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# A QoS Guaranteed Technique for Cloud Applications Based on Software Defined Networking

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ABSTRACT Due to the centralized control, network-wide monitoring and flow-level scheduling of software-defined-networking (SDN), it can be utilized to achieve quality of service (QoS) for cloud applications and services, such as voice over IP, video conference, and online games. However, most existing approaches stay at the QoS framework design and test level, while few works focus on studying the basic QoS techniques supported by SDN. In this paper, we enable SDN with QoS guaranteed abilities, which could provide end-to-end QoS routing for each cloud user service. First of all, we implement an application identification technique on SDN controller to determine required QoS levels for each application type. Then, we implement a queue scheduling technique on SDN switch. It queues the application flows into different queues and schedules the flows out of the queues with different priorities. At last, we evaluate the effectiveness of the proposed SDN-based QoS technique through both theoretical and experimental analysis. Theoretical analysis shows that our methods can provide differentiated services for the application flows mapped to different QoS levels. Experiment results show that when the output interface has sufficiently available bandwidth, the delay can be reduced by 28% on average. In addition, for the application flow with the highest priority, our methods can reduce 99.99% delay and increase 90.17% throughput on average when the output interface utilization approaches to the maximum bandwidth limitation.

**INDEX TERMS** Software defined networking, cloud computing, QoS, applications.

# I. INTRODUCTION

SDN enhances network flexibility and scalability by separating control plane from data plane. It is progressively dominating the dynamic management for timely network trouble shooting and fine grained traffic scheduling in the data center network infrastructure [1], [2], which is the foundation for building today's cloud computing services. More and more multimedia applications are deployed on cloud. So cloud users can access the multimedia data from any geographical location. As the cloud providers, they should meet QoS requirements for each cloud user application.

The Internet uses the traditional best-effort service model to route traffic flows. It cannot provide differentiated services for cloud applications. Then, some enforcement service models are proposed, such as Integrated Services (IntServ) [3], Differentiated Service (Diffserv) [4] and Multi Protocol Label Switching (MPLS) [5]. However, these models are lack

of traffic control, or are hard to configure, manage and troubleshoot. Compared with the traditional best-effort service model and the enforcement service models, SDN has the potential to provide a better QoS guarantee for cloud applications and services due to its centralized control, network-wide monitoring and flow-level scheduling [6], [7], [22].

Existing studies have tried to use SDN to provide end-to-end QoS routing [8] or multipath routing [9] for multimedia applications. However, most existing approaches stay at the QoS framework design and test level, while few works focus on studying the basic QoS techniques supported by SDN. Different from previous works, we implement and verify a SDN-based QoS guaranteed technique for cloud applications. We combine application identification with queue scheduling to meet the application flows with different required QoS levels. Our methods supply a basic QoS technique for end-to-end QoS routing, as well as make a supplement to existing



approaches. The main contributions of this paper are summarized as follows.

- (1) We implement an application identification technique on SDN controller based on *C4.5 decision tree*. In addition to identifying application types, it also determines the required QoS level for each type of application and issues corresponding matching rules to SDN switches to meet the QoS requirements of different applications.
- (2) We implement a queue scheduling technique to allow delay-sensitive data to be dequeued and sent first. We create multi-queues for each output interface of the switch, including the *Expedited Forwarding (EF)* queue with the highest priority, the *Assured Forwarding (AF)* queue with the medium priority and the *Best Effort (BE)* queue with the lowest priority. We then implement two algorithms to queue the packets into the queues of each output interface and schedule the packets out of the queues with different priorities.
- (3) We evaluate the proposed SDN-based QoS technique through both theoretical and experimental analysis. Theoretical analysis shows that our methods can provide differentiated services for the application flows mapped to different QoS levels. And experimental results show that when the output interface has sufficiently available bandwidth, the delay can be reduced by 28% on average. In addition, for the application flow with highest priority, our methods can reduce 99.99% delay and increase 90.17% throughput on average when the output interface utilization approaches to the maximum bandwidth limitation.

According to *OpenFlow* switch specifications [32], it supports limited QoS futures by the *Hierarchical Token Based* (*HTB*) [33] queuing technique and the *Hierarchical Fair Sequence Curve* (*HFSC*) [34] queuing technique. Taking *HTB* as an example, it only allows to configure guaranteed minimum rate and limited maximum rate for the flows. And by default, each output interface has only one *First In Firs Out* (*FIFO*) queue. In this paper, we implement multi- *FIFO* queues for each output interface, and use *HTB* to achieve both rate-limiting and priority-scheduling.

The remainder of this paper is organized as follows. Related work is presented in Section II. Section III describes the SDN-based QoS technique. Section IV analyzes the effectiveness of the queue scheduling technique with the queuing theory. We conduct an experimental evaluation the proposed SDN-based QoS technique in Section IV. Section V concludes the whole paper and gives some future remarks.

#### **II. RELATED WORK**

SDN has become a promising network technology and it has been deployed in data center networks [1], [2]. Benefiting from the centralized control, network-wide monitoring and flow-level scheduling, SDN provides the opportunity to

achieve a better QoS guarantee for cloud applications and services.

Jeong et al. [10] extended the Network Operating System [11] for SDN with the QoS-aware ability of resource discover, routing computation, fault notification and restoration, etc. Wallner et al. [12] showed a basic idea to realize QoS through adding QoS modules and tools to the Floodlight controller [13]. Ishimori et al. [14] proposed the control of multiple packet schedulers to improve QoS for SDN. Bueno et al. [15] extended OpenNaaS [16] framework with SDN capacity to provides dynamic QoS control. Jarschel et al. [17] proposed an application-aware SDN approach to provide QoE for YouTube video streaming. They evaluated which types of application information can be exploited to enhance QoE. Bari et al. [18] proposed an autonomic SDN-based QoS policy enforcement framework by specifying QoS-based Service Level Agreements. Gorlatch et al. [19] used SDN to address the dynamic network demand and improve the QoS of real-time online interactive applications. Akella et al. [6] studied OoS-guaranteed bandwidth allocation for cloud users based on SDN. They introduced queuing techniques and considered the performance metrics of response time and the number of hops. Tomovic et al. [8] presented a new SDN control framework for QoS provisioning. The framework could provide required QoS level for multimedia applications automatically and flexibly. Seddiki et al. [20] proposed a SDN-based approach to achieve per-flow QoS for broadband access networks. Yan et al. [9] proposed an SDN-based multipath QoS solution, which could reduce delay, increase throughput and quickly reroute traffic from path failure. Sieber et al. [21] proposed a Network Services Abstraction Layer on top of the network control and management plane. They then introduced a unified data model for both SDN and legacy devices to achieve QoS for time-critical tasks. Dwarakanathan et al. [22] proposed a framework to meet the QoS requirements of cloud applications while providing high availability guarantees. Adami et al. [23] designed and developed a network control application for QoS provisioning on top of the Floodlight controller.

Existing works have proved the benefits of SDN to achieve QoS provisioning. Most studies stay at the systematic framework design and test level, while few works focus on queuing and scheduling techniques. In this paper, we implement and verify a SDN-based QoS guaranteed technique, which combines application identification with queue scheduling to achieve QoS guarantee for each cloud user application flows. Our work makes a supplement to existing studies, and provides a basic support for end-to-end QoS routing and multipath routing.

# III. A SDN-BASED QoS GUARANTEED TECHNIQUE

We first briefly introduce the system framework of the proposed SDN-based QoS guaranteed technique. Then, we describe the application identification approach and the queue scheduling mechanisms.



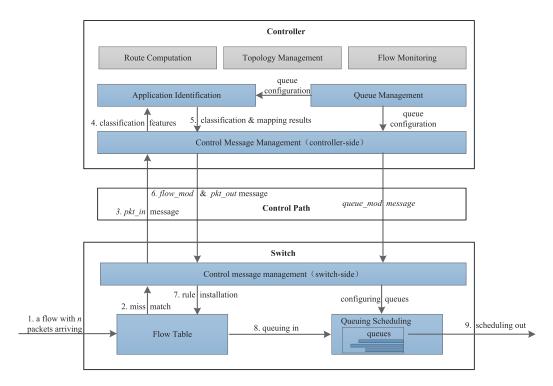


FIGURE 1. System framework of the SDN-based QoS guaranteed technique.

#### A. SYSTEM DESIGN

As depicted in Fig. 1, the system mainly contains three modules: the application identification module, the queue management module and the queue scheduling module. We also redesign the control message management modules for both the switch and the controller.

## 1) CONTROL MESSAGE MANAGEMENT

This module is responsible for sending, receiving and processing the control messages, mainly including the <code>packet\_in</code> message, the <code>packet\_out</code> message, the <code>flow\_mod</code> message and the <code>queue\_mod</code> message. If a packet can match a rule of the flow table, it will be forwarded at a line rate. Otherwise, the switch will generate a <code>pkt\_in</code> message and send it to the controller. After the controller decides how to forward the packet, it will send a pair of control operation messages (<code>flow\_mod</code> and <code>pkt\_out</code>) to the switch: <code>flow\_mod</code> message carries the forwarding rule that will be installed in the switch; <code>pkt\_out</code> message instructs to directly forward the miss-match packet through a specified output interface of the switch. The <code>queue\_mod</code> message is used to configure the queues on the output interface.

# 2) APPLICATION IDENTIFICATION

We implement the application identification technique in this module. it can identify application types according to the application features, and map different application types to different required QoS levels. This will instruct to configure QoS forwarding rules in the flow tables through the control message management module.

#### 3) QUEUE MANAGEMENT

This module is in charge of configuring queues on the output interfaces and maintaining the queue configuration information. It sends queue configuration commands to a switch through a *queue\_mod* message. The switch parses the *queue\_mod* message and configures the queues on a specified output interface.

# 4) QUEUE SCHEDULING

This module queues the packets of different applications into different queues, and then schedules the packets out of the queues with different priorities. Each output interface can configure no more than eight queues with different required QoS levels. In our study, we create three queues for each output interface. The relation between an application and a queue is presented in the *action* filed of a flow table item, marked as enqueue = x : y. It means packets of this application is queued into queue y of output interface x.

In addition to the primary modules, we also use the route computation, the topology management and the flow monitoring functions of the controller. They work together with the application identification module to calculate an output interface with a specific queue number for an application flow. The workflow of the proposed QoS guaranteed technique is described as follows.

Step 1-2: A flow contains many packets of  $\{p_1, p_2, \ldots, p_n\}$  arriving at a switch. If  $p_1$  matches a rule of the flow table, it will be directly forwarded through step 8-9. Otherwise, the switch needs to request the controller for forwarding decision through step 3-4 for the miss-match packet.



Step 3-4: The switch generates a  $pkt\_in$  message for  $p_1$  and send it to the controller. The control message management module captures the header fields of  $p_1$  included in the  $pkt\_in$  message. Then, the header information is sent to the application identification module.

Step 5: According to the header information, the application identification module extracts required features and queries the trained classification algorithm to determine the application types and the required QoS levels. The results are sent to the control message management module.

Step 6: According to the classification and QoS mapping results, as well as the forwarding interface computed by the route computation function, the control message management module generates a pair of control operation messages (flow mod and pkt out) and sends them to the switch.

Step 7: The switch installs the forwarding rule in the switch according to the  $flow\_mod$  message and directly queues  $p_1$  into a queue of a specified output interface according to the  $pkt\_out$  message.

Step 8-9: The switch queues the subsequently arrival packets of the application flow into the queue of the specified output interface. And at the same time, the queue scheduling module schedules the queued packets of the application out of the queue of the interfere.

#### **B. APPLICATION IDENTIFICATION**

In this section, we train the *C4.5 decision tree* to identify application types. Then we define the rules to map each application type to a specific required QoS level.

## 1) APPLICATION TYPES IDENTIFICATION

Features selection is the basis of application type identification. It starts at the flow setup phase. For TCP application flows, the features includes {source port, destination port, MSS (Maxitum Segment Size), window size}. It is well known that the port is closely related to application types. In addition, MSS and window size present great differences among the applications. For UDP application flows, the features includes {source port, destination port}.

According to the selected features, the C4.5 decision tree [24] is trained and implemented in the controller to identify the application types. Assuming the training dataset of S has k kinds of application types, the information entropy of S can be expressed as equation (1).

$$H(S) = -\sum_{i=1}^{k} p_i \log_2 p_i$$
 (1)

Each element in  $\{p_1, p_2, \dots, p_k\}$  represents the probability of an application type appearing in the dataset of S. Entropy is used to address the information uncertainty. The smaller the entropy is, the lower uncertainty the information is. And low information uncertainty means the dataset of S is concentrated in some application types. If we divide S into n subsets according to the attributes of X. The expectation entropy of X

to S (conditional entropy) is expressed as equation (2).

$$H(S|X) = \sum_{i=1}^{n} p(S_i)H(S_i)$$
 (2)

As shown in equation (3), the information gain is the difference between the entropy and the expectation entropy.

$$Gain(S, X) = H(S) - H(S|X)$$
(3)

The *C4.5 decision tree* introduces the information gain ratio (*GainRatio*) based on the information gain. It calculates the *GainRatio* for each attribute and chooses the attribute with the maximum *GainRatio* as the split node. The *GainRatio* is expressed by equation (4).

$$GainRatio(S, X) = \frac{Gain(S, X)}{SplitInfo(S, X)}$$
(4)

Where, SplitInfo(S, X) is the split information of X to S and expressed as equation (5).

$$SplitInfo = -\sum_{i=1}^{n} p(S_i) \log_2 p(S_i)$$
 (5)

**Algorithm 1** Application Types Identification With *C4.5 Decision Tree* 

**Input:** training dataset of S, attribute set of A

- 1:  $Tree \leftarrow \{\}$
- 2: **if** S belongs to the same class or A is  $\Phi$  **then**
- 3: Terminate;
- 4: end if
- 5: **for** all attribute  $X \in A$  **do**
- 6: compute H(S|X), SplitInfo(S, X) and GainRatio(S, X)
- 7: end for
- 8:  $X_{best}$ =Maximum(GainRatio(S, X));
- 9:  $Tree \leftarrow$  create a decision node with  $X_{best}$ ;
- 10:  $S_{subset} \leftarrow \text{split } S \text{ into } n \text{ different subsets based on } X_{best};$
- 11: **for** all  $S_{subset}$  **do**
- 12:  $Tree_n \leftarrow C4.5 (S_{subset}, A X_{best});$
- 13: attach  $Tree_n$  to the corresponding branch of the Tree;
- 14: **end for**
- 15: **return** Tree

Algorithm 1 describes how to use the C4.5 decision tree algorithm to identify application types. Since the algorithm is well known, we omit the details of the algorithm description. In this paper, we adopt the *Moore* dataset [25] to train and verify the algorithm. *Moore* dataset is popularly applies to application identification. The dataset contains 370000 flows composed by 10 application types shown in Table I [37]. We use k-folder  $Cross\ Validation\ [26]$  to test the effectiveness of  $Algorithm\ 1$ . We divide the  $Moore\$ dataset into  $K\$ subsets and consider K-1 subsets as the training data. The selected feature is the port number. Each subset of the  $K\$ subsets will be regarded as the test set, so the training and test process will be conducted  $K\$ times. As a result, we can get  $K\$ classification models. We find that when  $K\$ is set to 10, the average



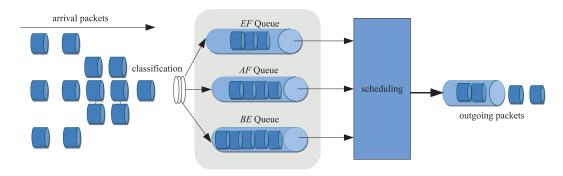


FIGURE 2. Queuing implementation at each output interface.

identification accuracy is 99%. Results show that the training *decision tree* can be used to identify the application types.

For the misclassification application flows, on one hand, we can still provide best effort services as the traditional way. On the other, during the subsequent transmission, more features (such as, *packet size*, *packet number* and *inter-packet gap*, etc.) can be collected to further determine the application types. If the result is different to that judged at the flow setup phase, corresponding QoS forwarding rules will be updated.

TABLE 1. Application types and corresponding required QoS levels.

Application Types	QoS Levels	
VOIP, GAME, SERVICES, CHAT	Expedited Forwarding	
MULTIMEDIA, WEB, INTERACTIVE	Assured Forwarding	
EMAIL, BULK, P2P	Best Effort	

#### 2) APPLICATION QOS-LEVELS MAPPING

Different application types have different QoS requirements, which should be mapped to different QoS levels. Application flows mapped to the same QoS level are queued into the same queue. Through QoS levels mapping, the switch can provide differential services for the application flows. According to the specification of IEEE 802.1Q [27], we classify the QoS into three levels: Expedited Forwarding (EF), Assured Forwarding (AF) and Best Effort (BE) forwarding. Considering the requirements of the applications in delay, jitter and bandwidth, we map the application types to the three required QoS levels. The mapping relations are shown in Table I. Realtime applications (VOIP, GAME, SERVICES and CHAT) are sensitive to delay and jitter. 2) Streaming application (MULTIMEDIA) focuses on unidirectional transmission and interactive applications (WEB and INTERACTIVE) are executed on the basis of the request-response model. Both of them are less sensitive to delay, but require bandwidth guarantee for availability. 4) Compared with the other applications, background applications have little demands on delay and bandwidth, e.g., Email, Bulk and P2P.

# C. QUEUE SCHEDULING

Queue scheduling aims to queue the application flows into the queues with different QoS levels, and schedule the queued packets out of the output interfaces. In this paper, we design the queue scheduling algorithms based on *LLQ* (*Low Latency Queueing*) [28].

# 1) QUEUE IMPLEMENTATION

LLQ brings the ability to specify low latency behaviour for a traffic class. Each queue is equipped with a priority and application flows queued in the queue with higher priority will be scheduled out of the interface first. As shown in Fig. 2, we implement three queues at each output interface, including EF queue, AF queue and BE queue. EF queue (highest priority): packets queued in the EF queue are served with strict QoS guarantee. AF queue (middle priority): packets queued in the AF queue can get a certain degree of minimum bandwidth guarantee. BE queue (lowest priority): packets queued in the BE queue enjoy the best effort service without QoS guarantee.

**Algorithm 2** Queue the Packets Into the Queues of an Output Interface

```
Input: n packets arriving at the output interface of I
 1: for each arrival packet p_i do
      if p_i belongs to EF QoS level then
 2:
         queue p_i to EF Queue;
 3:
       else
 4:
 5:
         if p_i belongs to AF QoS level then
 6:
            queue p_i into AF Queue;
 7:
 8:
            queue p_i into BE Queue;
         end if
 9:
10:
       end if
11: end for
```

**Algorithm II** describes how to queue the packets into the queues of an output interface. Assume that n packets of different applications arrive at a switch, and all the packets are forwarded to the same output interface of the switch. According to the forwarding decisions, the packets belonging to different QoS levels are queued into different queues of the output interface. **Algorithm III** describes how to schedule the queued packets out of an output interface. If the EF queue is not empty, all the packets in the EF queue are scheduled out of the output interface (line  $2\sim4$ ). Otherwise, the scheduler



**Algorithm 3** Schedule the Queued Packets Out of an Output Interface

**Input:** EF, AF, BE1: while TRUE do while isNotEmpty(EF) do 2: schedule out one packet in the EF Queue; 3: 4: end while if isNotEmpty(AF) then 5: schedule out one packet in the AF Queue; 6: continue: 7: else 8: schedule out one packet in the BE Queue; 9: end if 10: 11: end while

checks whether the AF queue is empty or not. If the AF queue is not empty, one packet in the AF queue is scheduled out of the output interface (line  $5\sim6$ ). Then, the scheduler moves out from the current *while* loop and moves into the next *while* loop (line 7). If the AF queue is empty, the scheduler checks the BE queue. If the BE queue is not empty, one packet in the BE queue is scheduled out of the output interface (line 9).

#### IV. ANALYSIS BASED ON OUEUING THEORY

To prove the effectiveness of the queue structure and scheduling algorithms, we conduct a theoretical analysis based on the queuing theory. The metrics of queue length, delay and throughput are used to evaluate the proposed queue scheduling technique.

## A. MODEL CREATION

A switch equips each output interface with three kinds of queues, i.e., EF queue, AF queue and BE queue. These queues provide differentiated services for various application flows. Assuming the arrival interval of application type i follows the exponential distribution with the parameter of  $\lambda_i$  (i=EF, AF, BE). The service time and the polling time also have the exponential distributions with the parameters of  $\mu_i$  (i=EF, AF, BE) and  $\theta$ . According to the LLQ-based scheduling algorithms, packets in the EF queue are always served first. Only when the EF queue is empty, the scheduler polls the AF queue and BE queue in turn. Assuming the total bandwidth is  $\mu_{all}$ , thus, the bandwidth of EF queue is  $\mu_{EF} = \mu_{all}$ . If the minimum guaranteed bandwidth of AF queue is  $\mu_{AF}$ , the available bandwidth of BE queue is  $\mu_{BE} = \mu_{all} - \mu_{AF}$ .

Correspondingly, the service state is divided into three types, including serving the EF queue, serving the AF queue and serving the BE queue. We use  $Q_{EF}(t)$ ,  $Q_{AF}(t)$  and  $Q_{BE}(t)$  to represent the number of packets in the three kinds of queues respectively at time t. As expressed by equation (6), S(t) denotes the state at time t, i.e., the queue the scheduler is serving at time t.

$$S(t) = \begin{cases} 0, & \text{serving the EF queue} \\ 1, & \text{serving the AF queue} \\ 2, & \text{serving the BE queue} \end{cases}$$
 (6)

The state space of  $\{S(t), Q_{EF}(t), Q_{AF}(t), Q_{BE}(t)\}$  is depicted as equation (7). In equation (7),  $m_{EF}$ ,  $m_{AF}$  and  $m_{BE}$  represent the maximum length of each type of queue respectively.

$$\{(f, i, j, k) | 0 \le i \le m_{EF}, 0 \le j \le m_{AF}, \\ 0 \le k \le m_{BE}, f \in \{0, 1, 2\}\}$$
 (7)

The state transition rate matrix of EF queue is shown in equation (8). In equation (8), each element of Q is a  $(m_{EF}+1)(m_{AF}+1)(m_{BE}+1)$  order matrix.  $A_1$ ,  $A_2$  and B are the transition rate matrices for state  $0 \to 1$ , state  $1 \to 2$  and state  $2 \to 1$  respectively.  $A_0$  is the transition rate matrix for state  $0 \to 0$ .

$$Q = \begin{cases} 0 & A_0 & B \\ 0 & A_1 & B \\ 0 & A_2 \end{cases}$$
 (8)

The value of B is shown in equation (9). When the state transition  $0 \to 0$  happens, the scheduler serves the EF queue. The value of  $A_0$  is shown in equation (10).

$$B = diag(\theta \dots \theta)_{(m_{FF}+1)(m_{AF}+1)(m_{RF}+1)}$$

$$\tag{9}$$

$$A_{0} = \begin{bmatrix} C_{0} & E & & & \\ 1 & C_{0} & E & & \\ & \ddots & \ddots & \ddots & \\ & & & D & C_{0} \end{bmatrix}$$
(10)

Where, the values of E, D and  $C_0$  are shown in equation (11), equation (12) and equation (13) respectively.

$$E = diag(\lambda_{EF} \dots \lambda_{EF})_{(m_{AF}+1)(m_{BE}+1)}$$
 (11)

$$D = diag(\mu_{EF} \dots \mu_{EF})_{(m_{AF}+1)(m_{BF}+1)}$$
 (12)

$$C_{0} = \begin{cases} 0 & G & & \\ 1 & F_{0} & G & \\ \vdots & & \ddots & \ddots \\ & & & F_{0} \end{cases}$$
 (13)

Where,  $F_0$  and G are depicted in equation (14), as shown at the top of the next page, and equation (15).

$$G = diag(\lambda_{AF} \dots \lambda_{AF})_{(m_{BE}+1)(m_{BE}+1)}$$
 (15)

When the state transition  $0 \to 1$  happens, the scheduler goes to serve the AF queue from the EF queue. The value of  $A_1$  is shown in equation (16).

$$A_{1} = \begin{cases} C_{1} & E \\ & C_{1} & E \\ & & \ddots & \ddots \\ & & & C_{1} \end{cases}$$

$$m_{EF} \begin{cases} C_{1} & E \\ & & C_{1} & E \\ & & \ddots & \ddots \\ & & & C_{1} \end{cases}$$

$$(16)$$

Where,  $C_1$  is shown in equation (17).

$$C_{1} = \begin{cases} 0 & F_{1} & G \\ H & F_{1} & G \\ & \ddots & \ddots & \ddots \\ & & H & F_{1} \end{cases}$$
 (17)



$$F_{0} = \frac{1}{1} \begin{cases} -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE}) & \lambda_{BE} \\ -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE} + \mu_{EF}) & \lambda_{BE} \\ & \ddots & \ddots & \ddots \\ & & -(\theta + \mu_{EF}) \end{cases}$$

$$F_{1} = \frac{1}{1} \begin{cases} 0 \\ -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE}) & \lambda_{BE} \\ -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE} + \mu_{AF}) & \lambda_{BE} \\ & \ddots & \ddots & \ddots \\ & & -(\theta + \mu_{AF}) \end{cases}$$

$$F_{2} = \frac{1}{1} \begin{cases} -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE}) & \lambda_{BE} \\ \mu_{BE} & -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE} + \mu_{BE}) & \lambda_{BE} \\ & \ddots & \ddots & \ddots \\ & \mu_{BE} & -\theta \end{cases}$$

$$(14)$$

$$F_{2} = \frac{1}{1} \begin{cases} -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE}) & \lambda_{BE} \\ \mu_{BE} & -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE} + \mu_{BE}) & \lambda_{BE} \\ & \ddots & \ddots & \ddots \\ & \mu_{BE} & -\theta \end{cases}$$

$$F_{1} = \begin{cases} 0 \\ 1 \\ \vdots \\ m_{RF} \end{cases} \begin{cases} -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE}) & \lambda_{BE} \\ -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE} + \mu_{AF}) & \lambda_{BE} \\ \vdots & \ddots & \ddots & \ddots \\ -(\theta + \mu_{AF}) \end{cases}$$
(18)

$$F_{2} = \frac{1}{\vdots} \begin{cases} -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE}) & \lambda_{BE} \\ \mu_{BE} & -(\theta + \lambda_{EF} + \lambda_{AF} + \lambda_{BE} + \mu_{BE}) & \lambda_{BE} \\ \vdots & \ddots & \ddots & \ddots \\ \mu_{BE} & -\theta \end{cases}$$
(22)

Where, the value of G is shown in equation (15).  $F_1$  and H are depicted as equation (18), as shown at the top of this page, and equation (19).

$$H = diag(\mu_{AF} \dots \mu_{AF})_{(m_{RF}+1)(m_{RF}+1)}$$
 (19)

Similarly, when the state transition  $0 \rightarrow 1$  happens, the scheduler goes to serve the BE queue from the AF queue. The value of  $A_2$  is shown in equation (20).

$$A_{2} = \begin{cases} C_{2} & E \\ & C_{2} & E \\ & & \ddots & \ddots \\ & & & C_{2} \end{cases}$$
 (20)

Where,  $C_2$  is shown in equation (21). The value of G is shown in equation (15) and the value of  $F_2$  is depicted as equation (22), as shown at the top of this page.

$$C_{2} = \begin{cases} 0 & F_{2} & G \\ 1 & F_{2} & G \\ \vdots & \ddots & \ddots \\ m_{AF} & & F_{2} \end{cases}$$
 (21)

# **B. MODEL SOLVING**

According to the state transition matrices, we try to solve the queue system. Equation (23) shows the steady-state distribution of the queuing system.

$$\Pi \cdot Q = \vec{0} \tag{23}$$

The steady-state probabilities of each state are expressed by equation (24).

$$\Pi = \{ \pi_{f,i,j,k} | 0 \le i \le m_{EF}, 0 \le j \le m_{AF}, \\ 0 \le k \le m_{BE}, f \in \{0, 1, 2\} \}$$
 (24)

Through calculating the steady-state probabilities, we can get the performance metrics of the queuing system, including 1) the average queue length denotes the number of the packets waiting to be served in the queue; 2) the throughput refers to the packet forwarding rate of the queue without packet loss; 3) the average waiting time starts from the packet being queued into the queue to being scheduled out of the queue. Under the steady state, the average queue length of the EF queue is shown in equation (25).

$$L_{EF} = \sum_{f=0}^{2} \sum_{i=0}^{m_{EF}} \sum_{j=0}^{m_{AF}} \sum_{k=0}^{m_{BE}} i \cdot \pi_{f,i,j,k}$$
 (25)

The throughput of the EF queue and its average waiting time are shown in equation (26) and equation (27) respectively.

$$\tau_{EF} = \sum_{i=0}^{m_{EF}} \sum_{j=0}^{m_{AF}} \sum_{k=0}^{m_{BE}} \mu_{EF} \cdot \pi_{0,i,j,k}$$
 (26)

$$W_{EF} = \frac{L_{EF}}{\tau_{EE}} \tag{27}$$

similarly, the performance values of the AF queue and BEqueue are shown in equation (28)  $\sim$  equation (33).

$$L_{AF} = \sum_{f=0}^{2} \sum_{i=0}^{m_{EF}} \sum_{i=0}^{m_{AF}} \sum_{k=0}^{m_{BE}} j \cdot \pi_{f,i,j,k}$$
 (28)

$$\tau_{AF} = \sum_{i=0}^{m_{EF}} \sum_{i=0}^{m_{AF}} \sum_{k=0}^{m_{BE}} \mu_{AF} \cdot \pi_{1,i,j,k}$$
 (29)

$$W_{AF} = \frac{L_{AF}}{\tau_{AF}} \tag{30}$$

$$L_{BE} = \sum_{f=0}^{2} \sum_{i=0}^{m_{EF}} \sum_{j=0}^{m_{AF}} \sum_{k=0}^{m_{BE}} k \cdot \pi_{f,i,j,k}$$
 (31)

$$\tau_{BE} = \sum_{i=0}^{m_{EF}} \sum_{j=0}^{m_{AF}} \sum_{k=0}^{m_{BE}} \mu_{BE} \cdot \pi_{2,i,j,k}$$
 (32)

$$W_{BE} = \frac{L_{BE}}{\tau_{BE}} \tag{33}$$

# C. NUMERICAL ANALYSIS

This section conducts a numerical analysis on the performance metrics of the queue system. Assuming the total



bandwidth  $\mu_{all}=0.1~Gbps$ , the average packet size p=1000~Bytes, and the parameters of  $\lambda_{EF}=\lambda_{AF}=\lambda_{BE}=0.0333~Gbps$ ,  $m_{EF}=m_{AF}=m_{BE}=5$  and  $\theta=0.01~s$ . Then, we can get  $\mu_{EF}=\mu_{all}=0.1~Gbps$ . At the initial phase, the bandwidth of AF queue is  $\mu_{AF}=0.04~Gbps$ , and the bandwidth of BE queue is  $\mu_{BE}=\mu_{all}-\mu_{AF}=0.06~Gbps$ .

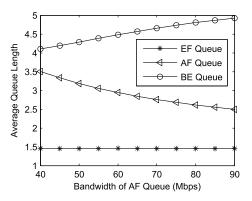


FIGURE 3. Average queue length under different bandwidth of AF queue.

# 1) AVERAGE QUEUE LENGTH

As shown in Fig. 3, with the increase of the bandwidth assigned to AF queue, the average length of AF queue decreases gradually. Correspondingly, the bandwidth assigned to the BE queue decreases gradually, because only the remaining bandwidth is available for the BE queue. Less assigned bandwidth means low packet forwarding efficiency, causing many queued packets waiting for being scheduled out of the BE queue. As a result, the average length of BE queue increases gradually. Since the EF queue has the highest priority to use the bandwidth, the average length of EF queue is always kept in a stable state.

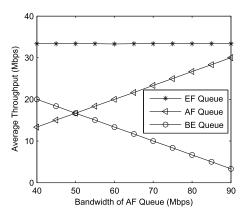


FIGURE 4. Average throughput under different bandwidth of AF queue.

## 2) AVERAGE THROUGHPUT

As shown in Fig. 4, the average throughput of AF queue is lower than that of BE queue at the beginning. With the increase of the bandwidth assigned to AF queue, the average throughput of AF queue presents a growth, and the average throughput of BE queue decreases gradually. The EF queue

has the highest priority of bandwidth utilization, so it keeps the high throughput and is not affected by the other two queues.

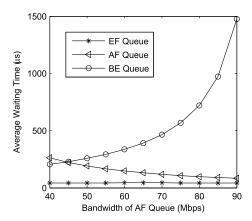


FIGURE 5. Average waiting time under different bandwidth of AF queue.

### 3) AVERAGE WAITING TIME

As shown in Fig. 5, the average queue waiting time of AF queue presents a decrease with the increase of the bandwidth assigned to AF queue. Conversely, the average queue waiting time of BE queue presents a obvious growth when more bandwidth can be utilized. Packets queued in the EF queue are scheduled out of the interface with the highest priority, so the average queue waiting time of EF queue keeps at a low level all the time.

In summary, the *LLQ*-based scheduling algorithms can provide strict QoS guarantee for the application flows queued in the *EF* queue. In addition to this, the algorithm can supply a certain degree of minimum bandwidth guarantee for the application flows queued in the *AF* queue. As a sacrifice, performance cannot be guaranteed for the packets queued in the *BE* are generic data flows, which are not sensitive to delay, jitter and bandwidth, etc. Theoretical analysis results prove the effectiveness of the *LLQ*-based scheduling algorithms.

## **V. EXPERIMENTAL EVALUATION**

In this section, we first describe the experimental environment. Then, we evaluate the effectiveness of the proposed SDN-based QoS technique.

#### A. EXPERIMENT DESCRIPTION

Fig. 6 shows our experiments setup. *Open vSwitch*(OVS) [29] is an open source *OpenFlow* virtual switch. *Floodlight* [13] is an open source SDN controller. We run *OVS* and *Floodlight* on two commodity PCs respectively. Table I shows the configurations of the experimental devices. *Host*<sub>1</sub> and *Host*<sub>2</sub> connect to OVS with 100Mbps interfaces. We run *pktgen* [30] on *Host*<sub>1</sub> to generate traffic at the rates of 5Mbps - 100Mbps with the Ethernet frame size of 1000 Bytes. We run *tcpdump* [31] to listen on the interfaces that are connected to the hosts and the controller respectively.



TABLE 2. Configurations of the experimental devices.

Device Name	CPU	Cores	RAM	NIC
$Host_1$	3.3GHZ	4	4GB	$1 \times 100$ Mbps
$Host_2$	3.3GHZ	4	4GB	$1 \times 100$ Mbps
Open vSwitch	3.3GHZ	4	4GB	$3 \times 100$ Mbps
Floodlight	3.3GHZ	4	4GB	$1 \times 100 \text{Mbps}$

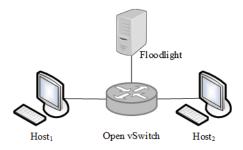


FIGURE 6. Topography of the experimental platform.

In this experiment,  $Host_1$  sends three kinds of flows to  $Host_2$ , including the voice flow (delay-sensitive), the video flow (bandwidth-hungry) and the generic data flow. These three application flows are sent out in cross sequences. We implement the C4.5-based application identification technique in the Floodlight and the LLQ-based queue scheduling technique in the  $Open\ vSwitch$ . The three types of flows will be mapped to the EF, AF and BE QoS levels and correspondingly, be queued into the EF queue, AF queue and BE queue respectively. By default, each output interface is configured with a FIFO queue. We conduct a comparison study between the default queue technique and the LLQ-based queue technique implemented in this study.

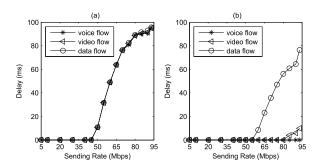


FIGURE 7. Delay variations across different sending rates: (a) The default FIFO mechanism. (b) The proposed LLQ-based mechanism.

#### **B. EFFECTIVENESS EVALUATION**

Fig. 7(a) shows the delay variations of the flows using the default *FIFO* queue. With the increase of the sending rate, the three types of flows compete for the bandwidth intensively. When the total sending rate reaches to the maximum bandwidth of the switch interface, the delays of the three types of flows increase quickly. Since the default *FIFO* queue mechanism does not distinguish the differences of the flows, the delays of the three types of flows present similar patterns.

Neither the voice flow nor the video flow is guaranteed with a tolerant delay. The *LLQ*-based queue mechanism treats the flows differently. As shown in Fig. 7(b), the voice flow is guaranteed with a small delay. It is not affected by the sending rate, because it is mapped to the *EF* QoS level, which has the highest priority to use the bandwidth. The video flow is mapped to the *AF* QoS level and guaranteed with the minimum bandwidth of 30*Mbps*, so the delay of this flow is less affected. However, when the sending rate exceeds 80*Mbps*, the delay of the video flow starts to be influenced slightly. The generic data flow is not guaranteed and the delay presents a similar pattern with the default *FIFO* queue mechanism.

We then conduct further experimental analysis about the proposed technique under different bandwidth conditions. 1) No-congestion: the total bandwidth of the interface is sufficient for the requirements of the three flows. 2) Congestion: the bandwidth requirements of the three flows approach to the total bandwidth limitation of the interface, i.e. 100Mbps. Metrics of delay, jitter and throughput are used to evaluate the performance of the three types of application flows.

# 1) DELAY

#### a: NO-CONGESTION

As shown in Fig. 8(a)  $\sim$  Fig. 8(c), when the interface is not congested, both the default and proposed mechanisms can meet the QoS requirements of the application flows. Using the default FIFO mechanism, the average delays of the three types of flows are  $6.31\mu s$ ,  $6.18\mu s$  and  $6.24\mu s$  respectively. While using the *LLQ*-based mechanism, the average delays of the three types of flows are 4.49  $\mu s$ , 4.43 $\mu s$  and  $4.46\mu s$  respectively. The delays of the three types of flows are reduced by 28.8%, 28.3% and 28.5% on average. For the default mechanism, the sending rate is small, the flows do not need to compete for the bandwidth. Packets queued into the queue can be scheduled out quickly without any queue waiting time. So the default mechanism can also meet the QoS requirements of these application flows. For the LLQ-based mechanism, the bandwidth assigned to each queue is larger than the sending rate of the application flows, so the bandwidth is enough to meet the QoS requirements. However, the proposed mechanism creates three queues for each output interface. Multi-queues mechanism reduces the complexity of packet processing, so the LLQ-based (multiqueues) mechanism performs a little better than the default FIFO (single-queue) mechanism.

# b: CONGESTION

As shown in Fig. 9(a)  $\sim$  Fig. 9(c), the proposed *LLQ*-based mechanism obviously outperforms the default *FIFO* mechanism when the interface is congested. Using the default mechanism, more and more packets are congested in the queue over the time. As a result, the average delays of the three types of flows present fast growth. While using the proposed mechanism, there are there types of queues for each interface. Each queue is marked with a QoS level, and serves the application flows that are mapped to this QoS level. The voice

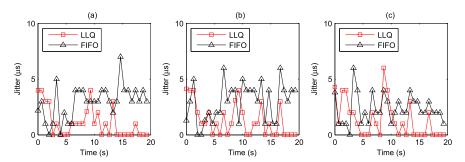


FIGURE 8. No-congestion: Delay variations of different application flows: (a) voice flow; (b) video flow; (c) generic data flow.

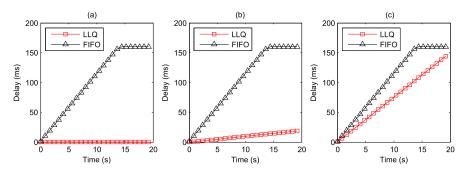


FIGURE 9. Congestion: Delay variations of different application flows: (a) voice flow; (b) video flow; (c) generic data flow.

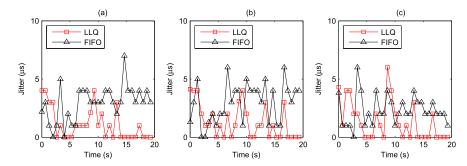


FIGURE 10. No-congestion: Jitter variations of different application flows: (a) voice flow; (b) video flow; (c) generic data flow.

flow has the highest priority to use the bandwidth, so the delay is small and keeps stable. The video flow is provided with a certain degree of minimum bandwidth guarantee, so the delay increases slowly. The generic data flow is served in the best effort way, so the delay increases quickly. However, its growth rate is smaller than that of the default mechanism. Under the congestion condition, the average delays of the three types of flows are reduced by 99.996%, 90.66% and 27.84% on average respectively. In view of delay, the voice flow is strictly guaranteed and the video flow can also be guaranteed to a great extent.

# 2) JITTER

# a: NO-CONGESTION

As shown in Fig.  $10(a) \sim \text{Fig. } 10(c)$ , when the interface is not congested, the jitters of the three types of flows are small. However, the proposed mechanism performs a little better

than the default mechanism. For the default mechanism, the average jitters of the three types of flows are  $3.02~\mu s$ ,  $3.93~\mu s$  and  $2.20~\mu s$  respectively. While for the proposed mechanism, the average jitters of the three types of flows are  $1.02~\mu s$ ,  $1.12~\mu s$  and  $1.09~\mu s$  respectively. The jitters can be reduced by 66.17%, 61.61% and 50.41% on average respectively. The reason is that the proposed mechanism adopts multiqueues to serve different application flows, which can avoid the inference when all the application flows are queued into a single queue.

# b: CONGESTION

As shown in Fig.  $11(a) \sim \text{Fig. } 11(c)$ , when the interface is congested, the proposed mechanism presents obvious advantages in view of jitter. Compared with the default mechanism, the jitters of the three types of flows can be reduced by 99.85%, 91.18% and 34.13% on average respectively.



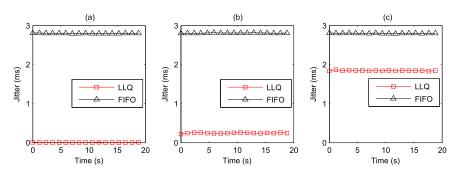


FIGURE 11. Congestion: Jitter variations of different application flows: (a) voice flow; (b) video flow; (c) generic data flow.

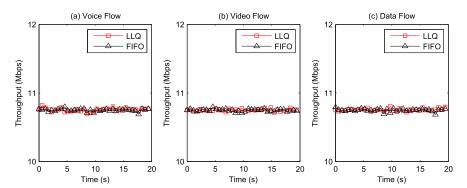


FIGURE 12. No-congestion: Throughput variations of different application flows: (a) voice flow; (b) video flow; (c) generic data flow.

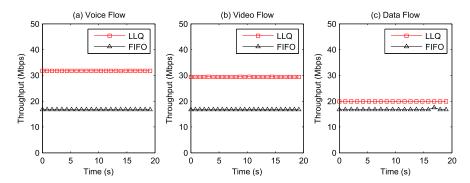


FIGURE 13. Congestion: Throughput of different application flows: (a) voice flow; (b) video flow; (c) generic data flow.

#### 3) THROUGHPUT

# a: NO-CONGESTION

As shown in Fig.  $12(a) \sim \text{Fig. } 12(c)$ , when the interface is not congested, the throughputs of the two mechanisms present similar patterns. This is because the interface can provide sufficient bandwidth for each kind of application flow. Packets queued into the queues can be scheduled out quickly by both of the default mechanism and the proposed mechanism.

## b: CONGESTION

As shown in Fig. 13(a)  $\sim$  Fig. 13(c), when the interface is congested, the proposed mechanism can achieve greater

throughput than the default mechanism. Compared with the default mechanism, the throughputs of the three types of flows can be increased by 90.17%, 76.06% and 18.5% on average respectively.

### **VI. CONCLUSIONS AND FUTURE REMARKS**

In this paper, we propose and evaluate a SDN-based QoS technique for cloud applications. 1) We first design the architecture of the SDN-based QoS technique, which combines application identification with queue scheduling. 2) Then, we implement an application identification method in the SDN controller. It can identify application types and map each type of application to a required QoS level. 3) Thirdly, we



implement a queue scheduling method in the switch. We also prove the effectiveness of the queue scheduling technique by a theoretical analysis. 4) At last, we evaluate the proposed SDN-based QoS technique through an in-depth experimental analysis.

In the future, more queues will be created at each output interface, and the proposed SDN-based QoS technique will be used to provide end-to-end QoS routing and multipath routing. In addition, more features will be extracted to classify the application flows at finer granularity. We will further evaluate the efficacy of proposed technique through commodity SDN switches with Gigabit Ethernet, and try to extend the use of the proposed QoS technique in the area of mobile computing [35], [36].

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