

# Considerations of deploying AI services in a distributed method

*draft-hong-nmrg-ai-deploy-05*

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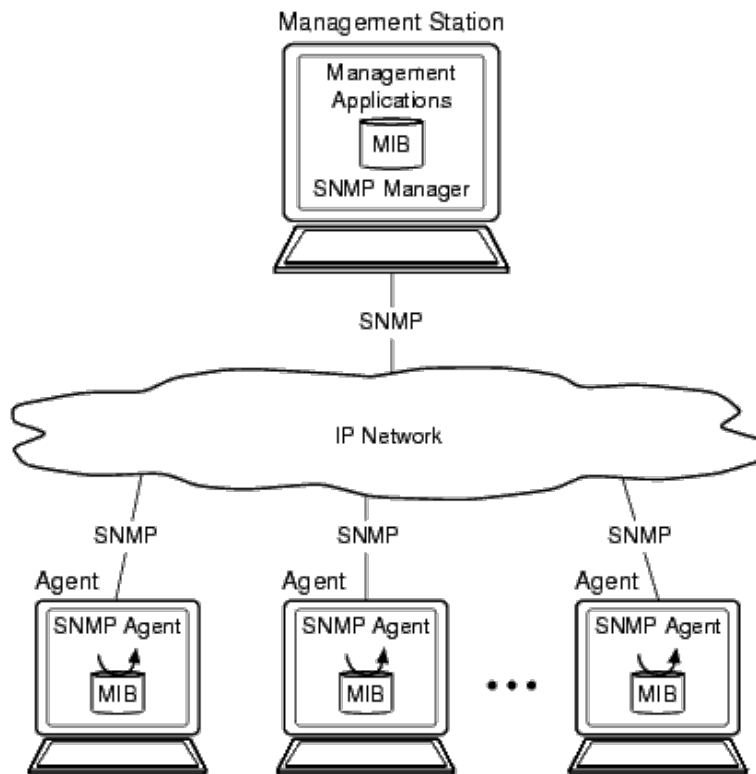
**nmrg Meeting@IETF 118 – Prague**  
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# History and status

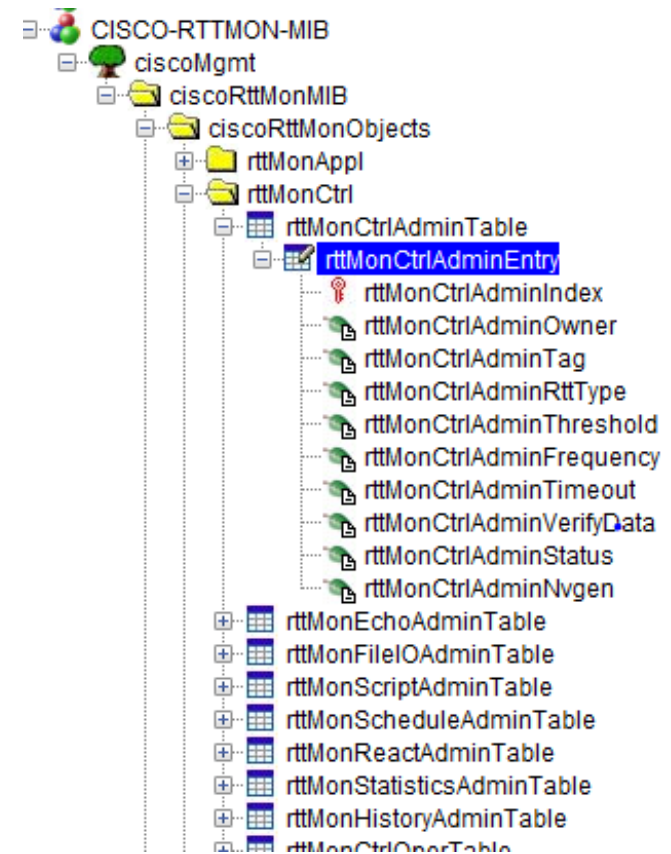
- 00 : draft-hong-nmrg-ai-deploy-00 (Mar. 2022)
- 1<sup>st</sup> revision : draft-hong-nmrg-ai-deploy-01 (Jul. 2022)
  - 1<sup>st</sup> presentation
- 2<sup>nd</sup> revision : draft-hong-nmrg-ai-deploy-02 (Oct. 2022)
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- 3<sup>rd</sup> revision : draft-hong-nmrg-ai-deploy-03 (Mar. 2023)
  - 3<sup>rd</sup> presentation
  - Updated by comments by Alexander Clemm, Jeff Tantsura, Jeferson Campos Nobre
- 4<sup>th</sup> revision : draft-hong-nmrg-ai-deploy-04 (Jul. 2023)
  - Updated to reflect the use case of digital twin networks
- **5<sup>th</sup> revision : draft-hong-nmrg-ai-deploy-05 (Oct. 2023)**
  - **4<sup>th</sup> presentation**
  - **Updated to reflect the use case of digital twin networks and self-driving car**

# Motivations (1/2)

## SNMP-managed Configuration



[Source: Oracle]



[Source: StackExchange]

# Motivations (2/2)

- Deployment of AI services
  - Focus : training (learning) -> inference (prediction)
  - For inference, not only high-performance servers, but also small hardware, microcontroller, low-performance CPUs, and AI chipsets are optimal target device (due to cost)
- Configuration of the system/network in terms of AI inference service
  - For training : accuracy of the model
  - For inference :
    - Target device : Local, edge, cloud
    - Objectives : Accuracy, Latency, Network traffic, Resource utilization, etc.
    - Considerations : Network configuration, AI model, Serving framework, Communication method, device capacity, inference data, etc.

# Intentions of this draft

- Share our experiences and implementation results to find optimal network/system for AI services
  - To find what is important information to provide optimal AI services
  - To find How to deliver these information between related devices
  
- Find common components to provide optimal AI services
  - Common information (similar to MIB)
  - Common system to provide AI services
  - Common network architecture to provide AI services
  - Common protocols to exchange information for AI services
  
- Find useful use cases
  - Self-driving cars
  - Digital twin networks

# Network configuration structure to provide AI services

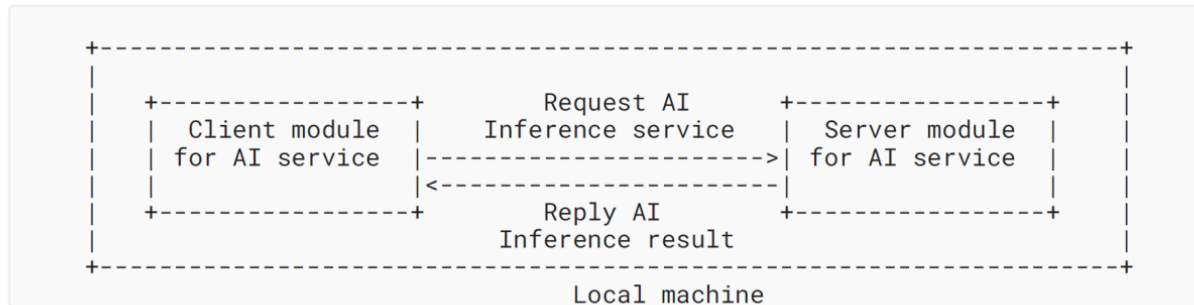


Figure 2: AI inference service on Local machine

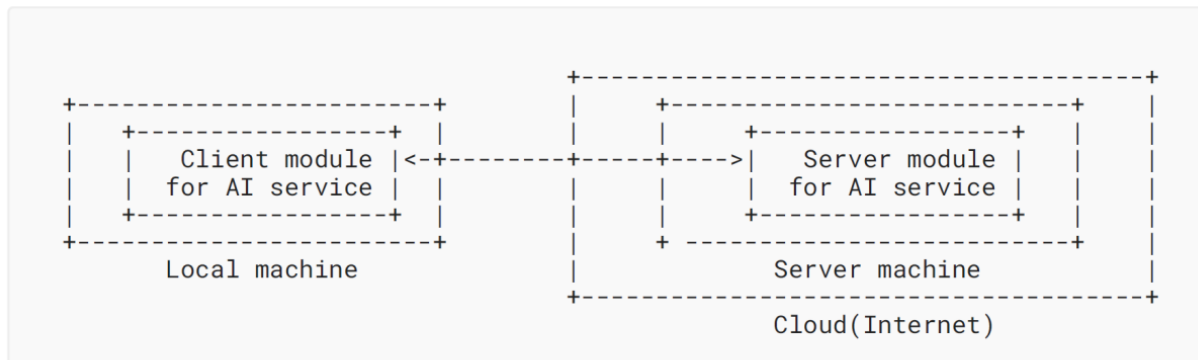


Figure 3: AI inference service on Cloud server

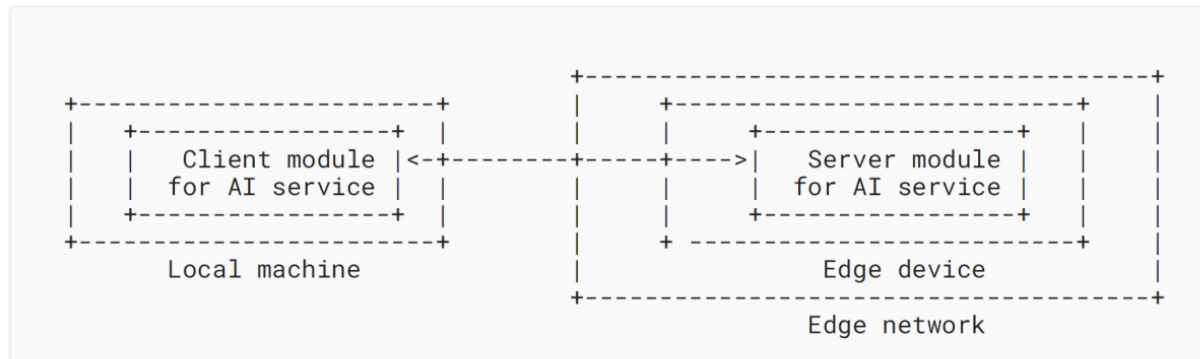


Figure 4: AI inference service on Edge device

# AI inference service on vertical/ horizontal servers

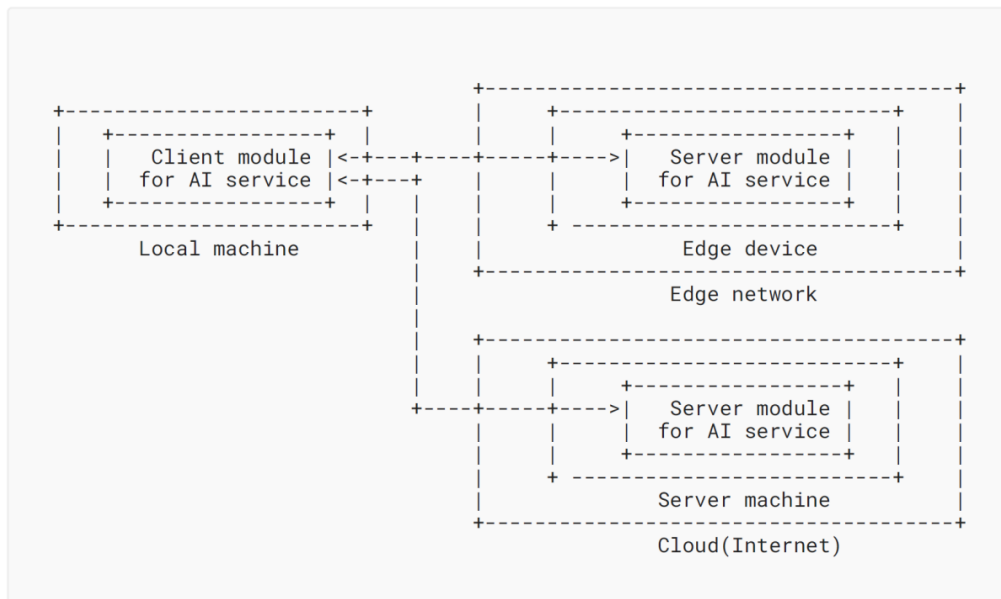


Figure 5: AI inference service on Cloud sever and Edge device

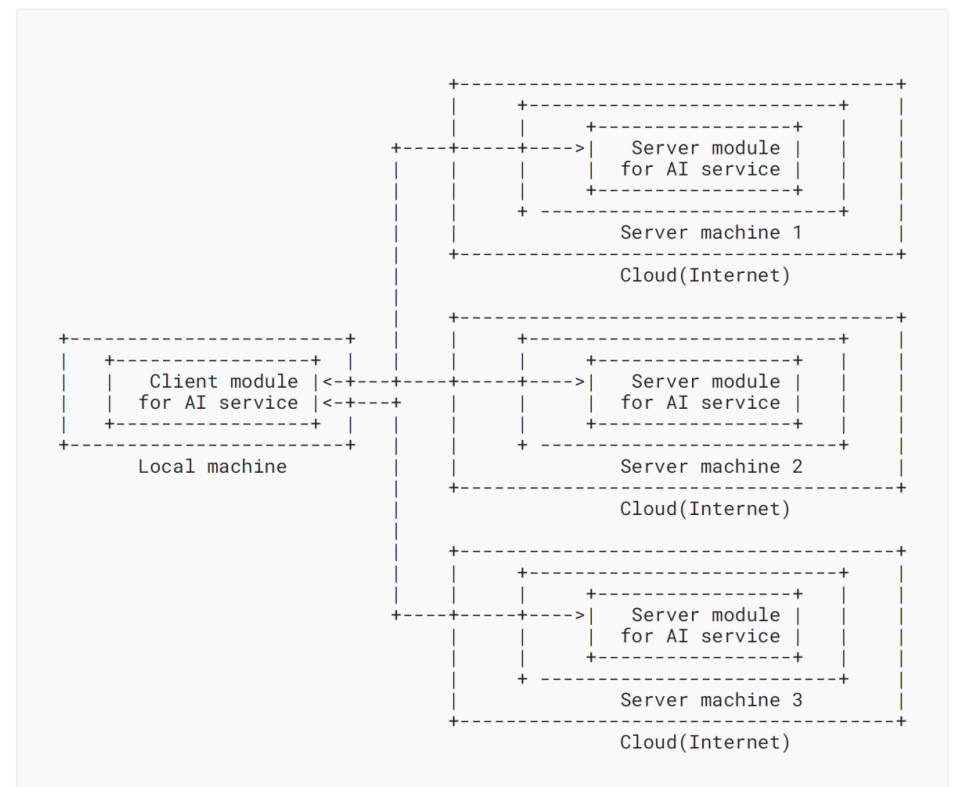


Figure 6: AI inference service on horizontal multiple servers

# Considerations according to the functional characteristics of the hardware

- The performance of AI inference service varies depending on how the hardware such as CPU, RAM, GPU, and network interface is configured for each cloud server and edge device.
- AI inference service can be deployed in the following locations
  - Distant cloud server : High performance and high cost
  - Near edge device : Medium performance and medium cost
  - Local machine : Low performance and low cost
- AI inference service result in (assumption: same AI model)
  - Distant cloud server : High accuracy, short inference time, and long delay to transmit
  - Near edge device : Medium accuracy, medium inference time, and medium delay to transmit
  - Local machine : Low accuracy, long inference time, and short delay to transmit



# Considerations according to the characteristics of the AI model

- AI inference service can be deployed in the following locations
  - Distant cloud server : Heavy AI model, high accuracy, Big size, long inference time
  - Near edge device : Medium AI model, medium accuracy, medium size, medium inference time
  - Local machine : Light AI model, low accuracy, small size, short inference time
  
- AI inference serving framework
  - Traditional web server : ex) FastAPI, Flask, and Django
    - It can be operated on low performance machines
  - Specialized serving framework : ex) Tensorflow serving
    - It can provide high performance.

# Considerations according to the characteristics of the communication method

- AI inference service can be utilized
  - Traditional REST method
    - Common and easily deployed
  - Specified communication method (e.g., gRPC)
    - Better performance but need some works
  
- AI Inference data can be classified
  - Real-time vs. Batch
  - Secure & non-secure

# Use cases

Deploying AI services in Self-driving car

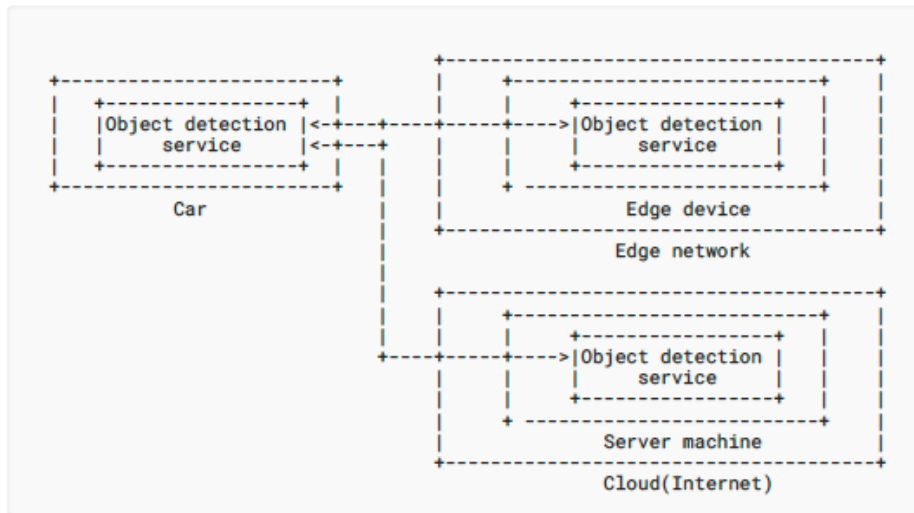


Figure 7: Distributed object detection service in self-driving car

Deploying AI services in DTN

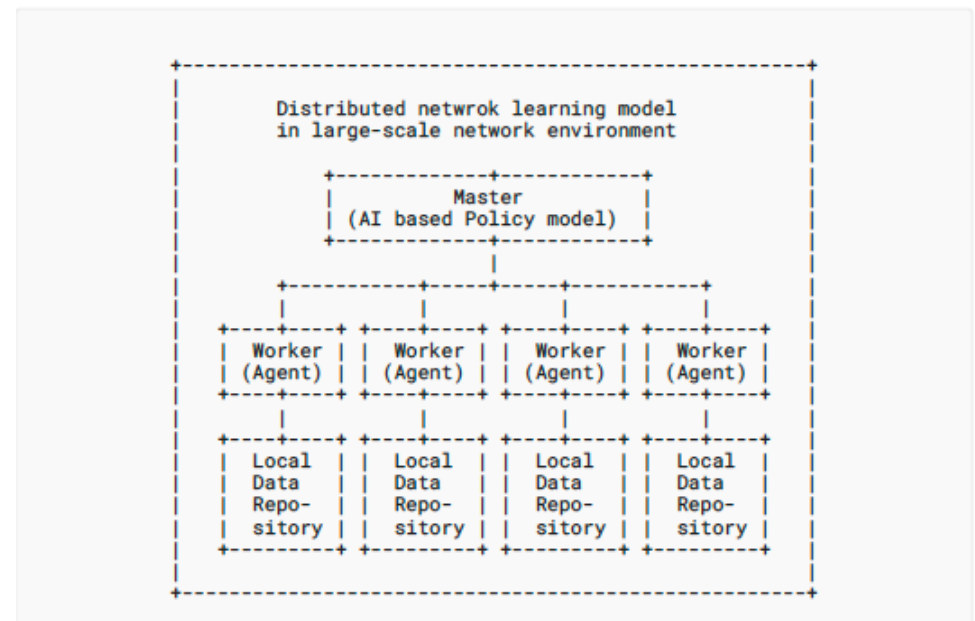


Figure 8: Distributed learning model of network learning for Digital twin network

# Thanks!!

## Questions to NMRG

1. Is it useful and appropriate in NMRG?
2. How to develop this draft?
3. Any feedbacks?