

Network Working Group
Internet-Draft
Intended status: Informational
Expires: September 14, 2017

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March 13, 2017

Usecases for Network Artificial Intelligence (NAI)
draft-zheng-opsawg-network-ai-usecases-00

Abstract

This document discusses the scope of Network Artificial Intelligence (NAI), and the possible use cases that are able to demonstrate the advantage of applying NAI.

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Table of Contents

1. Introduction	2
2. NAI Architecture	3
3. NAI Use Cases	3
3.1. Traffic Predication and Re-Optimization/Adjustment	3
3.2. Route Monitoring and Analytics	4
3.3. Multilayer Fault Detection In NFV Framework	5
3.4. Data Center Network Use Cases	7
3.4.1. Service Function Chaining	7
4. Contributors	8
5. Security Considerations	8
6. IANA Considerations	8
7. Acknowledgement	8
8. References	8
8.1. Normative References	9
8.2. Informative References	9
Authors' Addresses	9

1. Introduction

Current networks have become much more dynamic and complex, and pose new challenges for network management and optimization. For example, network management/optimization should be automated to avoid human intervention (and thus to minimize the operational expense). Artificial Intelligence (AI) and Machine Learning (ML) is a promising approach to realize such automation, and can even do better than human beings. Furthermore, the population of Software-Defined Networks (SDN) paradigm makes the application of Artificial Intelligence in networks possible, since the SDN controller has the complete knowledge of the network status and can control behavior of network nodes to implement AI decisions.

AI and ML technologies can learn from historical data, and make predictions or decisions, rather than following strictly static program instructions. They can dynamically adapt to a changing situation and enhance their own intelligence with by learning from new data. It can learn and complete complicated tasks. It also has potential in the network technology area especially with SDN and Network Function Virtualization (NFV).

This document presents the concept of Network Artificial Intelligence. It first discusses the scope of Network Artificial Intelligence (NAI). And then Some use cases are discussed to demonstrate the advantage of applying NAI.

2. NAI Architecture

The definition of the architecture of NAI could be refer to [I-D.li-rtgwg-network-ai-arch]. In the architecture of NAI, central controller is the core part of Network Artificial Intelligence which can be called as 'Network Brain'. The Network Telemetry and Analytics (NTA) engines can be introduced accompanying with the central controller. The Network Telemetry and Analytics (NTA) engine includes data collector, analytics framework, data persistence, and NAI applications.

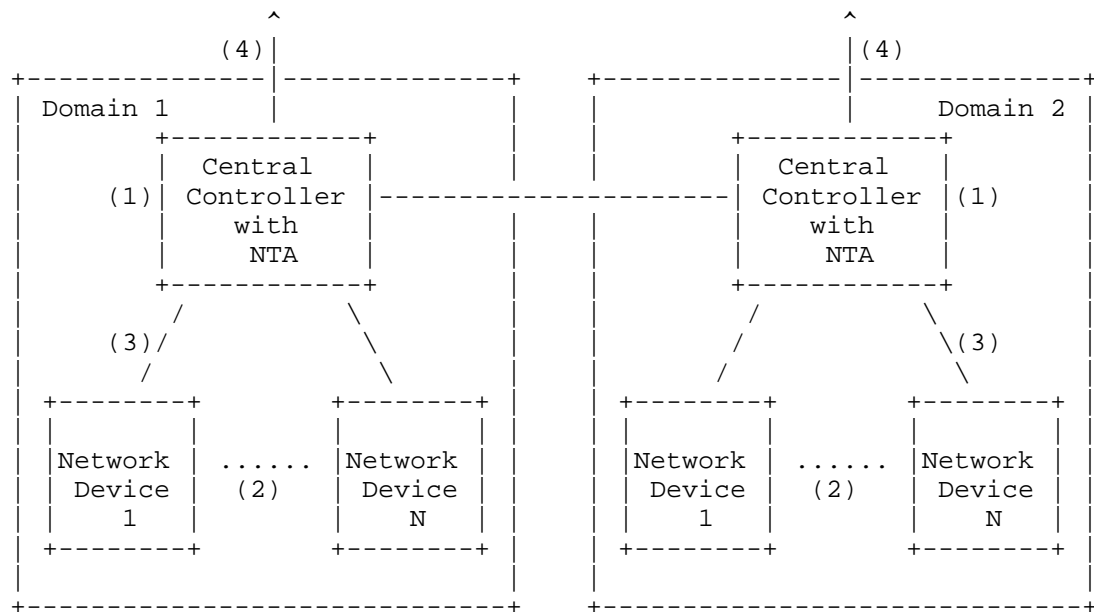


Figure 1: An Architecture of Network Artificial Intelligence (NAI)

3. NAI Use Cases

3.1. Traffic Predication and Re-Optimization/Adjustment

This subsection introduces the Path Computation Element (PCE) [RFC4655] use cases in wide area networks (WAN). In PCE scenario, network data collection is realized through the control plane protocols such as PCE protocol (PCEP) and BGP-LS [RFC7752] protocol and data are passed to the PCE application. PCEP receives the state of Label Switched Path (LSP) from the network, and BGP-LS receives the topology information from the network. If network telemetry is used, traffic information can be received from the network as well directly at the NTA engine using protocols such as gRPC.

PCE application (APP) only maintains the latest information. To enable NAI, history of all LSP and topology changes is stored in external data repository. Further traffic monitoring data could also be collected and stored, if network telemetry is used. There are two usecases in the application scenarios: (1) reroute/re-optimize using the historical trend and predications from AI; (2) traffic congestion avoidance and AI-enabled auto-bandwidth adjustment.

For the usecase (1), the analytics component in NTA (Network Telemetry and Analytics), can use stored data to build models to predict impact of network events and state of the LSPs. For example, it can use historical trends to guide path computation to include/exclude specific links. Finding correlations between data, finding anomalies and data visualization are also possible.

The analytics component in NTA can also use stored data to detect and predict network events and request PCE to take necessary actions. For example, it can use network bandwidth utilization historical trends to request for re-optimizations.

For the usecase (2), with network telemetry, the NTA can collect per-link and per-LSP traffic flow using gRPC from network. Such network telemetry data includes statistics for tunnels, links, bandwidth reservations, actual usage, delay, jitter, packet loss, etc. Meanwhile, it also collects data regarding network events and its impact on traffic flows. The analytics component can use telemetry data to build traffic models to predict traffic congestion when new years or sporting events are coming. According to the congestion prediction, the PCE app could reroute traffic to avoid congested links. Besides the case, NTA can also perform predication and make necessary changes to network. In particular, the PCE APP performs bandwidth usage prediction (i.e., bandwidth calendaring) by looking at the historical trends of all sampled data instead of the instant sampled data. The collected data are traffic engineering data base (TEDB) and LSP-DB, and can also include scheduling information. In addition, the collected data also include auto-bandwidth related changes under particular network events. Using machine learning algorithm, the analytics component is able to correct such changes with the events, and predicts network events and their impact.

3.2. Route Monitoring and Analytics

This subsection introduces the BGP Monitoring Protocol (BMP) [RFC7854] use case in wide area networks (WAN). The BGP protocol is known for its flexibility and ability to manage a large number of neighbors and routes. It is also the basis for many overlay services such as L3VPN, L2VPN and so on. The BMP protocol can be used by the

controller to monitor BGP protocol neighbor status and routing information on the routers.

According to [RFC7854], BMP client located in the router collects BGP neighbor status, routes for each neighbor, and events defined by the user. And then it passes the informations through the BMP protocol to the management station located on the controller. Based on BMP monitoring of BGP, there are three use cases: (1) BGP Route Leaks Monitoring; (2) BGP Hijacks Monitoring; (3) Traffic Analytics.

Route leaks involve the illegitimate advertisement of prefixes, blocks of IP addresses, which propagate across networks and lead to incorrect or suboptimal routing. For case (1), based on BMP, NAI apps can analyze BGP route leaks.

For case (2), by manipulating BGP, data can be rerouted in an attacker's favor out them to intercept or modify traffic. If the malicious announcement is more specific than the legitimate one, or claims to offer a shorter path, the traffic may be directed to the attacker. By broadcasting false announcements, the compromised router may poison the RIB of its peers. After poisoning one peer, the malicious routing information could propagate to other peers, to other Autonomous Systems, and onto the interactive Internet. Based on monitoring BGP routes, ML algorithms can be trained to determine when a hijack has taken place and take necessary actions.

In case (3), with BMP protocol providing BGP changes, together with Telemetry providing network traffic information, The NAI Apps can analyze traffic trends, predict traffic changes, and do traffic optimizing.

3.3. Multilayer Fault Detection In NFV Framework

The high reliability and high availability required for carrier-class applications is a big challenge in virtualized and software-based environment where failures are normal in a software-based environment. The interdependence between NFV's abstraction levels and virtual resources is complex as shown in Fig.. The dynamic characteristics of the resources in the cloud environment make it difficult to locate the fault. So multilayer fault detection for NFV networks and cloud environment will be very useful.

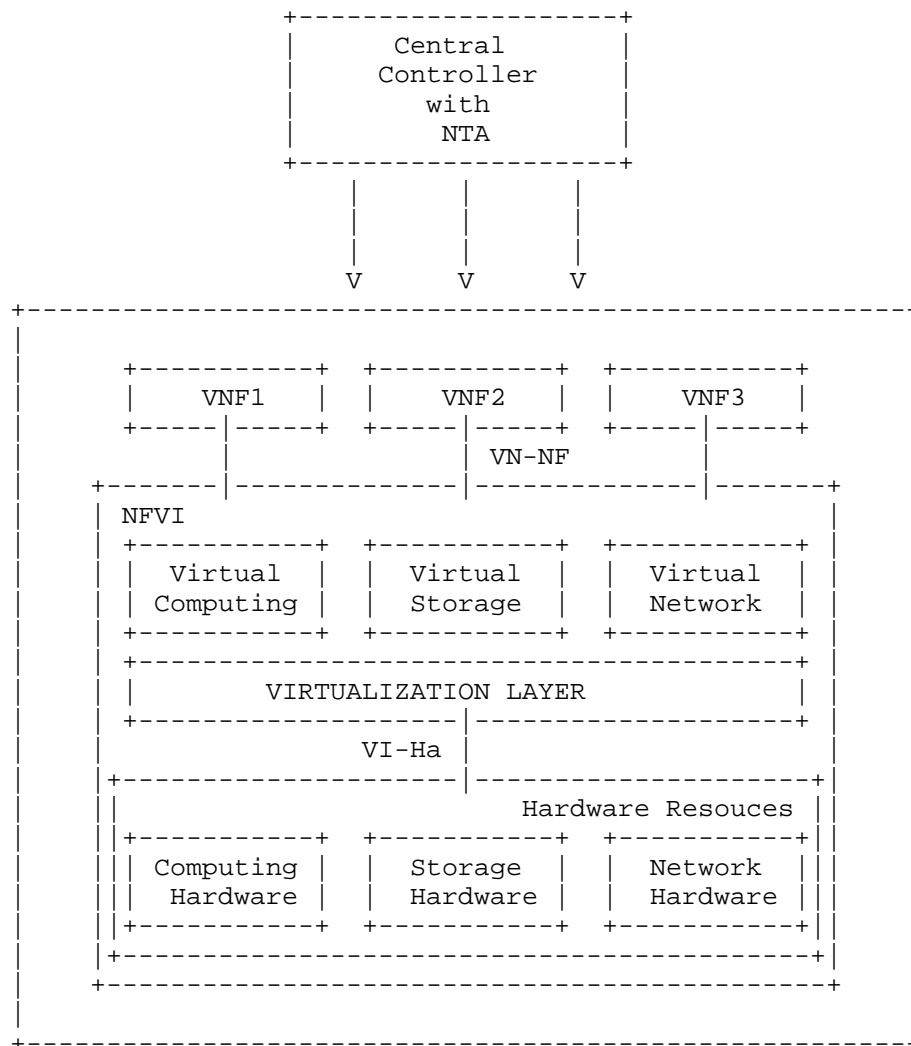


Figure 2 NAI in Multi-layer NFV Framework

For the virtualization layer, CPU performance, memory usage, interface bandwidth and other KPI indicators can be monitored. At the same time resource occupancy and the life cycle of NVF software process can also be monitored. Through the NAI, the relevant statistical data in multiple levels can be analyzed and the models can be setup to locate the root cause for the possible fault in the multi-layer environment.

3.4. Data Center Network Use Cases

Traditionally, data center networks have comprised a large number of switches and routers that direct traffic based on the limited view of each device. With help of SDN/NFV the data center networks are more agile and dynamic to changing usage and traffic patterns. The real-time traffic data and usage can be used to make the data center management and operations intelligent.

Various protocols such as sFLOW, IPFIX could be used to get the port statistics as well as traffic sampling. Over time this information can help build the traffic usage models on a per port and per flow basis. With historical data as the base the NTA engine can predict the traffic usage and make necessary instructions to the SDN controller or NFV orchestrator. These instructions could be reroute a flow to avoid a congested port or scale-in another switch to share load based on the predicted traffic demand.

The NTA engine should find correlation between the various network data to build models and predict the impact of network events, congestions, network utilization patterns etc. Further NTA could detect anomalies based on the historical patterns and help in root cause analysis. The policy framework can be enhanced to consider the analytics.

NTA engine could also get the usage and health information from the Host (servers). Correlation between this information with the information received from network could help in finding security flows and anomalies when the information does not match.

3.4.1. Service Function Chaining

This sub section introduces how to apply NAI to SFC scenario to intelligently reroute/re-optimize the service chains; increase utilization for both Service Functions(SF) and network; intelligent selection of the Service Function Path (SFP) based on data traffic trends.

As per [RFC7665], Service function chaining (SFC) enables the creation of composite (network), services that consist of an ordered set of SFs that must be applied for specific treatment of received packets and/or frames and/or flows selected as a result of classification. The SFs of chain are connected using a service function forwarder (SFF), which is responsible for forwarding traffic to one or more connected SFs according to information carried in the SFC encapsulation, as well as handling traffic coming back from the SF.

The various network telemetry information like delay, jitter, packet loss from the network and the CPU/memory usage utilizations from the SFs, can be collected using sFLOW/gRPC protocol and stored in persistent data repository. The analytics component in NTA can use stored data to build statistics models to predict the impact on various Service Function Paths due to network events, traffic and state of the SFPs and instruct the SDN controller to take necessary actions SDN controller can calculate new paths/reroute the SFC path to avoid congested Ports/SFFs or overloaded SFs. This correlation of application analytics from the SFs and the network analytics from the SFFs could enhance the intelligent management of the service chains for the operators.

The usage and traffic pattern over time can help increase the utilization of SF as well as the underlay network.

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5. Security Considerations

TBD

6. IANA Considerations

This document has no actions for IANA.

7. Acknowledgement

Thanks to Li Zhenbin and Liu Shucheng for their comments and contribution.

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