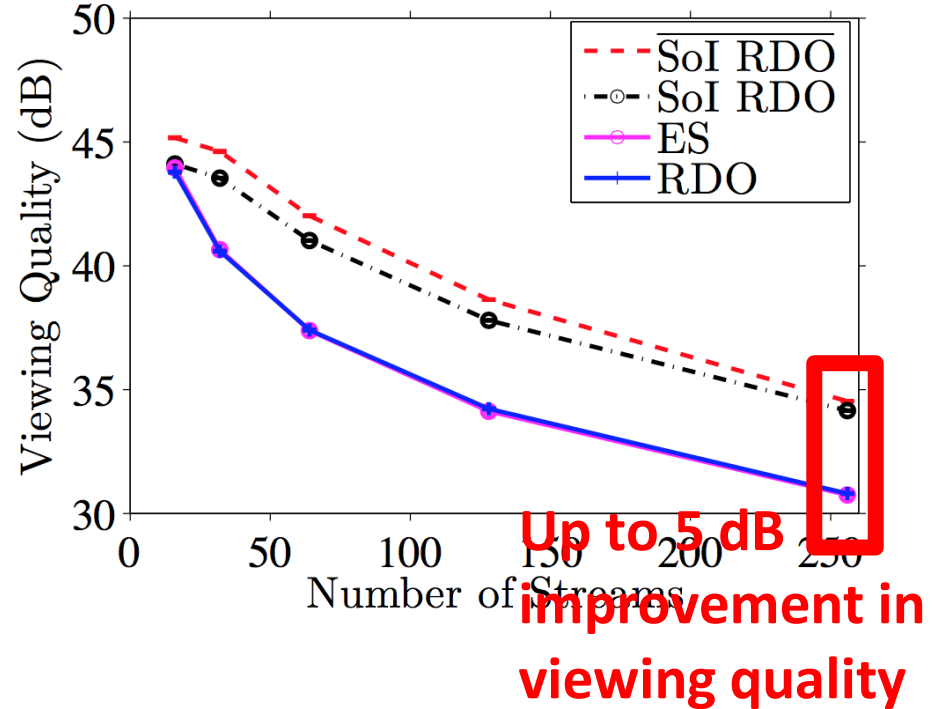
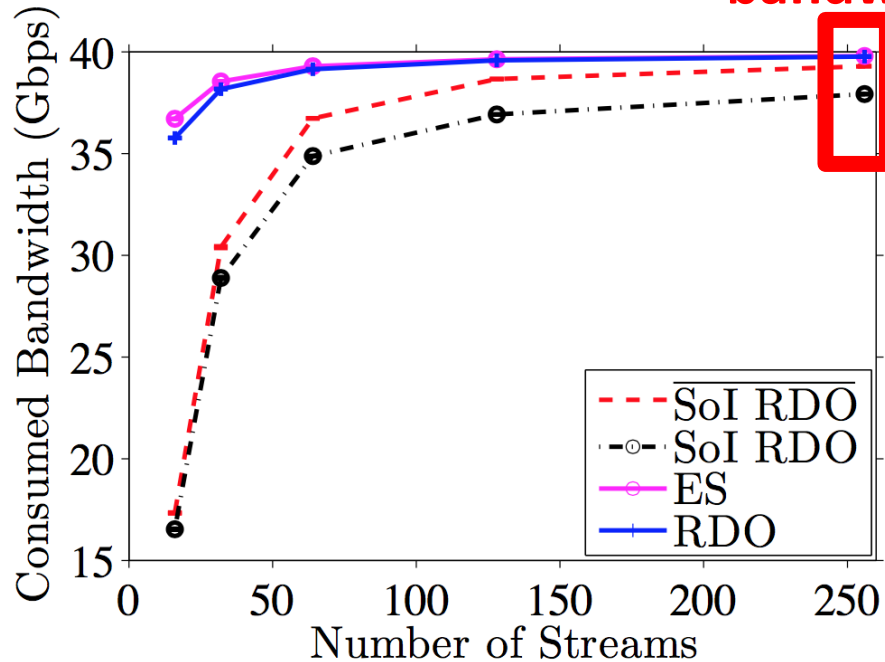
Our Algorithm Outperforms RDO and ES

- Different viewer arrival rate
With less consumed bandwidth



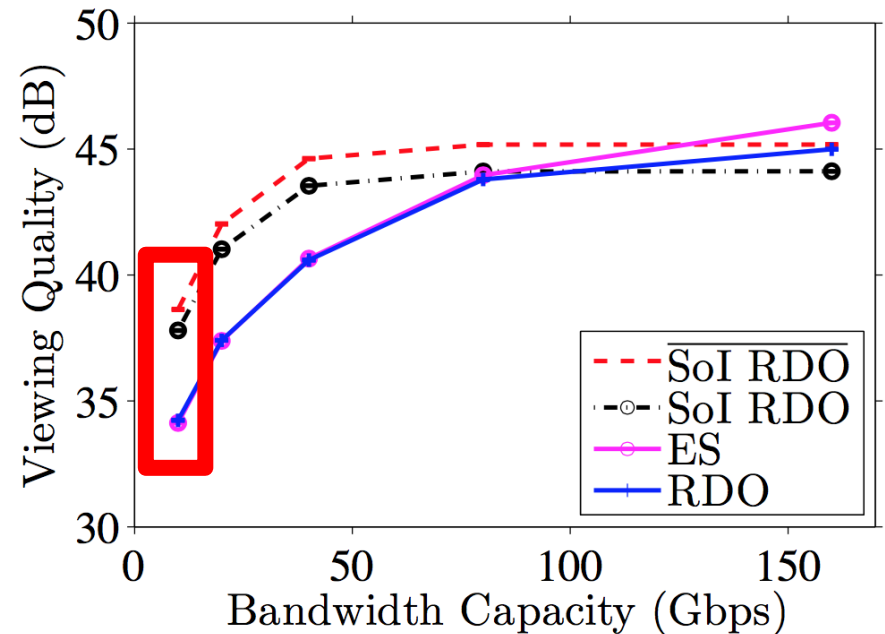
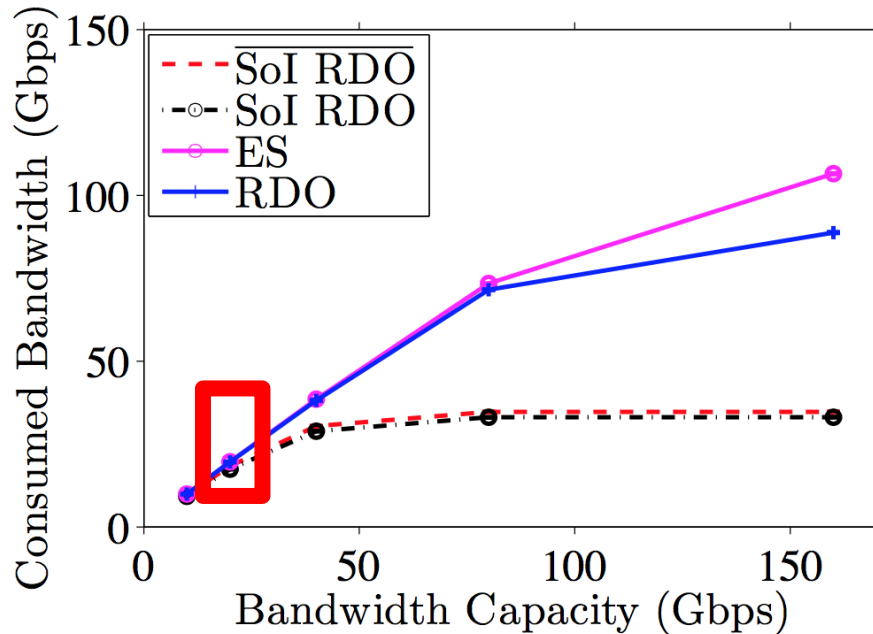
Our Algorithm Outperforms RDO and ES

- Different number of stream
With less consumed bandwidth



Our Algorithm Outperforms RDO and ES

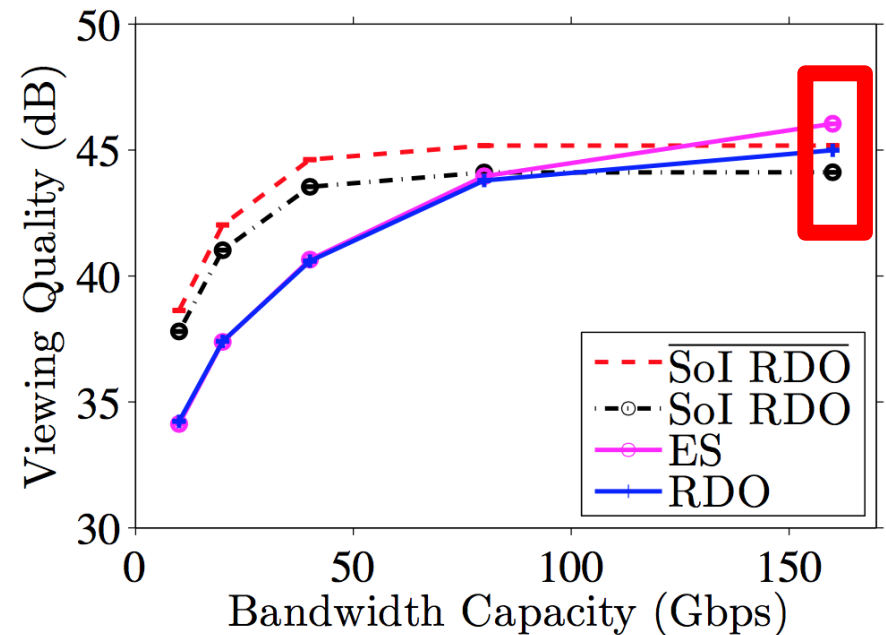
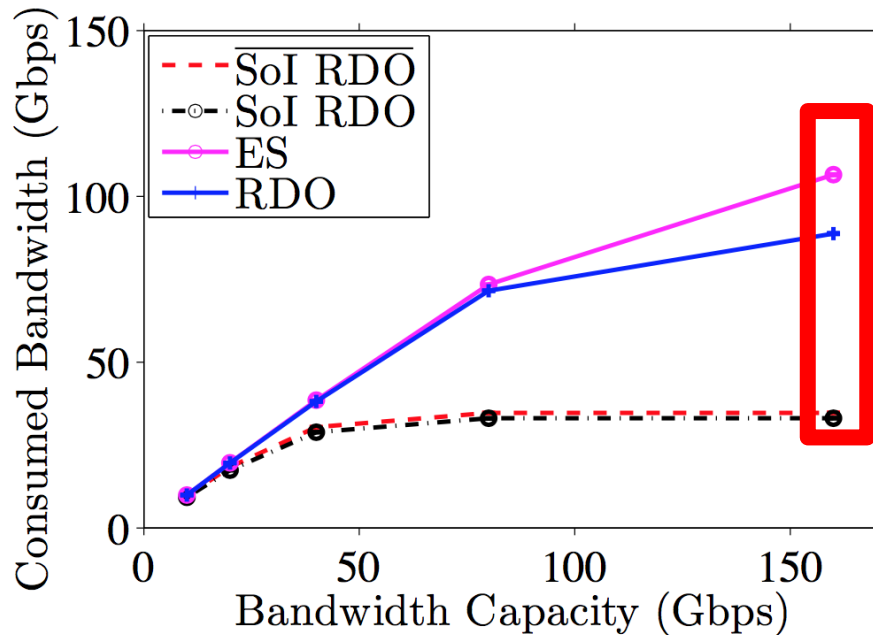
- Different bandwidth capacity



Better quality under the same bandwidth usage

Our Algorithm Outperforms RDO and ES

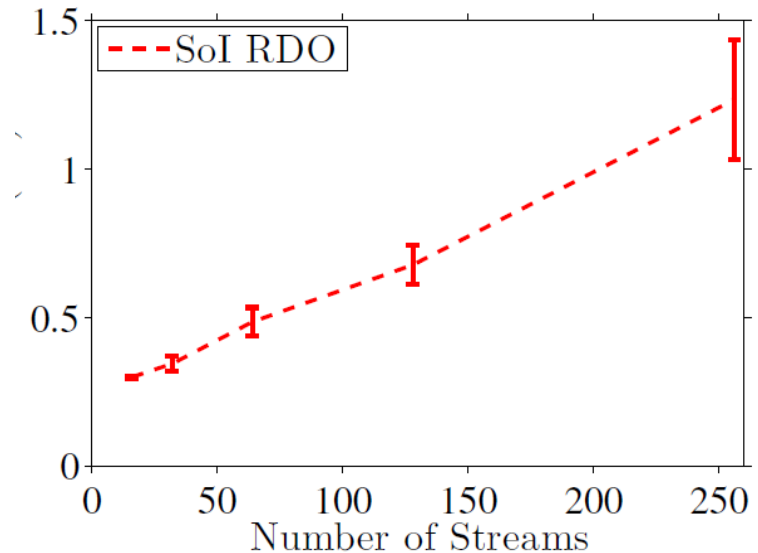
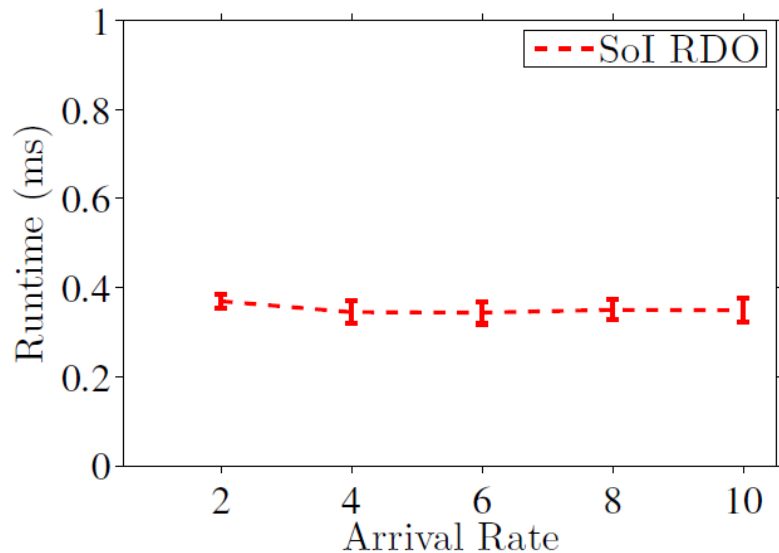
- Different bandwidth capacity



Slightly outperformed by ES and RDO, but that was with significantly larger bandwidth consumption

Our Algorithm Scales Well

- Always below 1.5 ms
 - 100+ thousand viewers
- Matlab solution may take over 6 minutes
 - With 58% chance of feasible solution



Outline

- Introduction & Motivation
- System Overview
- Sol Detector
- Resource Allocator
- **Evaluation**
- Conclusion

Outline

- Introduction & Motivation
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Conclusion

- We proposed the concept of **Segment of Interest (Sol)**
 - New possibility for optimization
- We build an open source testbed
 - Collect real world traces
- We develop accurate Sol detector for MOBA game
 - Up to 0.95 F-measure and 0.87 R-squared score
- Efficient Sol driven resource allocation algorithm
 - Outperforms the current solution up to 5 dB

A SHORT LIVE DEMO OF OUR SYSTEM

Future Work

- Consider device capability in Sol detector
 - Only homogeneous device in evaluation
- Leverage features collected from viewers
 - More fine-grained optimizations
- Resource allocator with finer transcoder control
 - Not only bandwidth, e.g., with FPS, resolution options
- Integrate with a real transcoder
- Provide API for deep integration with game engines

Research Highlight

- **T. Fan-Chiang**, H. Hong and C. Hsu, “Segment-of-Interest Driven Live Game Streaming: Saving Bandwidth Without Degrading Experience”, in **Proc. of ACM Annual Workshop on Network and Systems Support for Games (NetGames '15)**
- Y. Huang, M. Lee, **T. Fan-Chiang** and C. Hsu, “Minimizing Flow Initialization Latency in Software Define Networks”, in **Proc. of Asia-Pacific Network Operations and Management Symposium (APNOMS'15)**
- H. Hong, **T. Fan-Chiang**, C. Lee, K. Chen, C. Huang and C. Hsu, “GPU Consolidation for Cloud Games: Are we There Yet?”, in **Proc. of ACM Annual Workshop on Network and Systems Support for Games (NetGames '14)**
- C. Mao, **T. Fan-Chiang**, M. Lee and C. Hsu, “Taming Flow Initialization Delay in Multi-Site Software Defined Enterprise Networks”, in preparation for **IEEE/ACM Transaction on Networking (TNET)**
- TMM submission (based from this thesis)

THANK YOU

Q & A

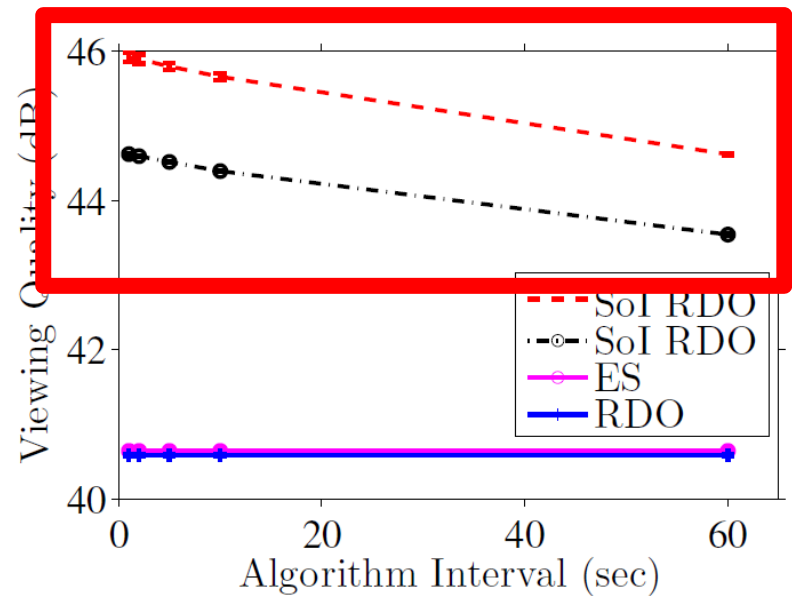
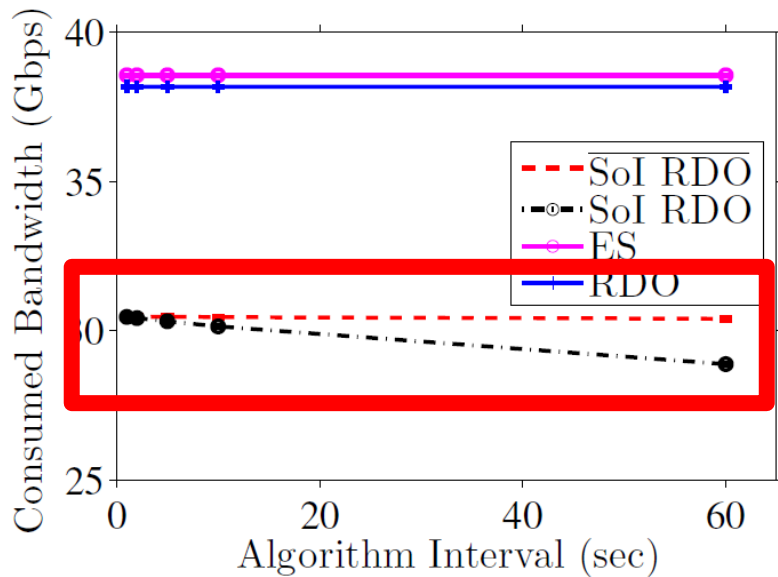
tyfanchiang92@gmail.com



BACKUP SLIDES

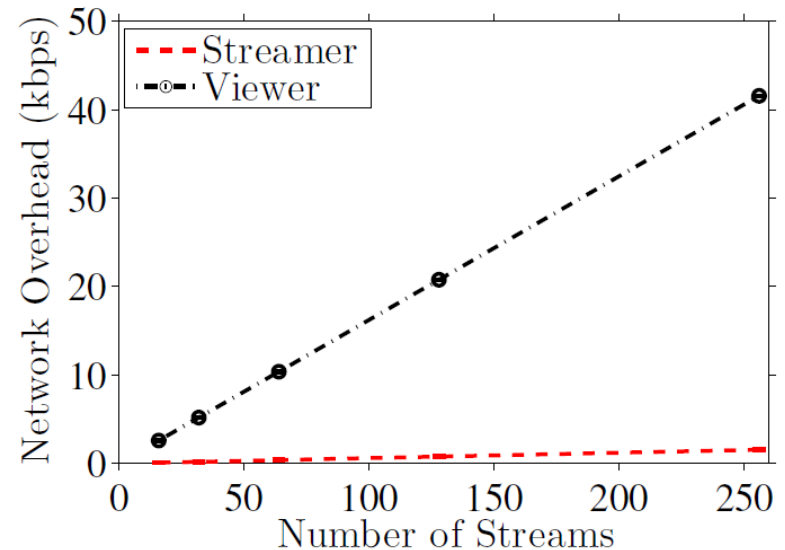
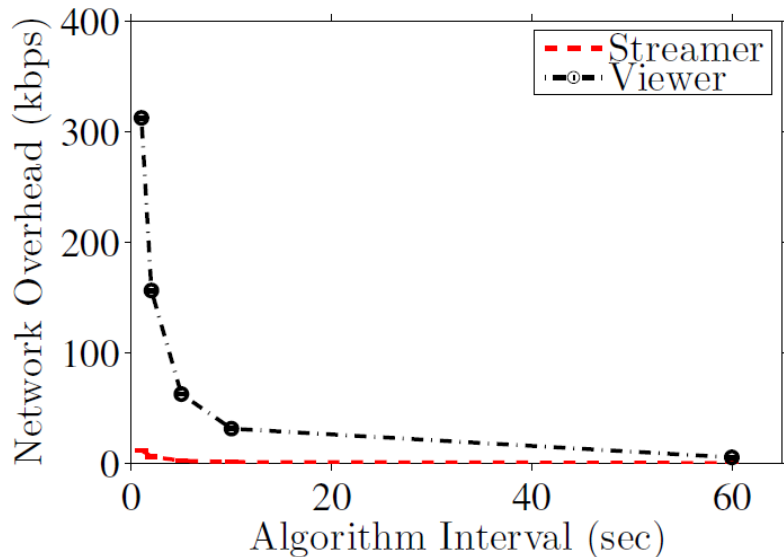
Implication Of Calling Interval

- Slightly better quality and more bandwidth consumption



Negligible Network Overhead

- Just around 300 kbps with 12000+ viewers

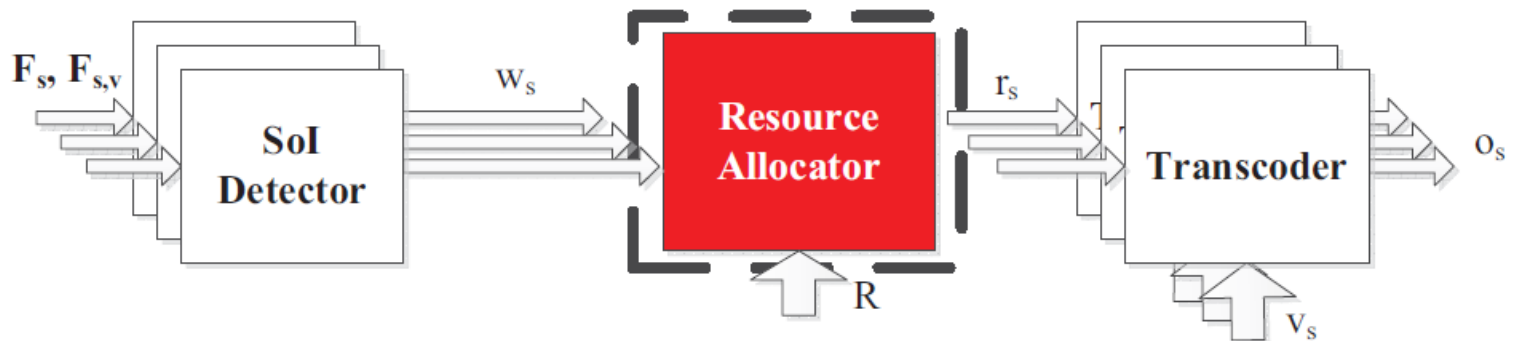


Sol-Driven Streaming Platform

- Develop Sol-driven streaming platform
- Answer three question
 - How to detect Sol efficiently?
 - How to allocation resources among channels?
 - How to perform transcode on the streaming videos?

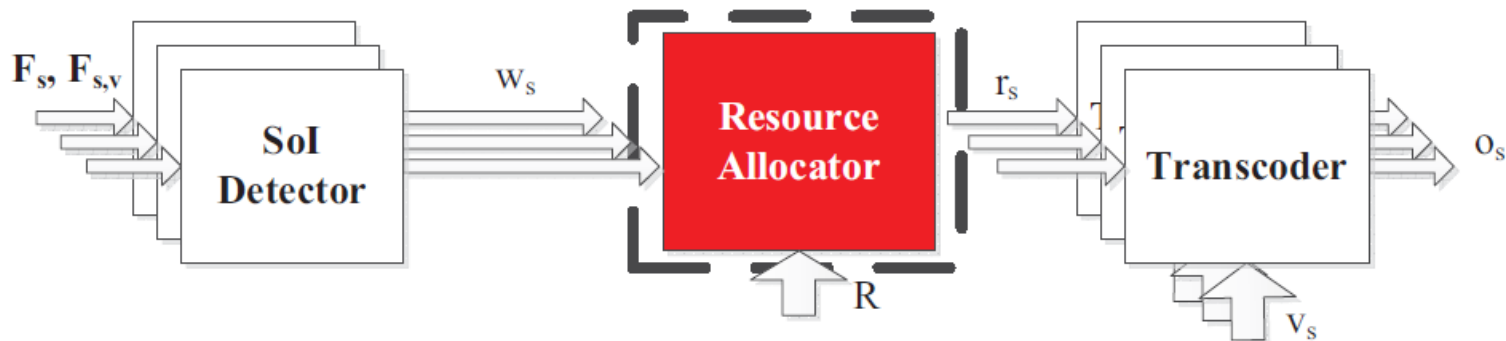
Notations

- The system collects
 - A set of features F_s from streamer s
 - A set of features $F_{s,v}$ from viewer v of streamer s
- Sol detector take F_s and $F_{s,v}$ as input
 - Output Sol weight w_s for current segment

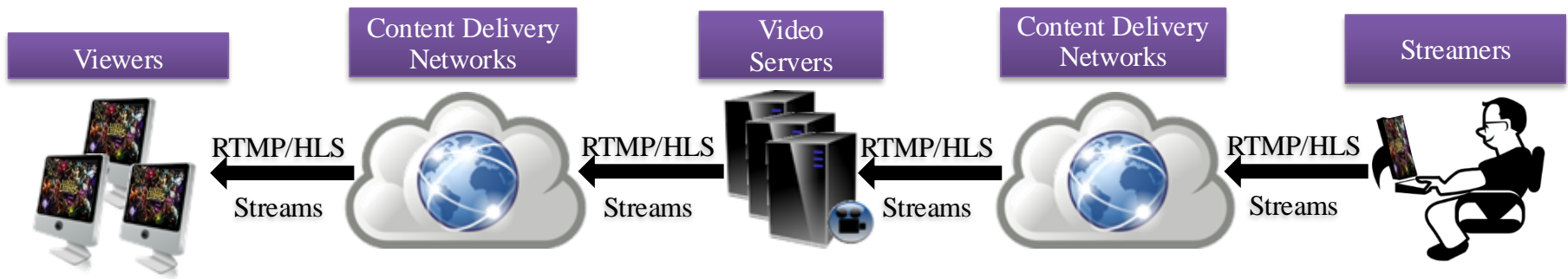


Notations

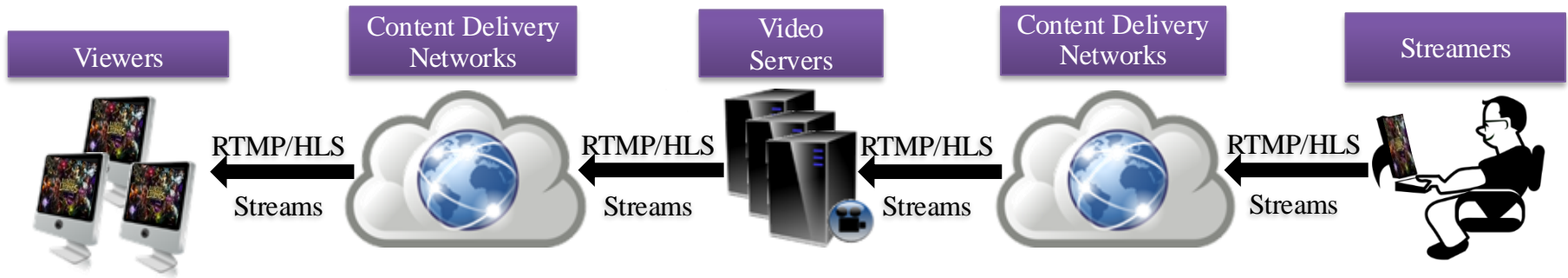
- Resource allocator take all the \mathbf{w}_s and bandwidth capacity \mathbf{R} as input
 - Output \mathbf{r}_s as the bitrate for video segment \mathbf{v}_s from s
- Transcoder use \mathbf{r}_s as parameter to transcode \mathbf{v}_s
 - Output \mathbf{o}_s stream to viewers



WHY DO LAG HAPPENS?

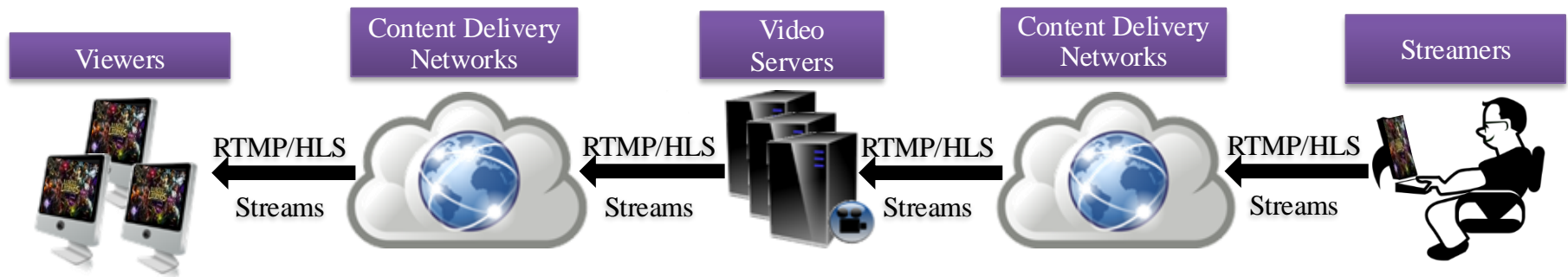


Reason 1: Server Too Busy



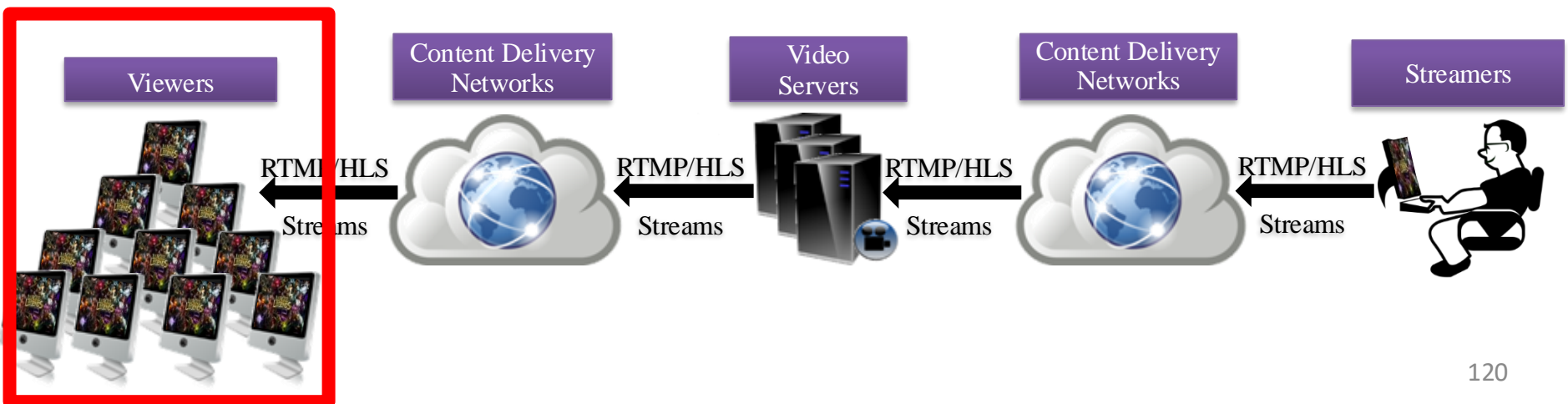
Reason 1: Server Too Busy

- Too many viewers



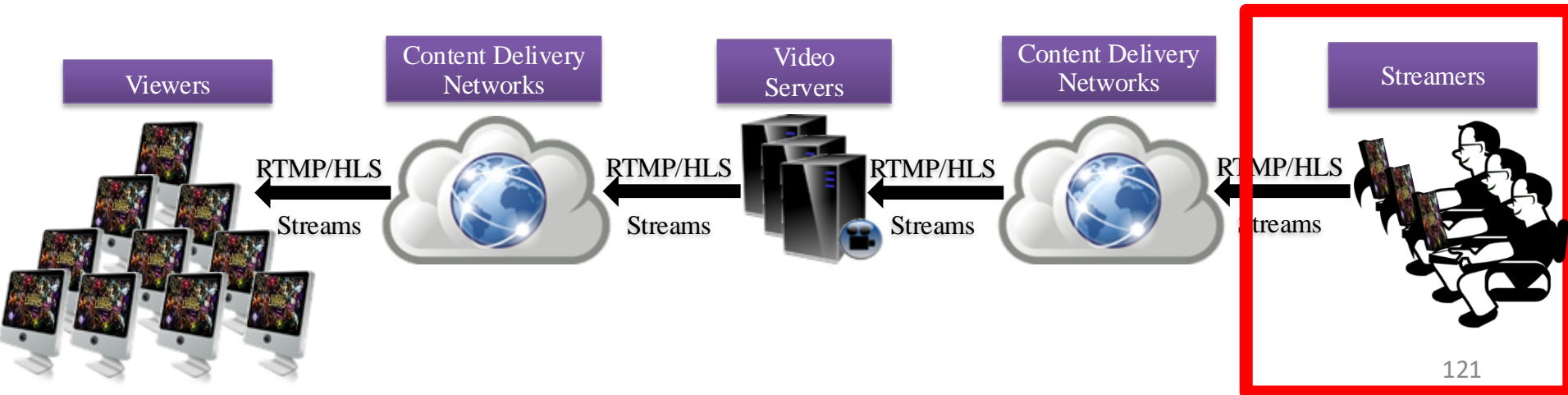
Reason 1: Server Too Busy

- Too many viewers



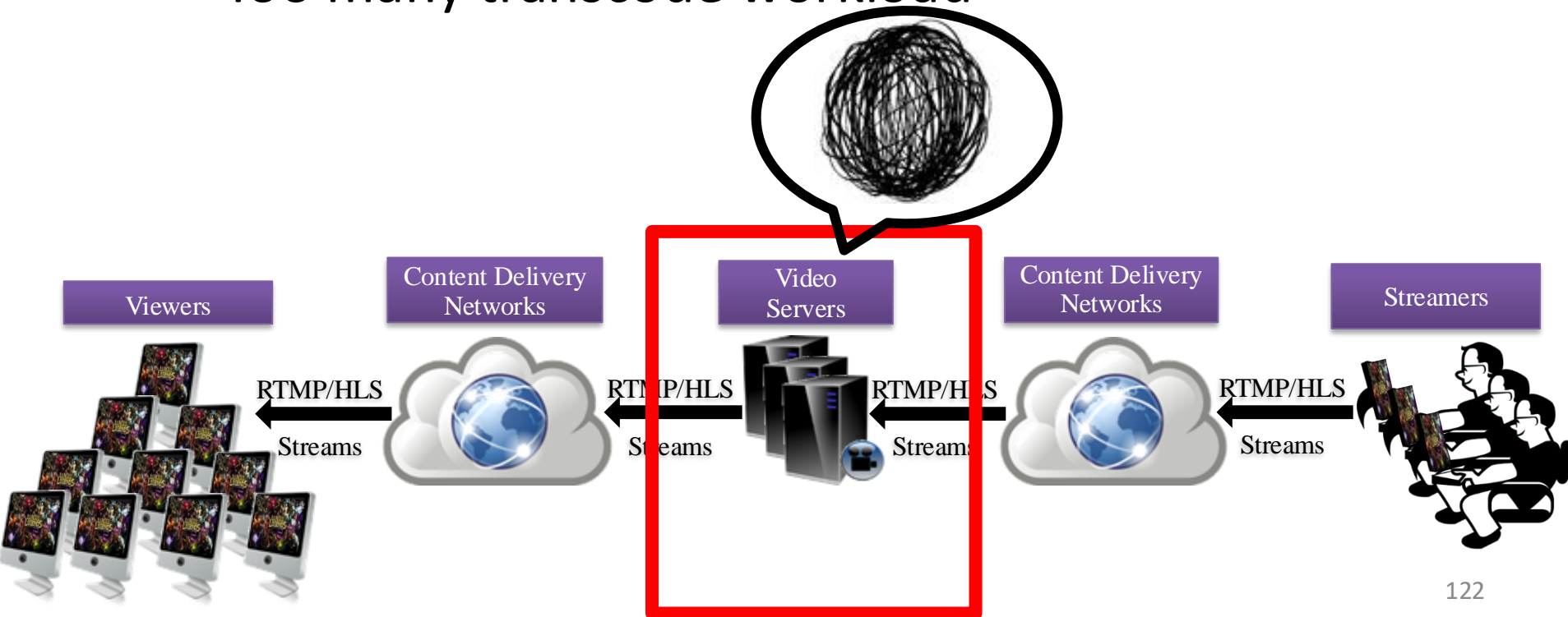
Reason 1: Server Too Busy

- Too many viewers
- Too many streamers
 - Too many transcode workload



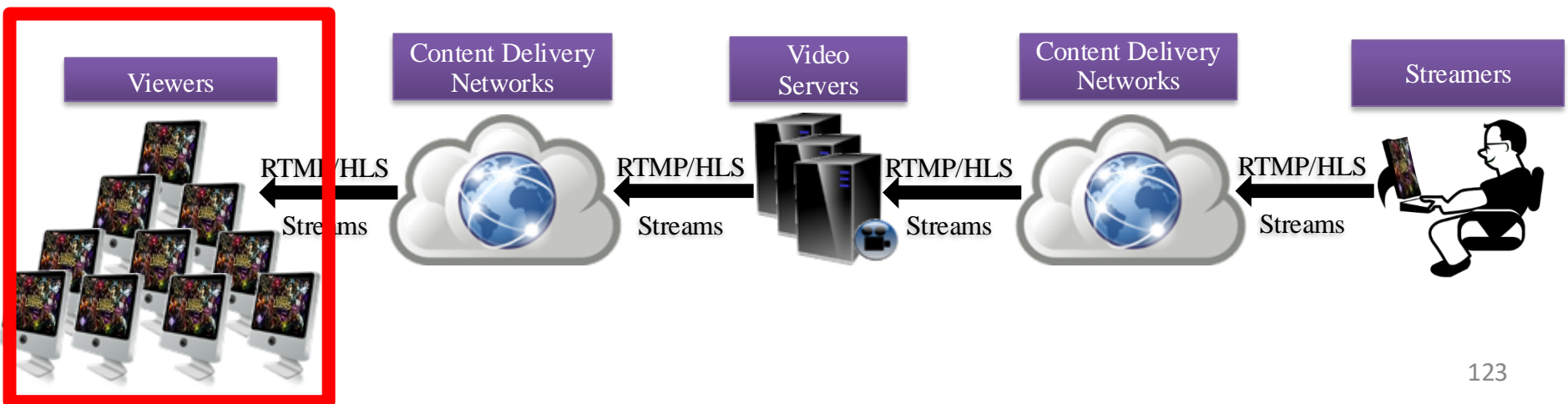
Reason 1: Server Too Busy

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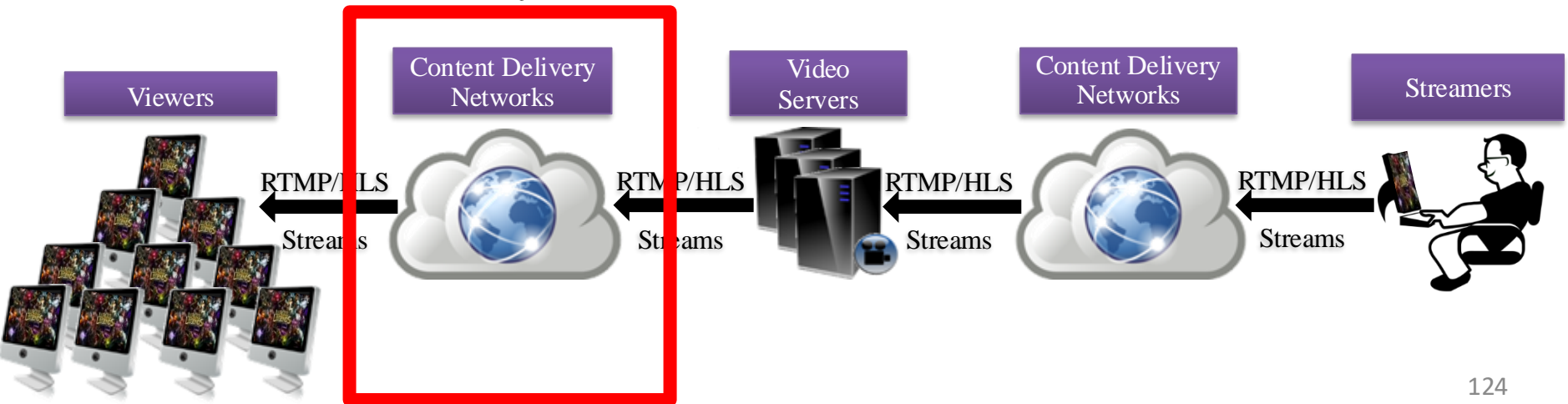
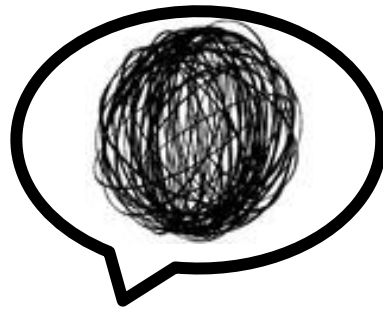
Reason 2: Outbound Bandwidth

- Too many viewers



Reason 2: Outbound Bandwidth

- Too many viewers



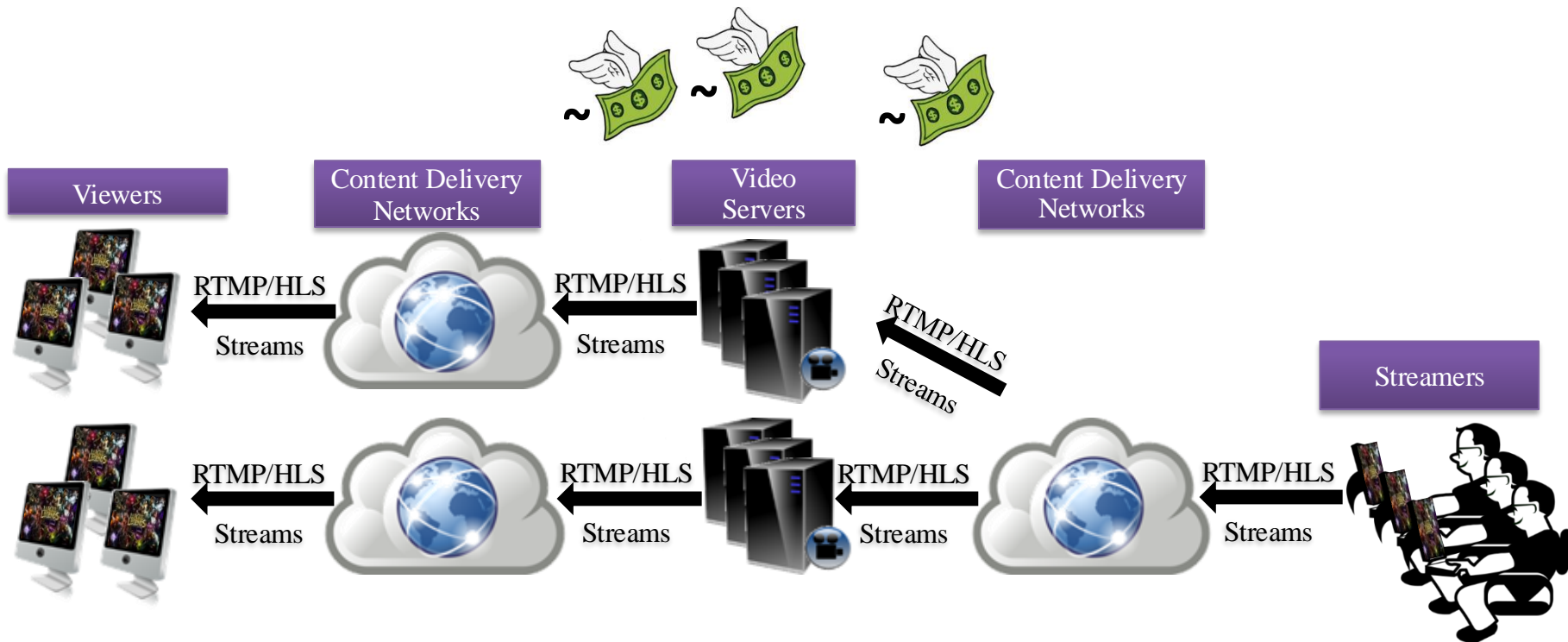
POSSIBLE SOLUTION

Add more resource

Reduce the video quality

Add More Resource

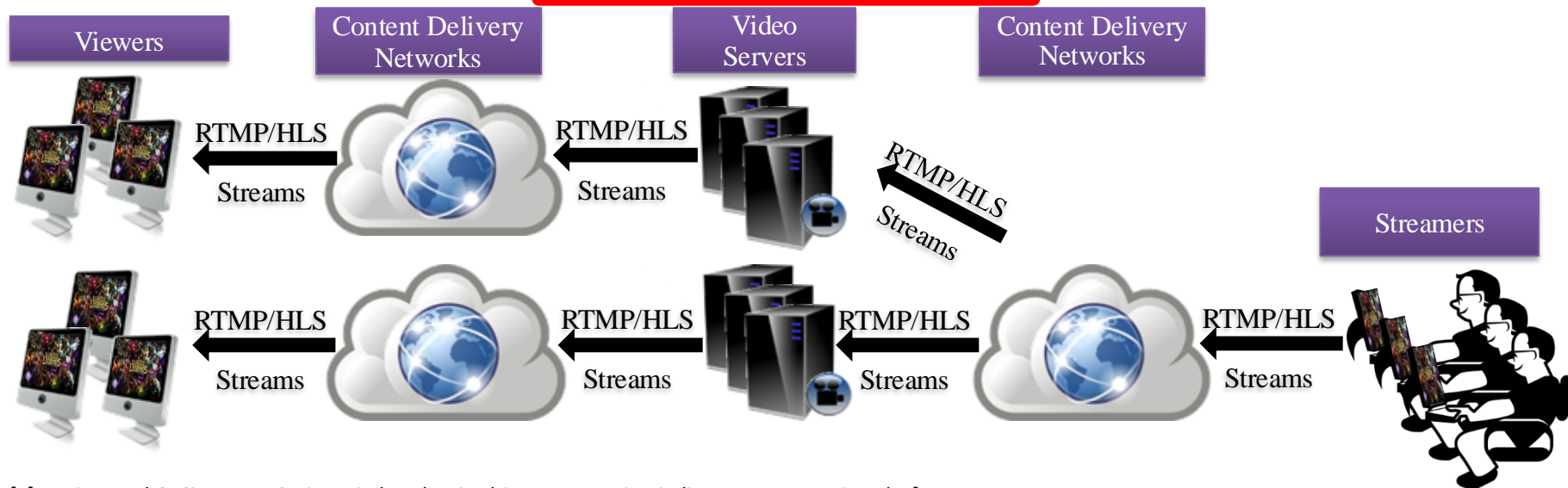
- Twitch upgraded it's transcoding server in 2014
 - Also doubled it's CDN bandwidth in Europe region



Add More Resource

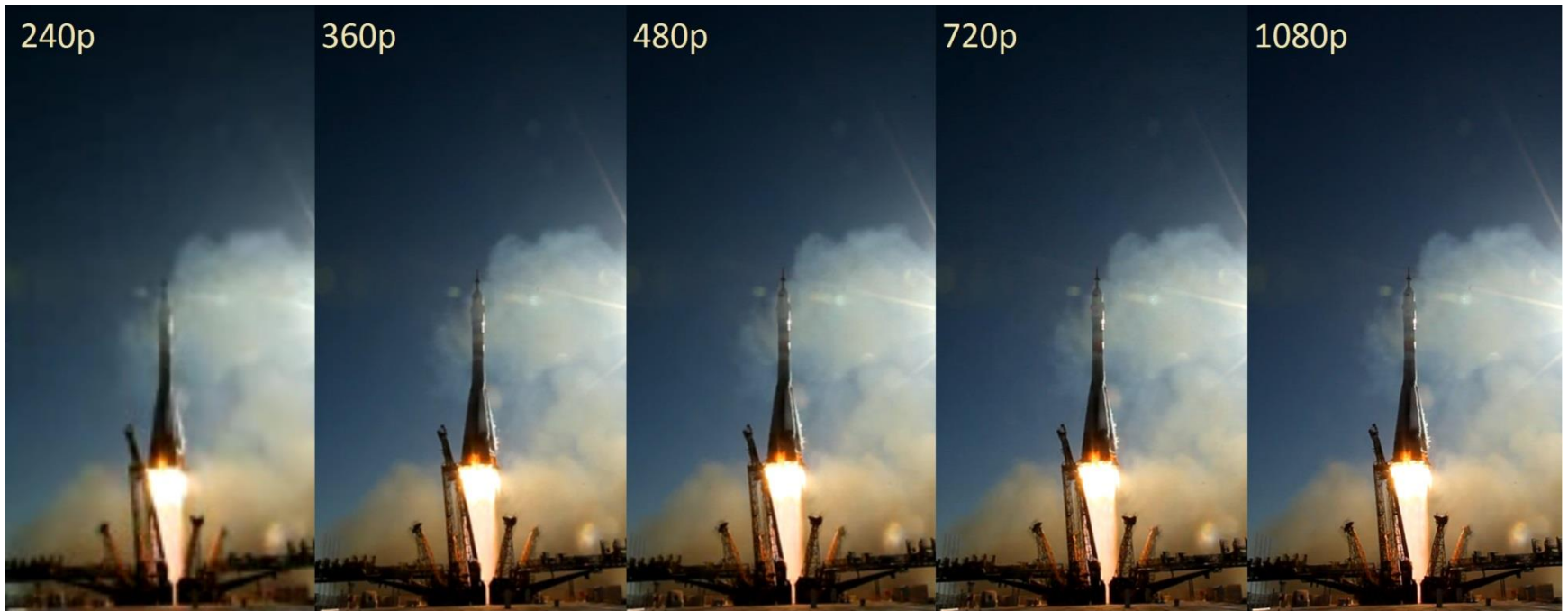
- Twitch have avg 1Tbps bandwidth usage in 2014 [1]
 - May cost as much as \$ 11.34 million per month

Bad for business

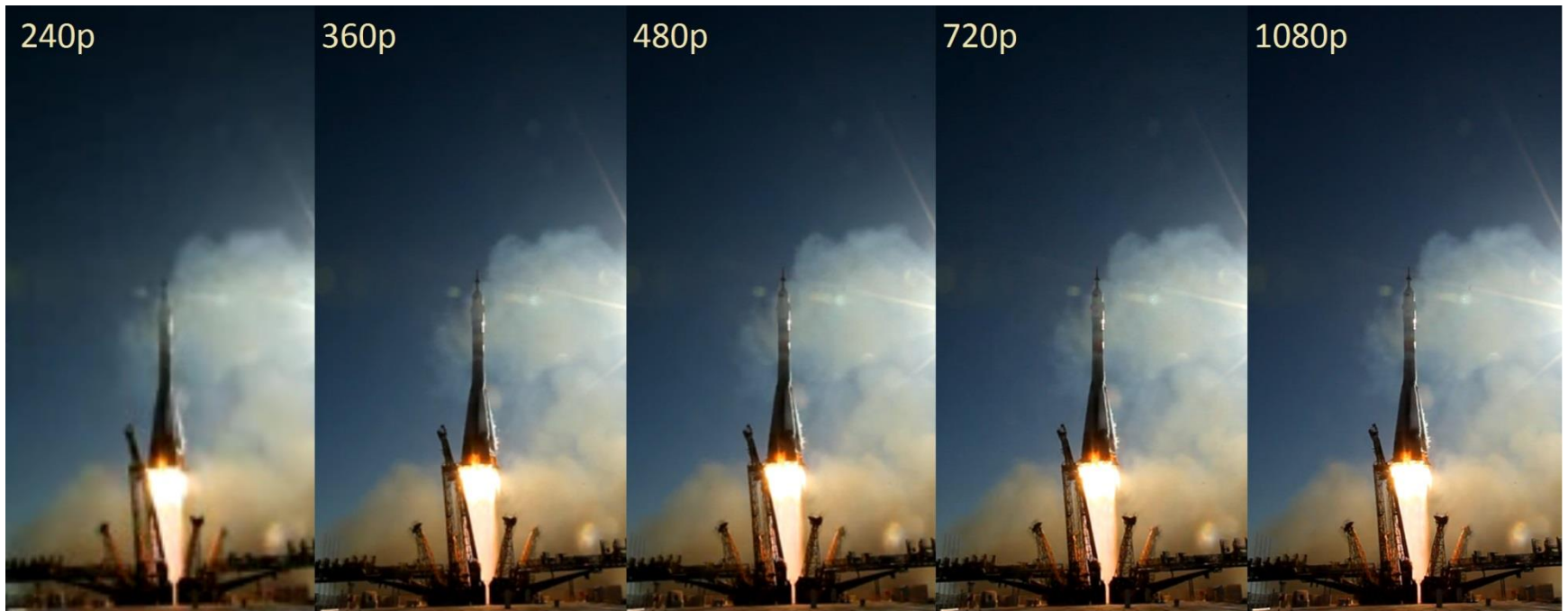


[1] K. Pires and G. Simon. DASH in Twitch: Adaptive bitrate streaming in live game streaming platforms. In *Proc. of ACM Workshop on Design, Quality and Deployment of Adaptive Video Streaming (VideoNext'14)*

Lower Video Quality



Lower Video Quality



How to do this without degrading user experience?

How To Leverage Sol

- Devote more resource into streaming Sol
 - Lower the quality of non-Sol if necessary
- Higher user experience
- Lower bandwidth consumption

How To Leverage Sol

- Devote more resource into streaming Sol
 - Lower the quality of non-Sol if necessary
- Higher user experience
- Lower bandwidth consumption

Wait, how to determine Sol?

Detecting Sol

- The game is running on the machine of streamer
 - Unlike other kind of live streaming
- Collect features in the system
 - CPU/GPU usage, keyboard input event...e.t.c.
 - Use these features to help us determine