Towards Hybrid Cloud-assisted Crowdsourced Live Streaming: Measurement and Analysis

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- Background
- Data description
- Measurement results
- HyCLS: Hybrid design and solution
- Trace-driven simulation and results
- Conclusion and further discussion

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Background-1

Crowdsourced live streaming (CLS) has attracted a substantial amount

of attentions from both industry and academia.



Due to the growth of e-sports games and the development of highperformance personal devices and networks, Twitch became the biggest crowdsourced live streaming platform.

Background-3



Number of broadcasters: **2.1 million** Number of monthly streams: **11 million** Number of monthly unique users: **100 million** Amount of game content that has been streamed: **241 billion minutes**

Periscope

Number of total users: **10 million** Number of daily active users: **2 million** Number of broadcast to date: **200 million** Amount of video content that is streamed daily: **350,000 hours of video**

Background-2

The generic framework of CLS.



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Data Description

Broadcaster datasets

- Total number of views
- Playback bitrate, resolution, partner status
- About **1.5 million** broadcasters

Stream datasets

- The number of viewers per five minutes
- Start time, duration, game name
- About **9 million** streams

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- CLS highlights the event-related live streams with different broadcasters.
- In each CLS event, streaming contents have an event-based correlation, but show broadcaster-based differences.



- We also explore the popularity of broadcasters.
- We plot the highest number of concurrent views against the rank of the broadcasters (in terms of the popularity) in log-log scale.



Fig. 3 Broadcasters rank ordered by popularity

• The distribution of live duration



Fig. 4 The distribution of live duration

The total duration of all unpopular streams in one month is nearly 830 years, while the total duration of popular streams is only 310 years.

The daily activity of two broadcasters



Fig. 5a Popular broadcaster sample



Fig. 5b Unpopular broadcaster sample

A: regular live schedule, stable live duration, a large number of viewers. B: dynamic schedule and duration, a few number of viewers

• Broadcaster arrivals per five minutes.



CLS features

CLS feature:

- 1. Live sources
 - Controlled by broadcasters vs. Managed by service providers

• 2. Service cost

• Storage/bandwidth/.computation continually vs. storage

How about the resource consumption of hosting these unpopular broadcasters in Twitch?

The effectiveness of resource consumption.



R. Aparicio-Pardo, K. Pires, A. Blanc, and G. Simon. Transcoding live adaptive video streams at a massive scale in the cloud. In *ACM MMSys*, 2015.

Analysis

- Crowdsourced live events
- Unpopular broadcasters
 - Dynamic schedule
 - Unstable live duration
 - o Frequent arrival
- Dedicated resource consumption
 - o Bandwidth
 - Computation
- Can we use public cloud to assist existing private datacenter ?

• RTT comparison between public cloud and private data center



EC2 instances do not increase RTT significantly even in the degradation of networks.

We can use EC2 instance to ingest the live streams of broadcasters without extra latency.

• Performance comparison between different instance types (m3.medium vs. m3.large)



Source(i.e., 1080P, 3200Kbps) 720P (1500Kbps), 480P (800Kbps) 360P (500Kbps) 228P (200Kbps)

• Performance of different types of instance (m3.large).



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HyCLS Design

• Our goal is to assign broadcasting workloads cost-effectively in hybrid-design.



HyCLS-Initial Offloading

- Stable Index reflects the similarity of b's resource consumption in recent n days.
- Update threshold periodically to determine the initial offloading.

$$SI^{(b)} = \begin{cases} \frac{1}{n} \sum_{i=2}^{n} \frac{\sum_{j=1}^{m} d_{i,j}^{(b)} \cdot d_{i-1,j}^{(b)}}{\sum_{j=1}^{m} d_{i-1,j}^{(b)}} & \text{if } \sum_{j=1}^{m} d_{i-1,j}^{(b)} \neq 0\\ 0 & \text{otherwise} \end{cases}$$

Ingesting Redirection & Transcoding Schedule

Ingesting Redirection & Transcoding Schedule

$$\underset{e \in E^{(t)}}{\operatorname{Maximize}} F(A^{(t)}) = \underset{\substack{b \in e \\ r \in R}}{\min} \{ U^{(t)}(b, r) \}$$

Resource Availability Constraints:

$$\forall r \in R, W_r^{(t)} \le \mathbb{W}_r$$
$$\forall r \in R, C_r^{(t)} \le \mathbb{C}_r$$

Budget Constraints:

$$\sum_{r \in R} \frac{W_r^{(t)}}{\mathcal{W}} \cdot Cost_w(r) \cdot I(r) \le K_w$$
$$\sum_{r \in R} \frac{C_r^{(t)}}{\mathcal{C}} \cdot Cost_c(r) \cdot I(r) \le K_c$$

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H. Wang, R. Shea, X. Ma, F. Wang, and J. Liu. On design and performance of cloud-based distributed interactive applications. In *IEEE ICNP, 2014.*

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Simulation setup

- Partner status in Twitch
- Homogenous public instances (m3.large)
- α = 1 and β = 0.011: make the gain G(t)(·) ∈ [0, 1] with current Twitch broadcast latency interval.
- n = 2: Stable index is calculated by using the data-trace during latest two days.

Simulation Results-1



- Views-based (LB-V): only considers the current number of views in different live streams;
- Computation-based (LB-C): migrates workload based on the consumption of computation resources.

. Our HyCLS-based approach has the lowest cost, decreasing 16.9%-19.5% of LB-C approach and 17.8%-20.4% of LB-V approach.

Simulation Results-2



The daily lease cost performs the weekly pattern and provide elastic workload provisioning costeffectively. Moreover, more than 30% of broadcasters are migrated to the public cloud in every day.

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Conclusion and Further Discussion

- The characteristics of broadcasters in Twitch
- The challenges of bandwidth and computation comsuption
- Hybrid-cloud design and solution

- Re-design initial offloading strategy
- Amazon EC2 and PlanetLab-based practical deployment

Thank You!

Q&A