

Using AMOV ESoV Algorithms And Sharing In The Clouds

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Abstract—Demands on video traffic over mobile networks have been soaring, the wireless link capacity cannot keep up with the traffic demand. The gap between the traffic demand and the link capacity, along with time-varying link conditions, results in poor service quality of video streaming over mobile networks such as long buffering time and intermittent disruptions. Leveraging the cloud computing technology, we propose a new mobile video streaming framework, dubbed AMES-Cloud, which has two main parts: adaptive mobile video streaming (AMoV) and efficient social video sharing (ESoV). AMoV and ESoV construct a private agent to provide video streaming services efficiently for each mobile user. For a given user, AMoV lets her private agent adaptively adjust the streaming flow with a scalable video coding technique based on the feedback of link quality. Likewise, ESoV monitors the social network interactions among mobile users, and their private agent tries to prefetch video content in advance. We implement a prototype of the AMES-Cloud framework to demonstrate its performance. It is shown that the private agents in the cloud can effectively provide the adaptive streaming, and perform video sharing (i.e., prefetching) based on the social network analysis.

Index—streaming, cloud computing, mobile networks, scalable video coding, social video sharing.

Terms—Adaptive video

1. INTRODUCTION

The past decade, increasingly more traffic is accounted by video streaming and downloading. In particular, video streaming services over mobile networks have become prevalent over the past few years [1]. While the video streaming is not so challenging in wired networks, mobile networks have been suffering from video traffic transmission over scarce bandwidth of wireless links. Despite network operators' desperate effort to enhance the wireless link bandwidth (e.g., 3G and LTE), soaring video traffic demands from mobile users are rapidly overwhelming the wireless link capacity. While receiving video streaming traffic via 3G/4G mobile networks, mobile users often suffer from long buffering time and intermittent disruptions due to the limited bandwidth and link condition fluctuation caused by multi-path fading and user mobility [2]–[4]. Thus, it is crucial to improve the service quality of mobile video streaming while using the networking

and computing resources efficiently [5]–[8]. Recently there have been many studies on how to improve the service quality of mobile video streaming on two aspects:

- **Scalability:** Mobile video streaming service should support wide spectrum of mobile devices; they have different video resolutions, different computing powers, different wireless links (like 3G and LTE) and soon. Also, the available link capacity of a mobile device may vary over time and space depending on its signal strength, other user traffic in the same cell, and link condition variation. Storing multiple versions (with different bitrates) of the same video content may incur high overhead in terms of storage and communication.

Adaptability: Traditional video streaming techniques designed by considering relatively stable traffic links between servers and users, perform poorly in mobile environments [2]. Thus the fluctuating wireless link status should be properly dealt with to provide "tolerable" video streaming services.

Cloud computing techniques are poised to flexibly provide scalable resources to content/service providers, and process of loading to mobile users [13]–[19]. Thus, cloud data centers can easily provision for large-scale real-time video services as investigated in [9], [20]. Several studies on mobile cloud computing technologies have proposed to generate personalized intelligent agents for servicing mobile users, e.g., Cloudlet [21] and Stratus [22]. This is because, in the cloud, multiple agent instances (or threads) can be maintained dynamically and efficiently depending on the time-varying user demands.

Recently social network services (SNSs) have been increasingly popular. There have been proposals to improve the quality of content delivery using SNSs [23], [24]. In SNSs, users may share, comment or re-post videos among friends and members in the same group, which implies a user may watch a video that her friend has recommended (e.g., [24]). Users in SNSs can also follow famous and popular users based on their interests (e.g., an official Facebook or Twitter account that shares the newest pop music videos), which is likely to be watched by its followers. In this paper, we design an adaptive video streaming and prefetching framework for mobile users with the above objectives in mind, dubbed AMES-Cloud. AMES-Cloud constructs a private agent for each mobile user in cloud computing environments, which is used by its two main parts: (i) **AMoV** (adaptive mobile video streaming), and **ESoV** (efficient social video sharing).

The rest of the paper is organized as follows. We first introduce related work in Section II, and explain the AMES-Cloud

framework in Section III. The adaptive video streaming service and the efficient social video sharing will be detailed in Sections IV and V, respectively. Then the operations of AMES-Cloud is illustrated in Section VI. Finally, we evaluate the prototype implementation in Section VII, and conclude the paper in Section VIII.

II. RELATED WORK

A. Adaptive Video Streaming Techniques

In the adaptive streaming, the video traffic rate is adjusted on the fly so that a user can experience the maximum possible video quality based on his or her link's time-varying bandwidth capacity [2]. There are mainly two types of adaptive streaming techniques, depending on whether the adaptivity is controlled by the client or the server. The Microsoft's Smooth Streaming [27] is a live adaptive streaming service which can switch among different bitrate segments encoded with configurable bitrates and video resolutions at servers, while clients dynamically request videos based on local monitoring of link quality. Adobe and Apple also developed client-side HTTP adaptive live streaming solutions operating in the similar manner. There are also some similar adaptive streaming services where servers control the adaptive transmission of video segments, for example, the Quava live Adaptive Streaming. However, most of these solutions maintain multiple copies of the video content with different bitrates, which brings huge burden of storage on the server. Regarding rate adaptation controlling techniques, TCP-friendly rate control methods for streaming services over mobile networks are proposed [28], [29], where TCP throughput of a flow is predicted as a function of packet loss rate, round trip time, and packet size. Considering the estimated throughput, the bitrate of the streaming traffic can be adjusted. A rate adaptation algorithm for conversational 3G video streaming is introduced by [30]. Then, a few cross-layer adaptation techniques are discussed [31], [32], which can acquire more accurate information of link quality so that the rate adaptation can be more accurately made. However, the servers have to always control and thus suffer from large workload. Recently the H.264 Scalable Video Coding (SVC) technique has gained momentum [10]. An adaptive video streaming system based on SVC is deployed in [9], which studies the real-time SVC decoding and encoding at PC servers.

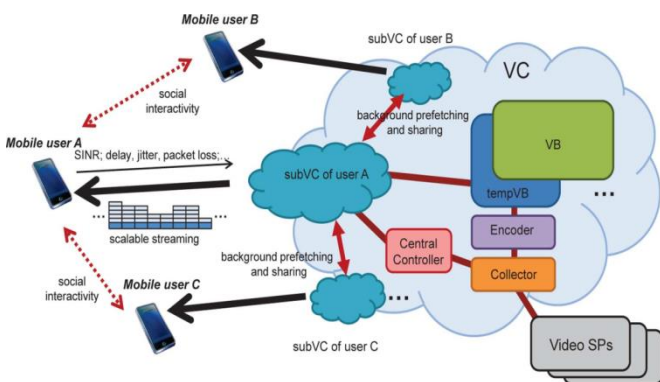


Fig. 1. Illustration of the AMES-Cloud framework with the Video Cloud (VC), subVCs for mobile users, the Video Base (VB), and the Video Service Providers (SPs).

III. AMES-CLOUD FRAMEWORK

In this section we explain the AMES-Cloud framework including the Adaptive Mobile Video Streaming (AMoV) and the Efficient Social Video Sharing (ESoV). As shown in Fig. 1, the whole video storing and streaming system is the cloud called the Video Cloud (VC). In the VC, there is a large-scale video base (VB), which stores the most of the popular video clips for the video service providers (VSPs). A temporal video base (tempVB) is used to cache new candidates for the popular videos, while tempVB counts the access frequency of each video. The VC keeps running a collector to seek videos which are already popular in VSPs, and will re-encode the collected videos into SVC format and store into tempVB first. By this 2-tier storage, the AMES-Cloud can keep serving most of popular videos eternally. Note that management work will be handled by the controller in the VC. Specialized for each mobile user, a sub-video cloud (subVC) is created dynamically if there is any video streaming demand from the user.

B. Mobile Cloud Computing Techniques

The cloud computing has been well positioned to provide video streaming services, especially in the wired Internet because of its scalability and capability [13]. For example, the quality-assured bandwidth auto-scaling for VoD streaming based on the cloud computing is proposed [14], and the CALMS framework [33] is a cloud-assisted live media streaming service for globally distributed users. However, extending the cloud computing-based services to mobile environments requires more factors to consider: wireless link dynamics, user mobility.

IV. AMOV: ADAPTIVE MOBILE VIDEO STREAMING

As shown in Fig. 2, traditional video streams with fixed bit rates cannot adapt to the fluctuation of the link quality. For a particular bitrate, if the sustainable link bandwidth varies much, the video streaming can be frequently terminated due to the packet loss.

In SVC, a combination of the three lowest scalability is called the Base Layer (BL) while the enhanced combinations are called Enhancement Layers (ELs). To this regard, if BL is guaranteed to be delivered, while more ELs can be also obtained when the link can afford, a better video quality can be expected.

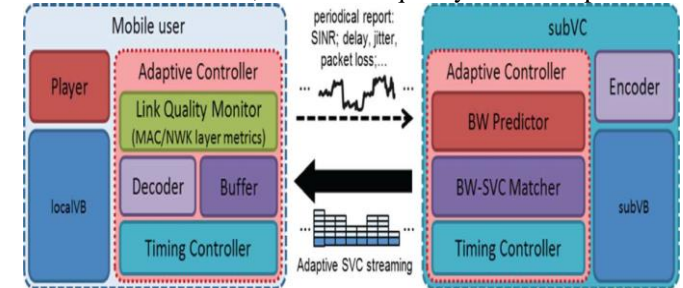


Fig 2. traditional video streams with fixed bit rates

V. ESOV: EFFICIENT SOCIAL VIDEO SHARING

A. Social Content Sharing

In SNSs, users subscribe to known friends, famous people, and particular interested content publishers as well; also there are various types of social activities among users in SNSs, such

as direct message and public posting. For spreading videos in SNSs, one can post a video in the public, and his/hersubscribers can quickly see it; one can also directly recommend a video to specified friend(s); furthermore one can periodically get noticed by subscribed content publisher for new or popular videos.

We classify the social activities in current popular SNSs into three kinds, regarding the impact of the activities and the potential reacting priority from the point of view of the recipient:

- **Subscription:** Like the popular RSS services, a user can subscribe to a particular video publisher or a special video collection service based on his/her interests. This interest-driven connectivity between the subscriber and the video publisher is considered as “median”, because the subscriber may not always watch all subscribed videos.

- **Direct recommendation:** In SNSs, a user directly recommends a video to particular friend(s) with a short message. The recipients of the message may watch it with very high probability. This is considered as “strong”.

- **Public sharing:** Each user in SNSs has a timeline-based activity stream, which shows his/her recent activities. The activity of a user watching or sharing a video can be seen by his/her friends (or followers). We consider this public sharing with the “weak” connectivity among users, because not many people may watch the video that one has seen without direct recommendation.

TABLE I SOCIAL ACTIVITIES AND BACKGROUND PUSHING STRATEGIES

	Direct recommendation	Subscription	Public sharing
VB → subVB	All	Parts	Little
subVB → locVB (via Wi-Fi)	All	Parts	Little
subVB → locVB (via 3G/4G)	Parts	Little	None

B. Prefetching Levels

Different strengths of the social activities indicated different levels of probability that a video will be soon watched by the recipient. Correspondingly we also define three prefetching levels regarding the social activities of mobile users:

- **“Parts”:** Because the videos that published by subscriptions may be watched by the subscribers with a nothigh probability, we propose to only push a part of BL and ELs segments, for example, the first 10% segments.
- **“All”:** The videos shared by the direct recommendations will be watched with a high probability, so we propose to prefetch the BL and all ELs, in order to let the recipient(s) directly watch the video with a good quality, without any buffering.

VII. IMPLEMENTATION AND EVALUATION

We evaluate the performance of the AMES-Cloud framework by a prototype implementation. We choose the U-cloud server (premium) in the cloud computing service offered by Korean Telecom, and utilize the virtual server with 6 virtual CPU cores (2.66GHz) and 32GB memory, which is fast enough for en-

coding 480P (480 by 720) video with H.264 SVC format in 30fps at real time [9]. In the cloud, we deploy our server application based on Java, including one main program handling all tasks of the whole VC, while the program dynamically initializes, maintains and terminates instances of another small Java application as private agents for all active users. We implement the mobile client at a mobile phone, Samsung Galaxy II, with android system version 4.0. The mobile data service is offered by LG LTE network, while in some uncovered areas the 3G network is used. Note that we still use “3G” to indicate the general cellular network. We test in the downtown area, so the practical bandwidth of the mobile link is not as high as we expected, but this won't impact our experiment results.

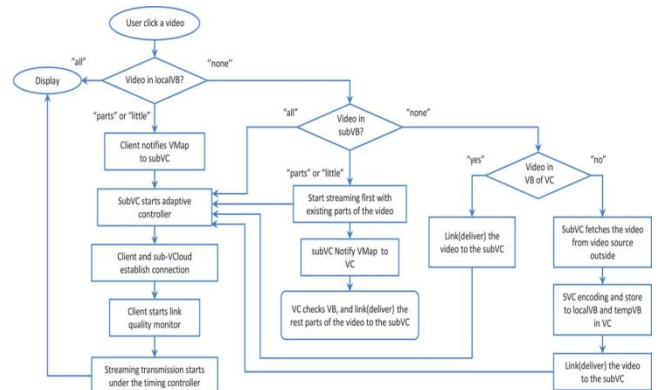


Fig. 3. Working flow of video streaming in the subVC and VC of AMES-Cloud framework.

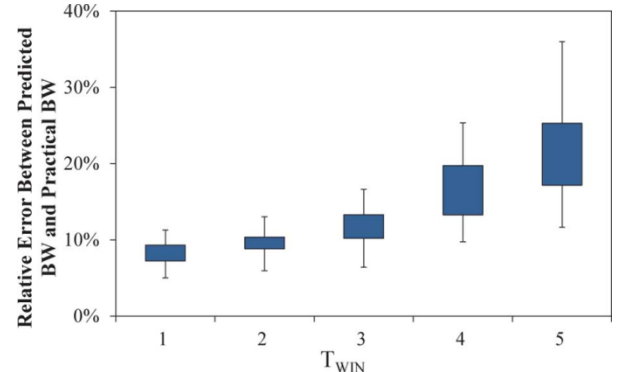


Fig. 4. Relative errors between predicted bandwidth and practical bandwidth.

Firstly we examine whether there is a deep relationship between the measured bandwidth of last time window and the practical bandwidth of next time window (goodput by Kbps). We test the video streaming service via cellular link, and move the device around in the building to try to change the signal quality.

We evaluate how H.264 SVC works in AMES-Cloud framework regarding the above mentioned SVC resolution configurations. As shown in Fig. 7(a), because of the strong computational capacity by the cloud computing, the encoding speed is fast. The best resolution configuration “with 5 second temporal segmentation scheme requires about 560m s for encoding. For shorter intervals of T_{II=II}, the encoding delay is very small under 50ms.

VIII. CONCLUSION

In this paper, we discussed our proposal of an adaptive mobile video streaming and sharing framework, called AMES-Cloud, which efficiently stores videos in the clouds (VC), and utilizes cloud computing to construct private agent (subVC) for each mobile user to try to offer “non-terminating” video streaming adapting to the fluctuation of link quality based on the Scalable Video Coding technique. Also AMES-Cloud can further seek to provide “non-buffering” experience of video streaming by background pushing functions among the VB, subVBs and local VB of mobile users. We evaluated the AMES-Cloud by prototype implementation and show that the cloud computing technique brings significant improvement on the adaptivity of the mobile streaming. We will carry out large-scale implementation and with serious consideration on energy and price cost. In the future, we will also try to improve the SNS-based prefetching, and security issues in the AMES-Cloud.

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