

SDN 交換機核心技術-流量分類以及應用辨識技術

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Today's routers



Routing, management, mobility management, access control, VPNs, ...

Feature Feature

OS

Custom Hardware

Million of 6,000 RFCs lines of source code

Billions of Bloated Power Hungry gates

- Vertically integrated, complex, closed, proprietary
- Networking industry with "mainframe" mindset

What SDN really is





App pp pp pp pp pp pp pp

Open Interface

Control Plane Or

Control Plane

Or Control Plane

Open Interface

Merchant Switching Chips

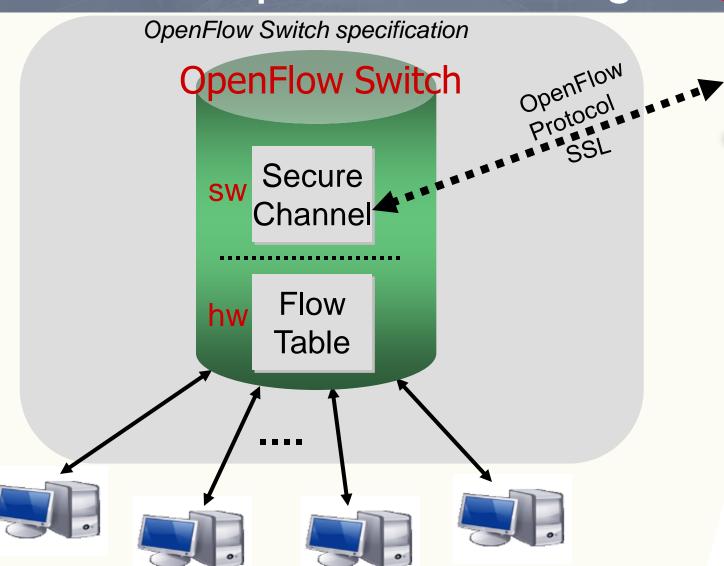
Vertically integrated Closed, proprietary Slow innovation



Horizontal
Open interfaces
Rapid innovation

OpenFlow Switching

Controller

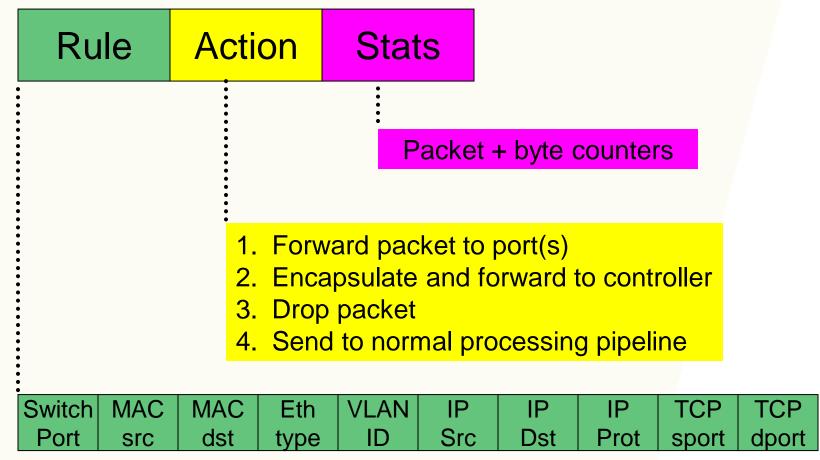


The Stanford Clean Slate Program

http://cleanslate.stanford.edu

Flow Table Entry

"Type 0" OpenFlow Switch



+ mask

The Stanford Clean Slate Program

http://cleanslate.stanford.edu

Applications for SDN Networks

- QoS management
- Security management
- Network management
- BYOD management

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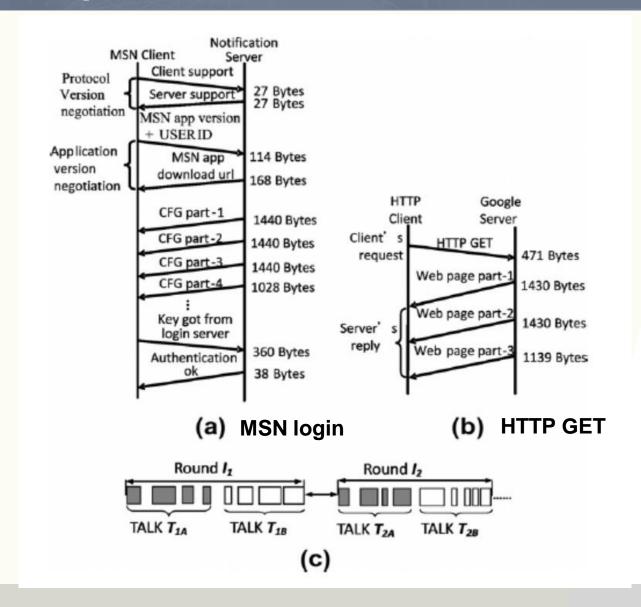
Key issues in SDN switches

- Fast and Accurate Flow Classification
 - Fast SDN Switching Fabric with TCAM based flow table (TACM lookup)
 - Accurate Flow table with accurate flow signatures
- But only Layer-4 fields are designed in the standard flow table of OpenFlow SDN switches
 - Good for port-based applications
 - FTP (TCP/20-21), DNS (UDP/53), Telnet (TCP/23), Http (TCP/80)
 - Not good enough for port-changing applications
 - P2p sharing, On-line Games, Skype, BT, ...

Key issues in SDN switches

- We can either classify the traffic with Deep Packet Inspection (DPI) technology
 - High Accuracy
 - But need to maintain the signatures dynamically
 - Unable to classify the encrypted flow
 - Not practical for cloud based platform
- Or we can use Machine Learning based approach
 - Use "statistical attributes" of flows to achieve high accuracy identification
 - Without inspecting the packet payload
 - Practical for cloud based platform as only "statistical attributes" are forwarded to the cloud.

Examples of Statistical Attributes

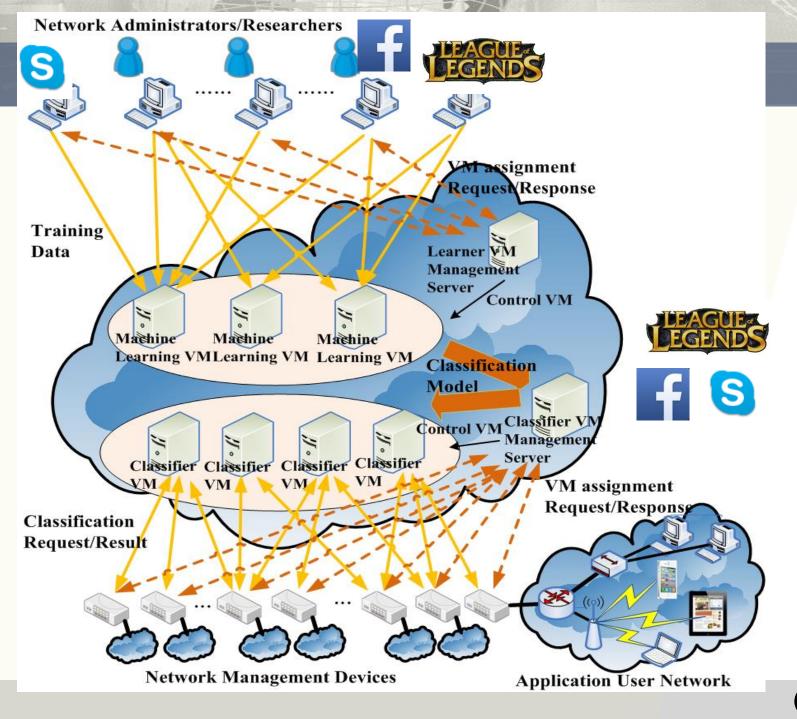


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A Cloud based Applications Classification Service Platform with Machine Learning Algorithms for SDN Networks

基於機器學習演算法之雲端應用辨識服務平台 與 SDN 網路應用

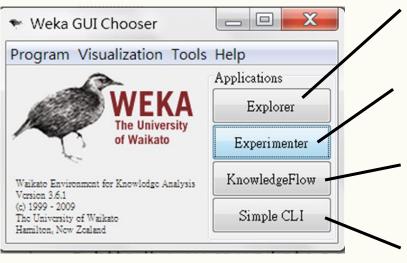


Architecture of Traffic Classification Service on Cloud

- Machine learning virtual machine (LVM)
 - Running training service
- Classifier virtual machine (CVM)
 - Running classification service
- LVM management server
 - LVM assignment and monitoring
- CVM management server
 - CVM assignment and monitoring
- Client (PCs running the traffic classification client)
 - Computing the statistical attributes of each flow
 - Retrieving the training/classification service.

The used machine learning tools-Weka

- Weka -Waikato Environment for Knowledge Analysis (running on single PC/NB)
 - A JAVA-based open source software suite that collects machine learning algorithms for data mining tasks
 - Four kinds of interface



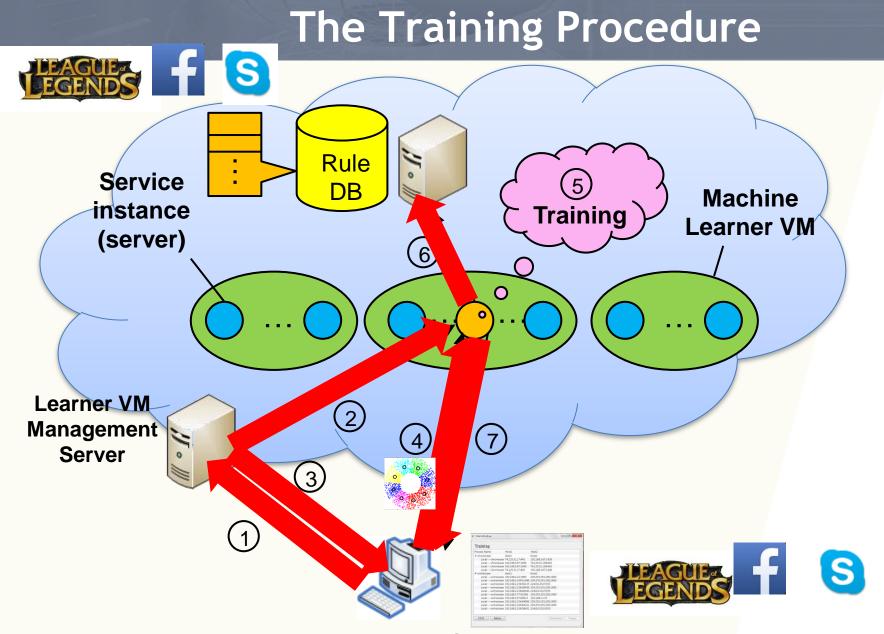
Explore: To see data distribution and apply ML algorithm directly.

Experimenter: Batch modes for running training-testing evaluation

Knowledge Flow: To combine the different ML algorithms component as graph-based processing flow

CLI: Command Line Interface

http://www.cs.waikato.ac.nz/ml/weka/



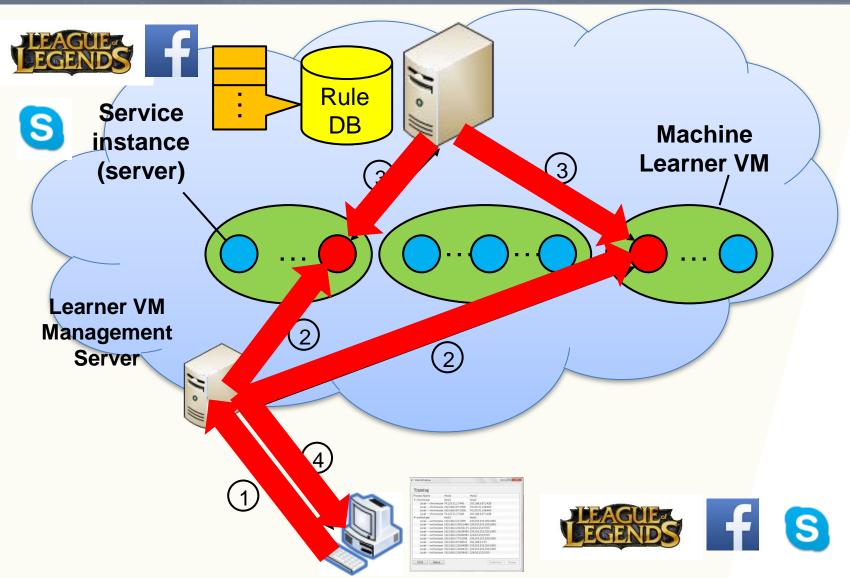
The Training Service

- Requesting the training service
 - 1. Client → LVM management server
- LVM assignment
 - 2. Checking the loading and status of each service instance on the LVM
 - IP address & port number of LVM
 (LVM Management Server → Client)

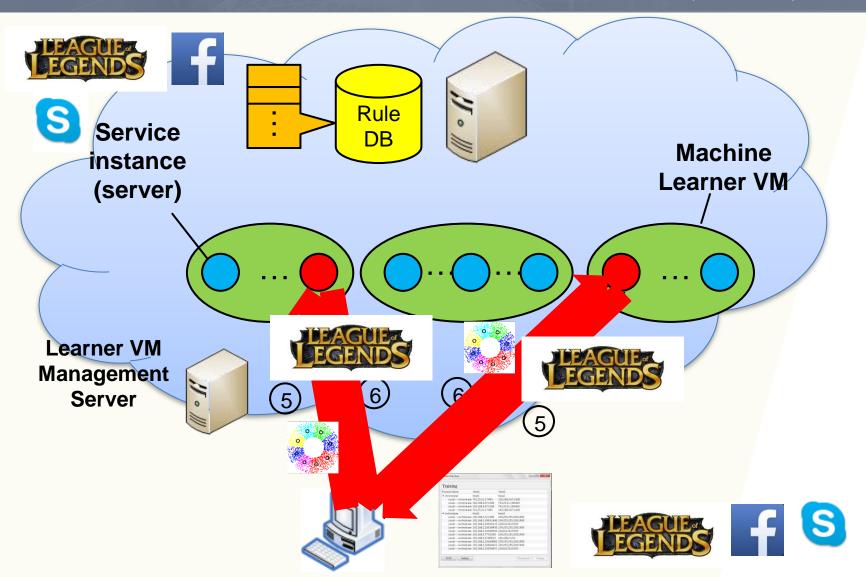
The Training Service

- Training
 - 4. Training Data (attributes of each flow to train):
 Client → LVM
 - 5.Training phase
- Training Results
 - 6. Storing the rules (LVM → Rule DB)
 - 7. Training result (e.g. number of rules.)
 - Attributes: Two way communications, packet size, packet number, packet time interval, download rate, upload rate, ..., etc

Traffic Classification Procedure



Traffic Classification Procedure (cont.)



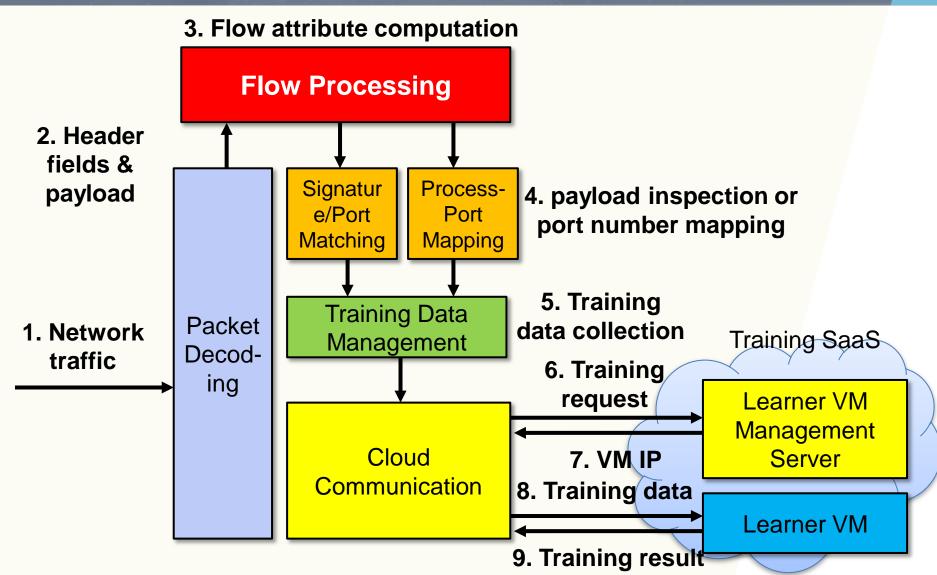
The Traffic Classification Procedure

- Requesting the classification service
 - 1. Client → CVM management server
- CVM assignment
 - 2. Checking the loading and status of each service instance on the CVM
 - 3. Loading the classification rules from DB.
 (Rule DB ←→ CVM)
 - 4. CVMs' IP address & port number
 (CVM Management Server → Client)

The Traffic Classification Procedure

- Classification
 - 5. Sending the attributes of flows to classify
 - Client → CVMs
 - 6. Receiving the classification result (application id)
 - CVMs → Client

Client Program for Training



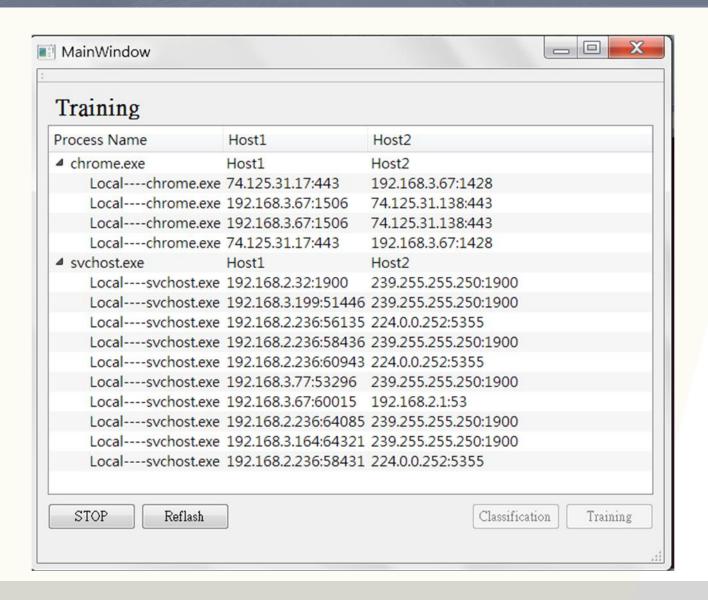
Implementation: Client Program Modules

- Packet decoding module
 - decoding the packet header fields
- Flow processing module
 - computing flow attributes
 - managing the flow table for storing the flow attributes
- Signature/port matching and process-port mapping modules (only for training)
 - To identify the original background truth (Training label).
 - User can choose the signature/port matching or processport mapping as the module to find the original application class for each flow.

Implementation: Client Program Modules

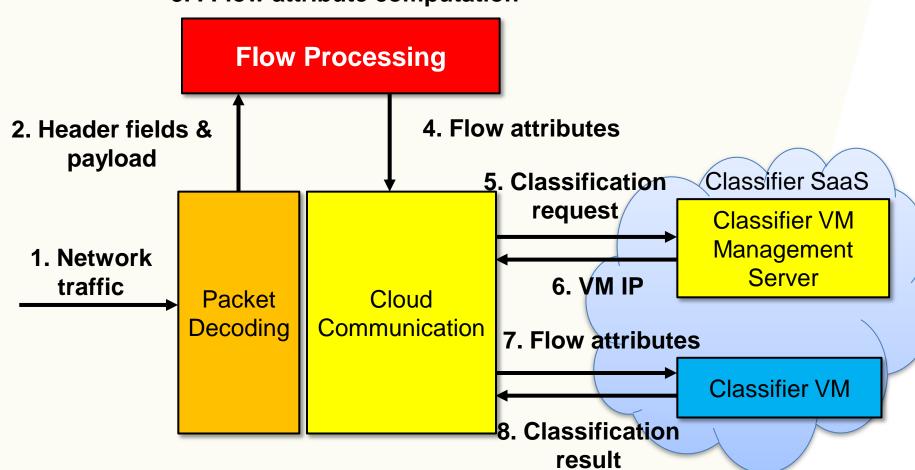
- Training data management module (only for training)
 - collecting the attributes of flows to form the training data set.
- The cloud communication module
 - Client ← → LVM/CVM Management server
 - Client ←→ LVM/CVM

The client program (Training)



Client Program for Classification

3. . Flow attribute computation



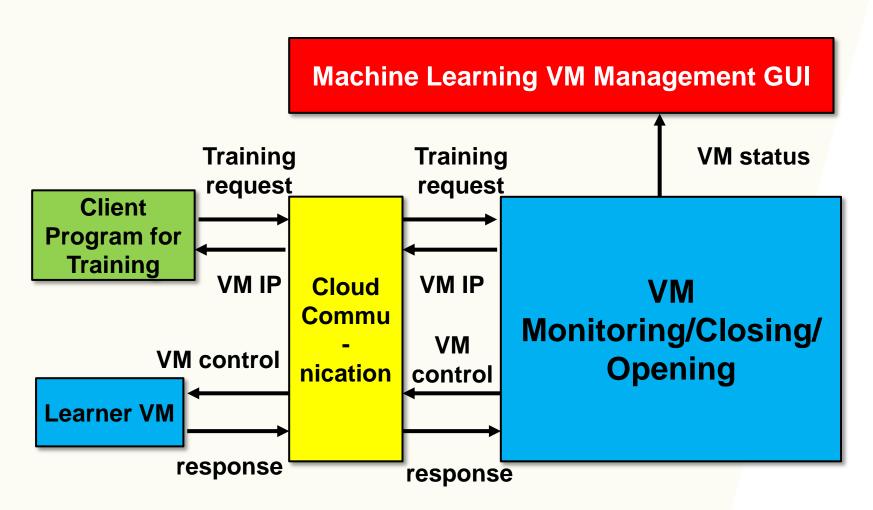
LVM and CVM management server (Cloud)

- Machine learning/classifier VM management server graphic user interface (GUI)
- VM monitoring/closing/opening module
- The cloud communication module in the LVM and CVM management server

Program modules of LVM and CVM (Cloud)

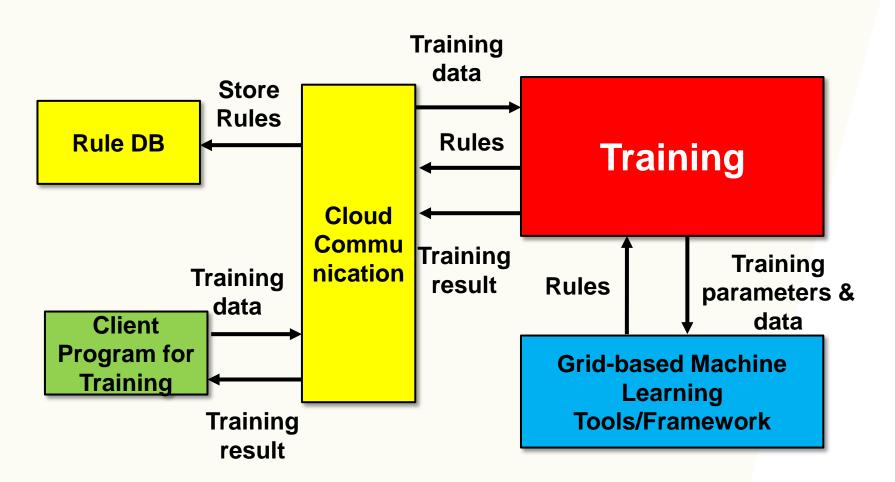
- The training/classification modules
- The cloud communication module in the LVM or CVM
- Grid-based machine learning tools/framework
 - The implementation of the machine learning algorithms on grid environment.

Learning VM Management Server

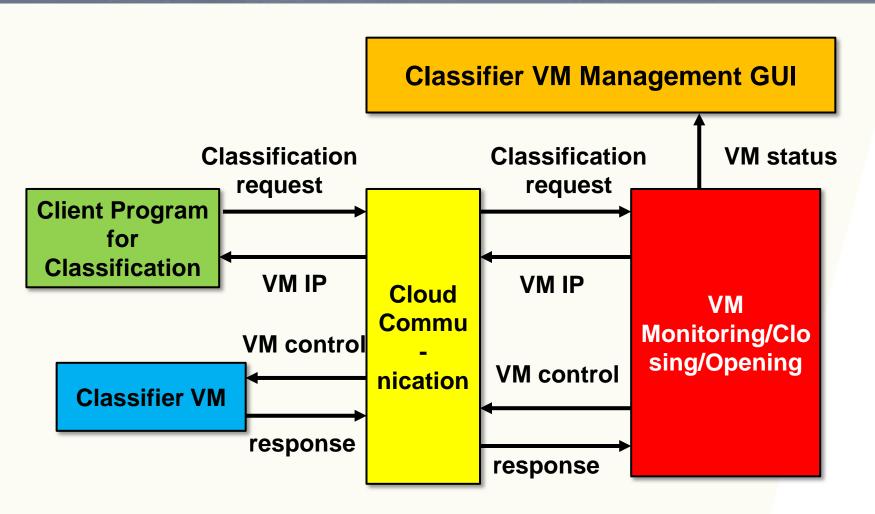


Learning VM Management Server

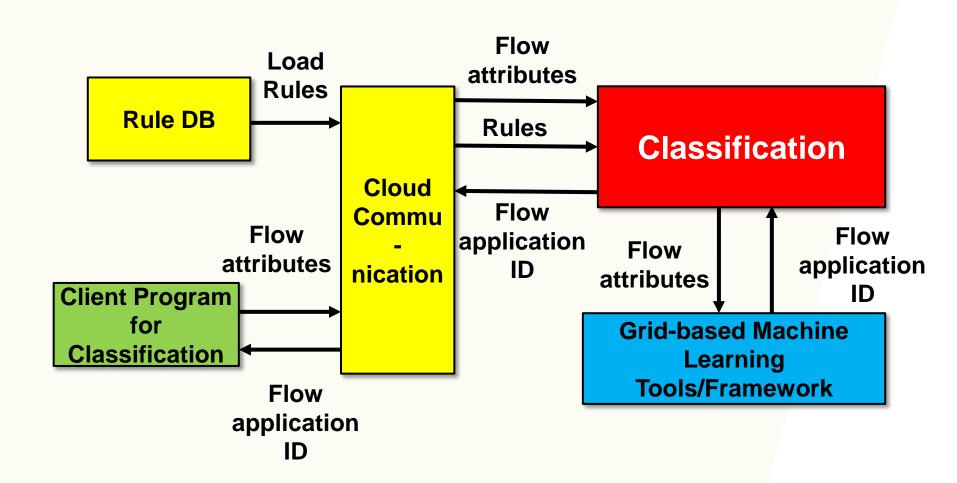
Learning VM



Classifier VM Management Server



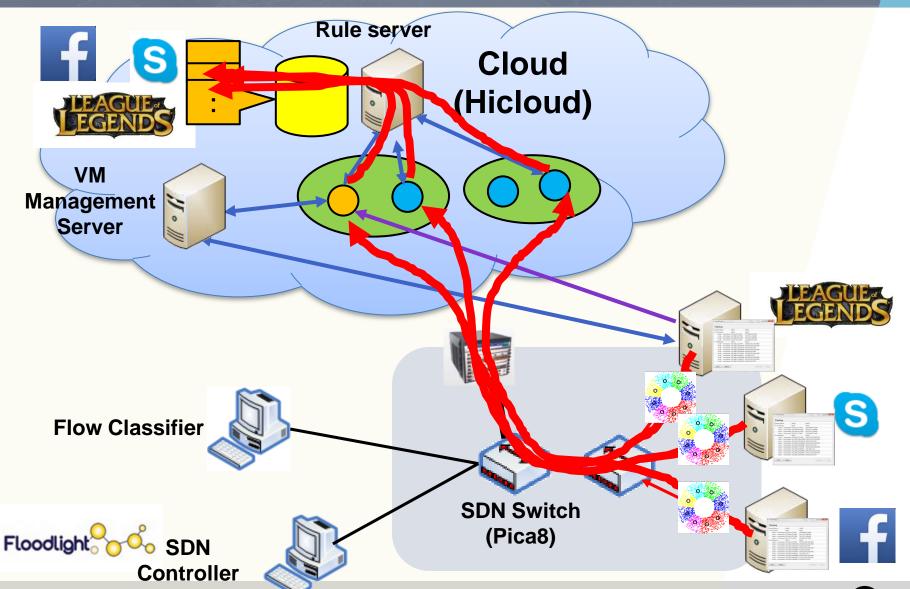
Classifier VM



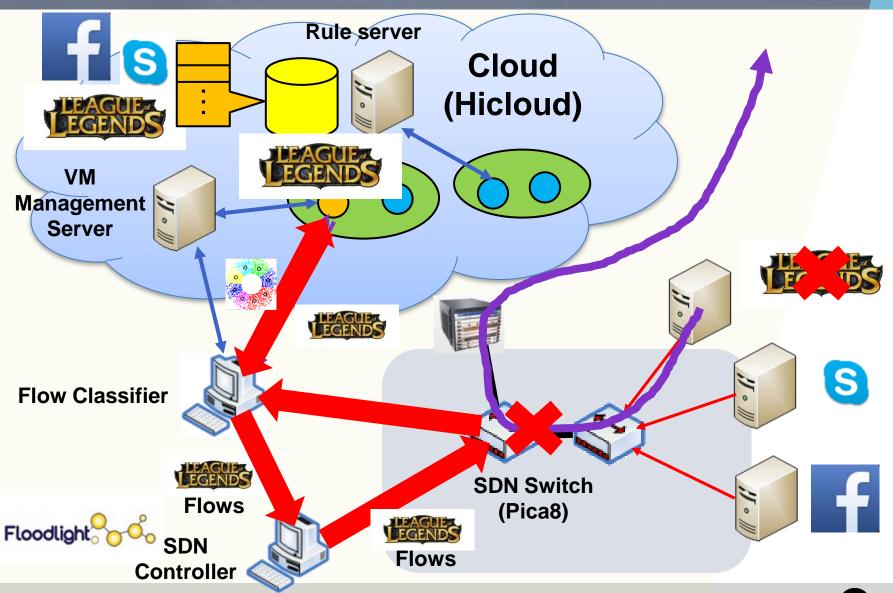
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Cloud + MLAC + SDN (Training)



Cloud + MLAC + SDN (Classification/Management)



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Demo of Applications Training (Video)

Demo of Applications Classification and Management with SDN Switch (Video)

Conclusions and future works

- The key issues of SDN networks include fast and accurate flow classification schemes
- The current SDN switches only support layer-4 flow table.
- A new service framework of applications classification for SDN networks
 - Cloud based for scalability, reliability
 - Machine Learning based for light communications
 - Global collected application attributes (statistical signatures) database.
 - Early stage application management
- Currently, this platform is still only a prototype, more training clients need to be installed (the client training software will be released later) to learn the worldwide applications as complete as possible.
- The accurate of the training and classification of this platform (currently around 80-98% of accuracy) needs also to be enhanced by tuning the machine learning model.

未來展望

- 建立全球最完整之網路應用軟體統計特徵資料庫
 - 開放式, 眾志成城 (more PCs to run applications)
- * 與雲端平台業者組織合作
 - 佈建此資料庫 (for millions of applications)
 - 可動態擴增之雲端應用軟體訓練虛擬機器 (scalable)
 - 可動態擴增之雲端應用軟體辨識虛擬機器 (scalable)
- ❖ 提供全球網路應用軟體即時辨識服務(SaaS)
 - 搭配 SDN 網路提供加值服務
 - QoS, Security, Network Management, ..., etc
- * 臺灣學術網路或校園網路示範建置



Thank you for your attentions!