



### Intent Based Networking in 5G Slicing Life-cycle Management and Orchestration

#### Eugen Borcoci University Politehnica Bucharest Electronics, Telecommunications and Information Technology Faculty (ETTI) Eugen.Borcoci@elcom.pub.ro





#### Acknowledgement

- This overview and analysis is compiled and structured, based on several public documents such as: conference proceedings, studies (overviews, tutorials, research papers), standards, projects, etc. (see specific references in the text and the Reference list).
- The selection and structure of this material belongs to the author.
- Notes:
  - Given the extension of the topics, this presentation is limited to a high-level overview only, mainly on concepts and architectural aspects.
  - The presentation takes as an example the utilization of IBN approach in 5G slicing life-cycle management and orchestration.

This study has been partially supported by the NO Grants 2014-2021, under project contract no. 42/2021, RO-NO-2019-0499 - "A Massive MIMO Enabled IoT Platform with Networking Slicing for Beyond 5G IoV/V2X and Maritime Services (SOLID-B5G)".





#### Motivation of this talk

- Current state in networks and services
  - Increased complexity (challenges: integration of cloud/edge computing and networking technologies, 5G, 6G, ... big data, large range of services, ...)
  - Driving forces for extension of IT&C technologies : IoT, smart cities, industry, governance, health systems IoV/automotive, safety/ emergency-oriented systems, environment, entertainment, etc.
- Orchestration, management, and control in networks and services became increasingly complex
  - Need for
    - automation, adaptability, intelligence, programmability, flexibility, sustainability
  - Appropriate technologies have been developed (> 2000)
    - Softwarization/virtualization, Policy Based Management, Cognitive and autonomic management, <u>Intent Based networking (IBN)</u>
    - Artificial Intelligence /Machine Learning (Ai/ML) can offer a strong support for the above





- 1. Introduction
- 2. Network and services management issues and technologies
- 3. Intent Based Networking (IBN)- concepts and architecture
- 4. IBN approach in 5G slicing life-cycle management and orchestration- examples
- 5. Conclusions





- 1. Introduction
- 2. Network and services management issues and technologies
- 3. Intent Based Networking (IBN)- concepts and architecture
- 4. IBN approach in 5G slicing life-cycle management and orchestration- examples
- 5. Conclusions





#### Current state in networks and services

- Increased complexity (challenges: integration of cloud/edge computing and networking technologies, big data, ...)
- IT&C driving forces: IoT, smart cities, industry, governance, IoV/automotive needs, safety/emergency-oriented systems, energy saving, green technologies, entertainment, environment preservation, etc.

#### Example: 5G mobility-capable networks

- high bandwidth, large range of services, large communities of users/terminals (e.g., IoT), inter-operability with clouds
- dedicated logical separated slices- customized for various business demands, programmability through softwarization, virtualization, open sources and open interfaces that allow access for third parties
- Orchestration, Management and Control (OMC) for 5G Multi-x (x= tenant, operator, provider, domain E2E) → – need enhanced concepts and methods and technologies





- **1.** Introduction
- 2. Network and services management issues and technologies
- 3. Intent-based networking (IBN)- concepts and architecture
- 4. IBN approach in 5G slicing life-cycle management and orchestration- examples
- 5. Conclusions





#### **2.1 FCAPS** classical management functions

- **F-Failure** detection (classically-based on monitoring) and repairing
- C- Configuration of the entities (physical, logical)
- A Accounting of resource usage (who, what, when, how much)
- P- Performance evaluation (including enforcement and checking Service Level Agreements fulfillment)
- S Security protection of the system

### FCAPS –tasks that can be supported by cognitive and ML techniques – examples

- Failure Prevention
  - Proactive fault prediction and mitigation
    - predict upcoming failures
    - identify the root cause of the predicted fault and select the mitigation steps





FCAPS –tasks that can be supported by cognitive and ML techniques – examples (cont'd)

- Fault Management in Cloud and Virtualized Environments
  - The multi-tenancy, E2E, multi-domain in cloud/NFV environment raises the complexity and dimensions of the fault space in a network
  - ML (e.g., DeepNNs) can model complex multi-dimensional state spaces -- > used to predict and locate faults
  - Any automated mitigation within a Virtual Network (VN)/slice should not affect other coexisting VNs
    - ML (in particular- Reinforcement Learning combined with DeepNNs) can learn to optimize mitigation steps

Adapted from : Sara Ayoubi, et.al., Machine Learning for Cognitive Network Management, IEEE Comm.Magazine , January 2018, pp.158-165





#### FCAPS –tasks that can be supported by cognitive and ML techniques – examples (cont'd)

#### Performance Management

- Adaptive Probing
  - Large number of devices, parameters, small time intervals to log data → high overhead produced by measurement traffic
  - Regression, mostly based on time series data usable for prediction
  - Solution: adjusting the probing rates, to lower the measuring traffic overhead, while minimizing performance degradation and providing high prediction accuracy

#### Detecting Patterns of Degradation

- detecting the characteristic degradation patterns before the quality drops below an acceptable level
- elastic resource (re)allocation to accommodate dynamic user demands goal: performance optimization and high resource utilization
- however, performance prediction for autonomic tuning of the network behavior is still a research challenge

Adapted from : Sara Ayoubi, et.al., Machine Learning for Cognitive Network Management, IEEE Comm.Magazine , January 2018, pp.158-165





- FCAPS tasks that can be supported by ML techniques examples (cont'd)
  Configuration Management:
  - **Mapping** High-Level Requirements to Low-Level Configurations:
    - There is still a gap between high-level services requirements and lowlevel configurations (e.g., resources to be provisioned)
    - RL techniques can be applied
    - The reward for selecting a configuration setting of a given network element can be seen as the utility indication of that particular setting in delivering the high-level requirements under a given network condition

#### Configuration and Verification

- Configuration changes (e.g., access control lists, routing tables) should comply with high-level requirements and not adversely affect the expected network behavior
- Interest exists in applying DL-aided verification, code correction, and theorem proving

Adapted from: Sara Ayoubi, et.al., Machine Learning for Cognitive Network Management, IEEE Comm.Magazine, January 2018, pp.158-165





#### 2.2 Specific classes of problems in network and service management

- extension of the classical FCAPS
- many of them can benefit from AI/ML support

#### Spectrum management in 5G, 6G

Traffic management and control

 Traffic classification: labeled and unlabeled traffic traces (payload-based; host-behaviour based, flow feature based)

 Traffic prediction: synthetic and real traffic traces with flow statistics; link load and traffic volume prediction in ISP networks; early flow-size prediction and elephant flow detection

•Congestion control: packet loss classification; queue management; congestion window control (for TCP); collect experience from network simulator

#### Resource management

 admission control; resource reservation and allocation (e.g., for slices (RAN, Core network, Cloud), VNFs, etc.)- in multi-domain multi-operator and multitenant contexts; adaptive and automatic re-allocation of resources to meet dynamic needs; synthetic workload with different patterns can be used for training the AI/ML machines





#### 2.2 Specific classes of problems in network and service management (cont'd)

- Fault magement: prediction; detection; localizing the faults; automated mitigation
  Network transport management and control
  - routing strategy; load balancing; QoS assurance with different levels of guarantees for real-time streams; route monitoring; E2E path bandwidth availability prediction and checking; trraffic patterns labeling with routing paths computed by routing protocols; decentralized/ centralized/mixed routing
- QoS and QoE management: QoE optimization; Session quality information; QoS/QoE correlation; QoS prediction under QoS impairment; QoS prediction for media-oriented communications
- Performance prediction: datasets with quality measurements e.g. from public CDNs; throughput prediction (datasets of HTTP throughput measurement)
  Network protection: intrusion detection (misuse-based, anomaly-based); anomaly detection; hybrid intrusion detection
- •Mobile networks: network-level mobile data analysis; mobility patterns analysis, user localization, mobile networks applications
- Internet of Things (IoT) specific
- Wireless sensor networks





#### 2.3 Policy Based Management (PBN) concepts

- **PBM** -separates the rules that govern the behavior of a system from the system functionality; it increases the management flexibility
  - PBM is an imperative management paradigm
  - PBM applied to networks: PBNM
- Today there exist several management architectures such as: SLA-driven, Business-driven, autonomous, adaptive, and self-\* management
- Policy: a set of rules used to manage and control the changing and/or maintaining of the state of one or more managed objects
  - It is defined at a higher level of abstraction, involving a canonical model of systems and devices to which the policy is to be applied
  - definition of a rule : [if.. then ... actions] structure
    - an event occurrence will trigger a rule if a set of conditions are valid and, a set of actions are carried out
    - It specify *what* to be done, *when* and in *which circumstance*
- PBNM offers simple control loops, allowing autonomic behavior: self-\* management properties, (\* = configuration, healing, optimization, protection) (CHOP) –in a similar way as in autonomic managed systems.





- 2.3 Policy and Policy Based Management (PBNM) concepts (cont'd)
- A policy agent on a controller or the device subsequently "renders" the policy, i.e., translates the canonical model into a *device-specific representation*
- The policy definition is de-coupled from policy instantiation and policy enforcement
- Usually the policies are pushed (by a controller) onto devices where they are rendered and enforced; the controller deploys policies across the network and monitors their proper operation
- However, PBM can also include a pull-component; the decision regarding actions to be taken is delegated to a Policy Decision Point (PDP), which:
  - can reside outside the managed device itself
  - typically, has global visibility on context where to make policy decisions
  - usually gets the policies from a Policy Data Base
- Whenever a network device observes an event, it asks the PDP for a decision
  - then, the device carries out the decision as returned by the PDP;
  - the device "enforces" the policy acting as a PEP (Policy Enforcement Point)
- Either way, PBNM architectures typically involve a central component from which policies are deployed across the network and/or policy decisions served.





#### 2.4 Cognitive Management concepts

- Cognitive network management
  - recent trend using AI/ML and in particular to develop self-x capabilities
  - (-x= -aware, -configuring, -optimization, -healing and -protecting systems)
- Cognitive management
   – extension of Autonomic Management (AM) (coined by IBM ~ 2001)- later extended in networking domain → ANM
  - AM + Machine learning = Cognitive Management (CogM)
- Challenge: to deploy the CogM and its orchestration across multiple heterogeneous networks:
  - Radio & Other Access Networks
  - Edge/ Aggregation/ Core Networks,
  - Edge and centralized Computing Clouds
  - Satellite Networks





- 2.4 Cognitive Management concepts (cont'd)
- Autonomous Network Management (ANM) : introduces self-governed networks for pursuing business and network goals while maintaining performance
  - Loop: The Monitor-Analyse-Plan-Execute over a shared Knowledge
  - (MAPE-K) is a control theorybased feedback model for selfadaptive systems
  - Full-duplex communication between *managing systems*, *managed system* and the *environment*
  - ANM hierarchical and recursive approach
  - IETF ANIMA working group

Source: 5GPPP Network Management & Quality of Service Working Group, "Cognitive Network Management for 5G", 2017







#### **2.4 Cognitive Management concepts** (cont'd)

#### Autonomic Network Management functions

- Monitoring: active/passive, centralized/distributed, granularity-based, timing-based and programmable
- Analysis: many approaches exist –relying, e.g., on probability and Bayesian models for anticipation on knowledge, timing, mechanisms, network-level, user applications
  - Challenge: to define a concentrated data set that comprehensively captures information across all anticipation points
  - Recent solutions use learning and reasoning to achieve specific goals
- Planning and Execution
  - The network adaptation plan several aspects: knowledge, strategy, purposefulness, degree of adaptation autonomy, stimuli, adaptation rate, temporal/spatial scope, open/closed adaptation and security
  - Current status: the adaptation solutions differ broadly and there is no unanimity in defining proper planning and execution guideline





#### 2.4 Cognitive Management concepts (cont'd)

#### Autonomic Network Management functions (cont'd)

- Knowledge base
- The network information is shared across the MAPE-K architecture
- Many approaches exist to build knowledge on network/topology, including models from learning and reasoning, ontology and DEN-ng models.
- Integrated solution- able to capture knowledge on: structure, control and behavior

#### Typically:

- In order to drive the decisions of Self Organizing Network (SON)-type (e.g., self-planning, self-optimization and self-healing), the knowledgebased framework should :
  - process the input data from multiple sources
  - extract relevant knowledge, through learning-based classification, prediction and clustering models





- **1.** Introduction
- 2. Network and services management issues and technologies
- 3. Intent Based Networking (IBN)- concepts and architecture
- 4. IBN approach in 5G slicing life-cycle management and orchestration- examples
- 5. Conclusions





- 3.1 General characteristics
- IBN
  - Novel intelligent technology for orchestration and automation of networks
  - IBN let operators focus on their desired outcomes, while leaving details about how those outcomes would be achieved

### Intent Based Networking is about "what" not "how"!

- BN is adopted by IETF, ITU, ... and also industriy (e.g., Cisco, Huawei, etc.)
- IBN can
  - automatically convert, deploy, and configure the network resources according to operator intentions
  - overcome and control abnormal events or failures
  - continuously monitor the network resources to get statistics, in order to help the life-cycle management (LCM) actions upon virtualized subsystems and adjust network functioning and performance towards desired business outcomes

See: A.Clemm et al., Intent-Based Networking - Concepts and Definitions , draft-irtf-nmrg-ibn-concepts-definitions-07, March 2022





- 3.1 General characteristics (cont'd)
- IBN transforms a hardware-centric, manually configured network into a controllerled network, that captures business intent and translates it into policies that can be automated and applied consistently across the network
- IBN uses a centralised controller (e.g., SDN)
- The closed-loop system of IBN operates with functional building blocks:
- *Translation*: capture , translation of intent into policies the network can act on;
- Activation: policies installation across the Physical/Virtual (P/V) network infrastructure, using networkwide automation;
- Assurance: use of analytics and ML to continuously monitor the network, to verify if the desired intent has been applied and the business outcomes are achieved
- Current trends (IETF, etc.) : to address E2E management and not individual devices in isolation : e.g. YANG models for network topology [RFC8345]: service models used by service orchestration systems and controllers [RFC8309])





- 3.1 General characteristics
- The IBN concepts are related to Autonomic Network Management (ANM)
  - ANM lowers operational expenses and creates adaptive management framework
  - ANM introduced "self-management" properties(CHOP)
  - However, ANM still require input from an operator or outside system to provide operational guidance and information about goals, purposes, and service instances
- Note: in this section the word "user" can also refer to operators and administrators who are responsible for the OMC.
- In IBN the input and operational guidance are commonly referred to as "intent"
- The system implementing the IBN is the Intent-Based Systems (IBS)
  - IBS can be implemented in several ways, e.g., a controller or management system are implemented
    - as an application that runs on a server or set of servers
    - or as a set of functions distributed across a network and that collectively perform their intent-based functionality





#### 3.2 IBN- related – definitions

- Intent: a set of operational goals and outcomes, defined in a declarative manner without specifying how to achieve or implement them
- Intent-Based Analytics (IBA): defined and derived from users' intent; it is used to validate the intended state
- Intent-Based System (IBS): s system that supports management functions, guided using intent
- Service Model: a model that represents a service that is provided by a network to an end user
- Single Source of Truth (SSoT): a functional block in an IBS that normalizes users' intent and serves as the single source of data for the lower layers
- Single Version of Truth (SVoT)
- User: In the context of this section, it is an operator and/or administrator responsible for the management and operation of communication services and networking infrastructure (as opposed to an **end user** of a service).

Source: A.Clemm et al., Intent-Based Networking - Concepts and Definitions, draftirtf-nmrg-ibn-concepts-definitions-07, March 2022





- 3.3 Intent based management (IBMg)
- Networks, even if autonomic, have no way of automatically knowing:
  - particular operational goals
  - which instances of networking services to support
- So, networks do not know the network provider (NP) intent (that gives the network the purpose to live). This still needs to be communicated to the network by what informally constitutes intent
- Intent defines goals and outcomes in a *declarative manner*, specifying what to accomplish, not how to achieve it.
- Intent applies several important concepts simultaneously:
  - data abstraction: users do not need to be concerned with low-level device configuration
  - functional abstraction from particular OMC logic: users (here they are operators) are not concerned with how to achieve a given intent
- The desired outcome, is specified by intent and the IBS automatically figures out a course of action (e.g., using an algorithm or applying a set of rules derived from the intent) to achieve the outcome





- 3.3 Intent based management (IBMg)
- Examples of intent (expressed in natural language) :
- Note: the intent specifies "what" and not "how"
  - "Avoid routing traffic originating from a given set of endpoints through a particular vendor's equipment, even if this occurs at the expense of reduced service levels."
  - "Maximize network utilization, unless service levels have deteriorated 20% or more from their historical mean
  - VPN service must have path protection at all times for all paths."

#### Examples of what would not constitute intent :

- "Configure a given interface with an IP address." This is device configuration, not an intent
- "If an interface utilization exceeds a specific threshold, then emit an alert" This is a rule that can support network automation, not an intent.

Source: A.Clemm et al., Intent-Based Networking - Concepts and Definitions, draft-irtfnmrg-ibn-concepts-definitions-07, March 2022





- 3.3 Intent based management (IBMg)
- Specific aspects of Intent, Policy, and Service Models
- All three models involve a higher-layer of abstraction of a network (not involving device-specifics) However differences exist
- Service model: data model describing instances of services
  - It has dependencies on lower-level models (device and network) to describe the service mapping onto the underlying network and IT infrastructure.
  - A service model instantiation requires orchestration
    - The logic to orchestrate/manage/provide the service model and its mapping onto underlying resources is not included in the model itself.
- Policy: set of rules, typically modeled around a variation of events/ conditions/ actions, used to express control loops that can be rendered by devices without intervention by the outside system
  - They let operators to define what to do under what circumstances, but they do not specify the desired outcome





- 3.3 Intent based management (IBMg)
- Specific aspects of Intent, Policy, and Service Models
- Intent : high-level, declarative goal, operating at the level of a network and services it provides, not individual devices
  - It defines outcomes and high-level operational goals, but does not specify how the outcomes should be achieved, neither how goals should specifically be satisfied
  - It does not enumerate specific events, conditions, and actions.
  - The algorithm or rules to apply can be automatically "learned/derived from intent" by the IBS.

#### Subcategories of intent :

- Operational Intent: defines intent related to operational goals of an operator; corresponds to the original "intent" term and the concepts defined in this document
- Rule Intent: a synonym for policy rules regarding what to do when certain events occur
- Service intent: a synonym for customer service model [see RFC8309]
- **Flow Intent:** a synonym for a *Service Level Objective* for a given flow





- 3.3 Intent based management (IBMg)
- Intent based networking architecture



Source: E.Zeydan and Y.Turk "Recent Advances in Intent-Based Networking: A Survey", <u>https://ieeexplore.ieee.org/document/9128422</u>, 2020, Spanish MINECO grant TEC2017-88373-R (5G-REFINE) and Generalitat de Catalunya grant 2017 SGR 1195.





- 3.3 Intent based management (IBMg)
- Intent based networking architecture (cont'd)
  - Business Layer: higher-level declarative policy that operates at the level of a network and services; provide semantic to consume network resources; allowing high-level guidance by a central entity; detect and resolve conflicts between multiple intents.
  - Intent Layer
    - Knowledge: access to knowledge and execute judgement; performs inference from relations between objects
    - Agent: capture the business intent and translate into policies; utilize ontology-based approach to communicate with users; communication interface directly to the network objects
    - Data: keep the state of each intent and the relation between network objects; provides models for the observed data; provides algorithms for data modeling
  - Network Layer : presents the abstraction of domain-specific data and control plane technologies; specifies context-aware architecture for enhancing the network intelligence





- 3.4 Intent based management (IBMg)
- IBN Operational principles
- Single Source of Truth (SSoT) the set of validated intent expressions
  - SSoT and the records of the operational states enable comparing the intended/ desired state and actual/operational states of the system and determining drift between them
  - SSoT + drift information provide the basis for corrective actions
  - IBS
    - can predict states
    - develop strategies to anticipate, plan, and pro-actively act on any diverging trends
- Single Version (or View) of Truth derives from the SSoT
  - It can be used to perform other operations, such as querying, polling, or filtering measured and correlated information in order to create so-called "views."
  - These views can serve the IBS users to create intents as single sources of truth; the IBS must follow well-specified and well- documented processes and models.

See: A.Clemm et al., Intent-Based Networking - Concepts and Definitions, draft-irtf-nmrg-ibn-conceptsdefinitions-07, March 2022





- 3.3 Intent based management (IBMg)
- IBN Operational principles
- One-touch but not one-shot
  - The user expresses intent (in different forms) and then the system takes over all subsequent operations (one-touch)
    - The I/F user-IBS could be designed as an interactive and iterative process.
    - The intent expressions may initially contain more or less implicit parts and unprecise or unknown parameters and constraints
    - The IBS parses, understands and refines the intent expression to reach a well- formed and valid intent expression, further used by the system for operations
    - An intent refinement process could use a combination of iterative steps

#### Autonomy and Supervision

- The IBS autonomous- to conduct its tasks and operations (no user intervention)
- It tqakes its own decisions as to meet the user expectations in terms of (performance, quality) and providing the proper level of supervision





- 3.3 Intent based management (IBMg)
- IBN Operational principles (cont'd)
- Learning
  - IBS can learn, reasoning, get knowledge representation and management
  - Different from an imperative-type of system, (e.g., PBM):
    - the user only declares outcome and not how to achieve these goals
  - The AI/ML naturally provide support for IBS ability for learning

#### Capability exposure

It consists in the need for expressive network capabilities, requirements, and constraints to be able to compose/decompose intents and map the user's expectations to the system capabilities

#### Abstract and outcome-driven (IBN – focused on "what" and not "how"

- Users are not concerned with outcomes
- They can refer to concepts at a higher level of abstractions, independent, e.g., of vendor-specific renderings





- 3.3 Intent based management (IBMg)
- Intent-Based Networking Functionalities
  - Intent Fulfillment : functions and I/Fs that
    - allow users to communicate intent to the network
    - **perform actions** (e.g., algorithms and functions to learn and optimize
    - It can include also traditional functions
      - orchestration of coordinated configuration operations
      - rendering of higher-level abstractions into lower-level parameters
- Intent Assurance : functions and I/Fs that
  - allow users to **validate and monitor** that the network is adhering with intent
  - this assesses the actionseffectiveness
  - feedback -allow training trained or tunning
  - addresses the "intent drift." Intent is not meant to be transactional, i.e., "set and forget", but expected to remain in effect over time (unless explicitly stated otherwise)
    - Intent drift : the system originally meets the intent, but over time gradually change its behavior





- 3.3 Intent based management (IBMg)
- Intent-Based Networking Functionalities (cont'd)
- Intent Fulfilment details
  - Intent Ingestion and Interaction with Users
    - "Ingesting" intent = obtaining intent through interactions with users.
    - The specific functions:
      - recognize intent from interaction with the user
      - allow users to refine their intent so that it becomes actionable by an IBS
      - go beyond those provided by a traditional API
      - may support unconventional human-machine interactions
  - One may have a set of intuitive workflows to guide users through the intent ingestion phase, (assuring that all inputs are gathered and translation have been done)





- 3.3 Intent based management (IBMg)
- Intent-Based Networking Functionalities (cont'd)
- Intent Fulfilment details (cont'd)
  - Intent Translation
    - The IBS core functions translate user intent into courses of action and requests to take against the network, which will be meaningful to network configuration and provisioning systems
    - They are bridging the gap between interaction with users and the traditional management and operations side that orchestrate provisioning and configuration across the network
  - Intent Orchestration- deals with the actual configuration and provisioning steps to be orchestrated across the network and that were determined by the previous intent translation step


#### **3. Intent Based Networking (IBN)- concepts and architecture**



- 3.3 Intent based management (IBMg)
- Intent-Based Networking Functionalities (cont'd)
- Intent Assurance- details
  - Functions to ensure that the network complies with the desired intent once it has been fulfilled.
  - Monitoring
    - It include the usual network monitoring (events, performance measurements) to assess service levels to be delivered
    - Monitored data are inputs for the next set of functions that assess whether the observed behavior is compliant to the intent

#### Intent Compliance Assessment

- Functions to compare the observed network behavior with the intended one
- This includes assessing the effectiveness of intent fulfillment actions
- The assessment results can be fed back to facilitate learning functions that optimize outcomes
- Assessing whether intent drift occurs over time



#### **3. Intent Based Networking (IBN)- concepts and architecture**



- 3.3 Intent based management (IBMg)
- Intent-Based Networking Functionalities (cont'd)
- Intent Assurance- details (cont'd)

#### Intent Compliance Actions

- When intent drift occurs or network behavior is inconsistent with desired intent, these functions are triggering corrective actions to resolve intent drift.
- Or, if necessary, reporting functions need will alert operators (provide information and tools to react appropriately).
- Maybe modifications of the original intent is necessary

#### Abstraction, Aggregation, Reporting

- The Intent Assurance outcome is reported back to allow the user to relate the outcomes to their intent
- Functions are needed to analyze, aggregate, and abstract the results of the observations accordingly





- 3.3 Intent based management (IBMg)
- Intent Life Cycle

|         | User Space   | : | Translat     | ion / IBS  | : Ne    | twork Ops |
|---------|--------------|---|--------------|------------|---------|-----------|
|         |              | : | Spa          | ce         | :       | Space     |
|         |              | : |              |            | :       |           |
|         | ++           | : | ++           | +          | + : +   | +         |
| Fulfill | recognize/ - | > | translate/ - | ->  learn/ | >  c    | onfigure/ |
|         | generate     |   | I I          | plan/      | el l    | rovision  |
|         | intent  <    |   | refine       | render     | I : $I$ | I         |
|         | ++           | : | ++           | +^         | + : +   | +         |
|         | I            | : |              | I          | :       | I.        |
|         |              |   |              |            |         |           |
|         | I            | : |              | ++         | + :     | v         |
|         | I            | : |              | validate   | : +-    | +         |
|         | I            | : |              | +*         | + <     | monitor/  |
| Assure  | ++           | : | ++           | ++         | ⊦ : I   | observe/  |
|         | report   <-  |   | abstract  <- | analyze    | <       | I         |
|         | ++           | : | ++           | aggregate  | : +-    | +         |
|         |              |   |              |            |         |           |

Source: A.Clemm et al., Intent-Based Networking - Concepts and Definitions, draft-irtf-nmrg-ibnconcepts-definitions-07, March 2022



#### **3. Intent Based Networking (IBN)- concepts and architecture**



- 3.3 Intent based management (IBMg)
- Intent Life Cycle (tied to interconnection functions)
  - Intent comes into being, may change over time, and retracted
  - **Two** functional planes, (**fulfillment and assurance**); **three** (vertical) spaces
  - User Space -its functions interface the network and IBS with the human user
    - Users can articulate intent and IBS recognize that intent
    - Functions reporting back the network status and which allow users to assess outcomes and whether their intent was achieved

#### IBS functions

- Translation
  - bridge the gap between intent users and the network operations infrastructure
  - translate an intent into a course of actions
- Algorithms to plan and optimize actions based on network feedback
- Analyze and aggregate network observations in order to validate compliance with the intent and to take corrective actions
- Abstract network observations
- Network Operations Space: traditional orchestration, configuration, monitoring, and measurement functions



#### **3. Intent Based Networking (IBN)- concepts and architecture**



- 3.3 Intent based management (IBMg)
- Intent Life Cycle (cont'd)
- There is an "inner" intent control loop, fully autonomic between IBS and Network Operations
  - The closed-loop involves automatic analysis and validation of intent based on observations from the network operations space
  - In order to make adjustments in the network configuration, the observations are input into the functions that plans the rendering of networking intent
  - The loop
    - addresses and counteracts any intent deviation/drift that may occur
    - allows to assess the effectiveness of actions taken in order to continuously learn and improvement
- The "outer" intent control loop is extended to the user space
  - It includes the user taking actions and (possibly) adjusting their intent based on observations and feedback from the IBS
  - Intent is thus subjected to a lifecycle: It comes into being, may undergo refinements, modifications, and changes in time, and retracted later





- **1.** Introduction
- 2. Network and services management issues and technologies
- 3. Intent Based Networking (IBN)- concepts and architecture
- 4. IBN approach in 5G slicing life-cycle management and orchestration- examples
- 5. Conclusions





#### 4.1 5G Slicing –overall architecture



Source: 5GPPP Architecture Working Group, View on 5G Architecture, Version 2.0, December 2017



- 4.2 5G slicing orchestration, management and control (OMC) aspects Work phases
  - Service/data model & mapping on slices
  - Customized slice design ,preparation, instantiation stitching / composition in a single domain and cross-domain
  - Network slice life cycle management- LCM (monitoring ,updating, ..)

#### OMC system:

- may
  - provision initial slice instances
  - or create them at tenants' requests
- must react during slice life cycle, by taking decision to adapt resources to dynamic conditions of the network and user needs
- Cognitive network management technologies can be a strong help to traditional OMC
  - Using AI/ML one can achieve: self-x characteristics of the network, where [x= awareness, configuration, optimization, healing, protecting]





### 4.2 Example: 5G slicing management and control (M&C) aspects Network functions requiring automation

- Planning and design: Requirements and environment analysis, topology determination
- Construction and deployment: Static resource allocation, VNF placement, orchestration actions
- Operation, control and management: Dynamic resource (re)allocation, adjustment; policy adaptation; it interact bi-directionally with
  - Fault detection: Syslog analysis, behavior analysis, fault localization
  - Monitoring: Workload, performance, resource utilization
  - **Security:** Traffic analysis, DPI, threat identification, infection isolation

Adapted from source: V. P. Kafle, et. al., "Consideration on Automation of 5G Network slicing with Machine Learning", ITU Caleidoscope Santafe 2018





### 4.2 5G slicing orchestration, management and control (M&C) aspects Network Slice Instance Life

Generic view



Source: A.Galis, T. Francesc, C.E, Rothenberg, S.Clayman, "Slicing 5G Networks: An Architectural Survey",2020, DOI: 10.1002/9781119471509.w5GRef095, <u>https://www.researchgate.net/publication/341437156</u>





#### 4.2 5G slicing orchestration, management and control (M&C) aspects

#### Network Slice Instance (NSLI) Life

- Preparation phase: creation and verification the feasibility of NSL template(s); preparation of the network environment to support the NSLIs lifecycle (e.g., provisioning databases)
- Instantiation / configuration:
  - Instantiation, configuration and activation of shared / non-shared NFs
  - All resources shared/ dedicated to the NSLI are created and configured, i.e. to a state where the NSLI is ready for operation
  - Activation makes the NSLI active, e.g., diverting traffic to it.
- Run-time phase: NSLI handles traffic to support services of certain type(s)
  - Supervision/reporting (e.g., for KPI monitoring)
  - Modification could be : upgrade, reconfiguration, NSLI scaling, changes of NSLI capacity or topology, association and disassociation of NFs with NSLI
- Decommissioning phase
  - Deactivation (taking the NSLI out of active duty); free the dedicated resources (e.g. termination or re-use of NFs) and re-configuration/updating of shared/dependent resources.
  - Finally, the NSLI does not exist anymore.

Source: A.Galis, T. Francesc, C.E,.Rothenberg, S.Clayman, "Slicing 5G Networks: An Architectural Survey",2020, DOI: 10.1002/9781119471509.w5GRef095, <u>https://www.researchgate.net/publication/341437156</u>





- 4.2 5G slicing orchestration, management and control (M&C) aspects
- 3GPP Vision on LCM of a network slice



Adapted from: 3GPP- TS 28.530 v 16.5.0 5G; Management and Orchestration; Concepts, Use Cases and Requirements, 2021, https://www.etsi.org/deliver/etsi\_ts/128500\_128599/128530/15.00.00\_60/t%s\_128530v150000p.pdf





- 4.2 5G slicing orchestration, management and control (M&C) aspects
- 3GPP Vision on LCM phases of a network slice instance (cont'd)
- LCM : commissioning, activation, runtime monitoring , decommissioning
  - Commissioning : the preparation and design for a slice is handled, e.g., a slice template is created (it contains information about the user requirements and resources to be deployed)
    - After that, the designed slice template is deployed on the underlying infrastructure with NFVO and network controllers
  - Activation: the specified resource should be assigned to a slice and activated.
  - Run-time operations and monitoring : the activated slice resources are monitored continuously
    - any failure → the NFVO orchestrator is notified in order to resolve the error
  - Decommissioning: the created slice instance is deleted automatically at the specified time

Source: 3GPP- TS 28.530 v 16.5.0 5G; Management and Orchestration; Concepts,Use Cases and Requirements, 2021, <u>https://www.etsi.org/deliver/etsi\_ts/128500\_128599/128530/15.00.00\_60/t%s\_128530v150000p.pdf</u>





#### 4.3 Example of Intent Based Management applied in 5G slicing LCM

- The main phases are those defined by 3GPP LCM i.e.:
  - commissioning, activation, runtime monitoring, and decommissioning
- The IBS includes a
  - Web-based Graphical User Interface (GUI)
  - IBN manager (mediator for all IBN components)
  - Knowledge base/ policy store, policy
  - Configurators/slice template generators
    - Performs Intent translation for
      - RAN policy generator for RAN network
      - Open Source MANO (OSM) policy generator- for Core network
  - Al-based intelligent update engine
    - e.g., engine belonging to Deep Learning (DL) family

Adapted from source: K. Abbas, T.A.Khan, M.Afaq, and W.C.Song, "Network Slice Lifecycle Management for 5G Mobile Networks: An Intent-Based Networking Approach", IEEE Access, 2021





#### 4.3 Example of Intent Based Management applied in 5G slicing LCM (cont'd)

- The users input their slice requirements (**user intents**) through IBN GUI
- The policy DB contains : information on VNFs, instance images, versions, IP addresses, and infrastructure and information related to the deployed intents
- The IBN manager
  - prepares a VNF forwarding graph (VNFFG) by mapping the intent requirements into the required resources and
  - sends it to the slice template generator which has policy configurators for translating higher-level QoS requirements into policy configurations
- The policy configurators forwards the slice templates to
  - the Open Source MANO (OSM) for core part control
  - the RAN controller
- The IBNS intelligent update and assurance engine automatically updates the resources at runtime in failure or overload cases





Learning

model

Update and insurance engine

Monitoring Data

Update

manager

#### 4.3 Example of Intent Based Management applied in 5G slicing

- Life Cycle Management
- Main functional components
  - **IBN System** 
    - Slice preparation and design
      - Data Base, Policy store
      - **IBN** Manager
    - Intent translation/Policy configurator (RAN, OSM)
    - Update and insurance engine
    - Run-time monitoring

#### Slice control and activation module

- **RAN** Controller
- **NEV** Orchestrator
- **RAN Domanin, Core Domain** 
  - (slice instances)

#### Adapted from source

K. Abbas, T.A.Khan, M.Afaq, and W.C.Song, "Network Slice Lifecycle Management for 5G Mobile Networks: An Intent-Based Networking Approach", IEEE Access, 2021

InfoSys 2022 Congress, May 22-26, 2022 - Venice, Italy



Intent

IBN

Manager

**OSM Policy** 

configurator

Slice template

**IBN System** 

DB

Knowledge

Base

Policy Store

**RAN Policy** 

Configurator

Slice template

RAN

controller

RAN

Domain

Slices

REST

API

Mon

Slice preparation and design

Sync

Intent translation through Policy

JSON

REST

API

Resource

activation

Mon





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- Slice Commissioning
- The users (customers, MNOs, resource vendors) input (via GUI) slice QoS requirements in the form of user intents
  - Intents : higher-level abstract requirements (SLAs) for a slice
    - e.g., slice information (slice ID/S-NSSAI), QoS requirements (up-rate, down-rate), service type (eMBB, URLLC, IoT), slice instantiation time, and deletion time-period
  - The IBN manager interacts with the network orchestrators the REST API I/F
  - When a user intent comes, the IBN manager fetches the resource and architecture information from the knowledge DB, maps them with the available resources, and creates a VNF graph (VNFG)
    - The knowledge base or catalog repository stores all the information regarding the underlying resources, (VNFs, PNFs, IP addresses scheme, instance images, and deployed intents)
    - The generated VNFG is sent to slice template generators/policy configurators
    - The slice template generators receive and translates the user intent forward into the underlying NFVO acceptable format





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management: Slice Commissioning (cont'd):
- Implementation examples
- The **OSM** slice template generator generates the slice template into JSON string format (OSM NFVO accepts this form) and similar actions are performed by slice template generator for RAN
- Multiple slice template generators can exist, for multi orchestrators (e.g., NFVO OSM, SDN RAN controller FlexRAN and ONOS controller)
- The CORD (Central Office Re-architected as a Datacenter) (https://opennetworking.org/cord/) is a platform that leverages SDN, NFV and Cloud technologies to build agile datacenters for the network edge. Integrating multiple open source projects, CORD delivers a cloud-native, open, programmable, agile platform for network operators to create innovative services.
- NX.Foukas, N.Nikaein, M.M. Kassem, M.K.Marina, K Kontovasilis ,"FlexRAN: A Flexible and Programmable Platform for Software-Defined RAN", European FP7 FLEX project (612050), <u>https://dl.acm.org/doi/10.1145/2999572.2999599</u>

https://opennetworking.org/cord/, https://opennetworking.org/m-cord/

"Open Source Mano" https://osm-download.etsi.org/ftp/Documentation/201902-osm-scopewhite-%paper/#!02-osm-scopeand-functionality.md





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management: Slice Commissioning
- Implementation examples (cont'd)
- *The* **M-CORD** is an open source reference solution (<u>https://opennetworking.org/m-cord/</u>), built on CORD for 5G mobile networks. It is a cloud-native solution built on SDN/NFV/Cloud technologies. It includes virtualized RAN functions and a *virtualized mobile core (vEPC)*. M-CORD enables the creation of use case-specific services that can be dynamically scaled.
- The **(TOSCA)** Topology and Orchestration Specification for Cloud Applications (https://www.techtarget.com/searchcloudcomputing/definition/TOSCA-Topology-and-Orchestration-Specification-for-Cloud-Applications)
  - open-source language used to describe the relationships and dependencies between services and applications that reside on a <u>cloud computing</u> platform.
- In the system example, the M-CORD policy configurators convert the user requirements in TOSCA format. The slice template is designed and prepared In the slice instantiation module, according to the underlying NFVOs.





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- Slice Activation
  - Slicing the Core network resources
  - Network Orchestrator (OSM) (ETSI)- enables MNOs to automatically deploy E2E network services, based on NFV.
  - Components: NFVO (NFV orchestrator); VNFM (Virtual Network Function Manager); VIM (Virtual Infrastructure Manager); see ETSI NFV
  - A GUI portal allows for automatically managing, monitoring, and deploying the network resources
  - The NFVO OSM
    - deploys Core network VNFs such as Evolved Packet Core (EPC) VNFs (vMME, vHSS, vSPGW)
    - facilitates slice LCM in E2E context
  - The OSM slice template generator/ OSM policy configurator generates the slice configurations for OSM NFVO (e.g., in the JSON string format)
  - The slice template is forwarded to OSM (vie REST API)
  - The OSM deploys the core network VNFs with the help of VIM (OpenStack) over the infrastructure





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Slice Activation (cont'd)
  - Slicing the RAN resources
  - Inside IBN, the RAN slicing policy configurator converts the high-level slice configuration provided at the IBN tool to JSON slice template format and send it to the underlying RAN controller
  - A SDN controller could be used, (e.g. FlexRAN), to manage multiple BSs and to slice RAN resources; it contains a Control plane (CPL) and an agent API
  - The agent acts as a local controller, interacting with the other agents and the master controller
    - It can control one eNodeB at a time
    - The agent API : SB I/F decoupling the CPL from the eNodeB DPL
  - On top of the RAN controller: control and monitoring applications exist and communicate with the master controller through the NB interface
    - for automatic management, modifications, and control of the RAN resources
  - RAN controller deploys these configurations at eNodeB for slice creation

Source: NX.Foukas, et al.,,"FlexRAN: A Flexible and Programmable Platform for Software-Defined RAN", European FP7 FLEX project (612050), <u>https://dl.acm.org/doi/10.1145/2999572.2999599</u>





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- Run-time monitoring and dynamic scaling
  - RAN domain (eNodeBs) monitoring
    - RAN controller and a special elasticMon tool
      - monitor the high real-time data traffic to control and manage it
      - store the statistics of the eNodeBs
  - Visualizing tools (e.g., Kibana, Grafana) can be used on the top of the RAN controller and *elasticMon* to visualize the data traffic on runtime

#### Core network monitoring

- Openstack telemetry service or ceilometer can monitor the VNFs
- On top of that, MON OSM monitoring service collects the VNFs data logs and further forward them to a DB (e.g., Prometheus -an open-source tool for monitoring the real-time data traffic and store them into a timeseries database).
- The DB keeps the VNFs data logs into the real-time database repository
- Visualizing tools are integrated with the DB to visualize theVNFs resource status (CPU, RAM usage, etc.)
- Failures  $\rightarrow$  reconfiguration and dynamic updating the slice resources





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Slice decommissioning
- Two ways for decommissioning
  - The slice termination request is generated through the IBN system web portal, which contains the slice SNSSAI (single network slice selection assistance information), SliceID.
    - The slice deletion configurations are rendered the same as the slice creation process
    - The slice deletion template is forwarded to the NFVOs; the OSM and RAN controller can delete the resources
  - The slice instance life time is inserted during the slice creation phase
    - When the life time is finished the IBS automatically generates the slice deletion request and performs the same process
    - A deletion confirmation notification is sent back to the operator through the IBN platform.



### 4.3 Example of Intent Based Management applied in 5G slicing Slice Life Cycle Management -summary



See for details- Source: K. Abbas, T.A.Khan, M.Afaq, and W.C.Song, "Network Slice Lifecycle Management for 5G Mobile Networks: An Intent-Based Networking Approach", IEEE Access,2021 InfoSys 2022 Congress, May 22-26, 2022 - Venice, Italy



### 4.3 Example of Intent Based Management applied in 5G slicing Slice Life Cycle Management –summary (cont'd)



See for details- source :K. Abbas, T.A.Khan, M.Afaq, and W.C.Song, "Network Slice Lifecycle Management for 5G Mobile Networks: An Intent-Based Networking Approach", IEEE Access,2021 InfoSys 2022 Congress, May 22-26, 2022 - Venice, Italy





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Slice Life Cycle Management –summary (cont'd) (see the previous slides
- Slice commissioning
  - The users define slice requirements (in intent form) of through the GUI of the IBN system
  - The IBN policy configurators translates the requirements into policy configurations for each domain and forwards them to OSM and RAN controller to deploy resources
  - The OSM prepares the core network EPC VNFs resources for the requested slice using Open-Stack
  - The RAN controller deploys the slice configurations at the RAN domain
  - The slice configurations include information about dedicated vMME, slice SNSSAI, and QoS requirements used to stitch RAN slice with core VNFs
  - The IBN notifies the user of successful slice activation





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Slice Life Cycle Management –summary (cont'd) (see the previous slides
- Slice decommissioning
  - The user inputs the network sliceID information
  - TheIBN manager checks the slice information from the catalog repository of the IBS
  - The IBN again translates the slice information into policy configurations and forwards them to OSM and RAN controller to delete resources associated with that slice instance
  - The OSM with OpenStack deletes the core EPC VNFs, and the RAN controller deletes the slice at eNodeB. The OSM and RAN controller notify the IBN platform about slice deletion





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- User Intents and contract design
  - The contract design in IBS define the underlying platform resources requirement in an understandable way by the network operators.
  - The contract design is the user intent for creating a slice (resource requirements)
  - User contract main fields: S-NSSAI (single network slice assistance information), architecture ID, up-rate, down-rate
    - S-NSSAI identifies the service domain for each slice (URLLC, eMBB, mMTC, etc.)
    - The network architecture type can be LTE, LTE advanced, 5G, etc.
    - The QoS parameters contains slice up-rate and down-rate data requirements (bandwidth, latency)
- The IBN framework has an interactive web-based front-end for NOs
  - The Nos can input the information of the user contracts
- The IBN framework generates the service graph for network resources allocation (RAN, core) and suitable network functions for a requested slice.





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- User Intents and contract design



Adapted from source: K.Abbas, M.Afaq, T.A.Khan, A.Rafiq and W-C Song, "Slicing the Core Network and Radio Access Network Domains through Intent-Based Networking for 5G Networks", Electronics 2020, 9, 1710; doi:10.3390/electronics9101710, www.mdpi.com/journal/electronics





- 4.3 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- Intelligent self- assurance and updating methods in IBN (ML –based)
  - Machine learning-based update and assurance engine example
    - Working phases: real-time cloud resource time-series dataset, data processing, data splitting, ML model processing, decision engine
    - The ML provides predicted resource utilization (CPU, RAM, storage, throughput); the decision engine decides on the updating



Adapted from source: K. Abbas, T.A.Khan, M.Afaq, and W.C.Song, "Network Slice Lifecycle Management for 5G Mobile Networks: An Intent-Based Networking Approach", IEEE Access, 2021

InfoSys 2022 Congress, May 22-26, 2022 - Venice, Italy





- 4.4 Network applications examples and relevant ML techniques
- ML summary
- Learning methods in basic ML
  - Supervised learning (SML)-predicting one or more dependent variables
    - based on (initially) labeled data
    - use cases examples: classification and regression
    - semi-supervised learning (SSML): not all data is labeled
  - Unsupervised learning (UML)- look for structure in (unlabelled) data sets
    - use cases examples: clustering or pattern mining
- Specific classes of ML
  - Reinforcement learning (RL) -using feedback to an agent actions in a dynamic environment; figures out predictions through trial and error
    - use cases examples: network and services management, self driving cars,
  - Deep learning (DL) use a cascade of multiple layers of nonlinear processing units for feature extraction and transformation
    - use cases examples: network and services management, ....
  - Deep Reinforcement Learning: DRL = RL + DL (general intelligence)
    - RL defines the objective; DL gives the mechanism
- Note: a complete ML taxonomy is very rich; it is out of scope for this presentation





- 4.4 Network applications examples and relevant ML techniques
- ML summary
- Deep Learning (DL) versus Reinforcement Learning (RL)
- Deep learning (DL)
  - subset of ML based on artificial neural networks (ANN/NN)
  - it is considered as one of the many ML methods
  - types: Convolutional NN, Recurrent NN, ..
  - use the current information in teaching algorithms to look for pertinent patterns which are essential in forecasting data
  - applies learned patterns to a new set of data (while RL gains from feedback)
  - is also known as hierarchical learning or deep structured learning
  - the applications are more often in recognition and area reduction tasks
  - can be supervised, unsupervised, semi-supervised, self-supervised, or even RL-mode (depending mostly on how the neural network is used)





- 4.4 Network applications examples and relevant ML techniques
- ML summary (cont'd)
- Deep Learning versus Reinforcement Learning
- Reinforcement Learning (RL)
  - is one of the main basic ML paradigms
  - in supervised and unsupervised learning, there is no 'consequence' to the ML engine if it fails to properly understand or categorize data
  - however, RL, generally figures out predictions through trial and error
  - applications are usually oriented on systems with environment interaction and optimal control objectives
  - does not need a current data set to learn (it can be unsupervised -type)

#### Working phases

**Training phase:** is to tell the agent which action shall be taken, under a given environment, from a series of trials

**Inference phase**: the agent takes appropriate actions according to the experience learned during the training







#### 4.4 Network applications examples and relevant ML techniques

| Networking application |                                 | Steps of ML workflow                                    |   |  |   |   |  |  |
|------------------------|---------------------------------|---|---|--|---|---|--|--|
| General objective      | Specific<br>works               | Problem<br>formulation                                  | Data collection   |  | Data analysis   | Offline model<br>construction   | Deployment and<br>online inference   |  |
|                        |                                 |   | Offline collection  | Online<br>measurement                                  |   |   |  |  |
| Traffic<br>prediction  | traffic<br>volume<br>prediction | SL: prediction<br>with Hidden-<br>Markov<br>Model (HMM) | Synthetic and<br>real traffic<br>traces with flow<br>statistics             | Observe the flow<br>statistics                         | The flow<br>count and the<br>traffic volume<br>are correlated                   | Training HMM<br>model with Kernel<br>Bayes Rule and<br>Recurrent Neural<br>Network with Long<br>Short Term Memory<br>unit | Take flow<br>statistics as<br>input and obtain<br>the output of<br>the traffic<br>volume |  |
| Resource<br>management | job<br>scheduling               | RL: decision<br>making with<br>deep RL                  | Synthetic<br>workload with<br>different<br>patterns is used<br>for training | The real time<br>resource demand<br>of the arrival job | Action space is<br>too large and<br>may have<br>conflicts<br>between<br>actions | Offline training to<br>update the policy<br>network   | Directly<br>schedule the<br>arrival jobs with<br>the trained<br>model                    |  |

Adapted from source (see details there): M.Wang, et al., "Machine Learning for Networking: Workflow, Advances and Opportunities", IEEE Network Vol.32, Issue 2, March-April 2018





#### 4.4 Network applications examples and relevant ML techniques (cont'd)

| Networking application a                       |                             | Steps of ML workflow  |  |   |  |   |   |  |
|--|-----------------------------|---|--|---|--|---|---|--|
| General<br>objective                           | Specific works              | Problem<br>formulation  | Problem Data collec<br>formulation   |   | Data analysis  | Offline model construction  | Deployment and online inference   |  |
|  |                             |   | Offline collection   | Online mon  |  |   |   |  |
| Network<br>adaptation<br>nd perf<br>prediction | routing<br>strategy         | SL: decision<br>making with<br>Deep Belief<br>Architectures<br>(DBA)                          | Traffic patterns<br>labeling with<br>routing paths<br>computed by<br>OSPF protocol | Online<br>traffic<br>patterns in<br>each router             | Pb.: difficult to<br>characterize the<br>I/O patterns to<br>reflect the<br>dynamic nature of<br>large-scale het.<br>networks | Take the Layer-Wise<br>training to initialize and<br>the backpropagation<br>process to fine-tune<br>the DBA structure | Record and collect the<br>traffic patterns in each<br>router periodically and<br>obtain the next routing<br>nodes from the DBAs |  |
|  | general QoE<br>optimization | RL: decision<br>making with a<br>variant of UCB<br>Upper<br>confidence<br>bound)<br>algorithm | Session quality<br>information with<br>features in large<br>time scale             | Session<br>quality<br>information<br>in small time<br>scale | Application<br>sessions sharing<br>the same features<br>can be grouped   | Backend cluster<br>determines the session<br>groups using CFA with<br>a long time scale                               | Front-end performs the<br>group-based<br>exploration-<br>exploitation strategy in<br>real time                                  |  |
|  | throughput<br>prediction    | SL: prediction<br>with HMM  | Datasets of<br>HTTP<br>throughput<br>measurement<br>from iQIYI                     | users' s<br>session<br>features as<br>input                 | Sessions with<br>similar features<br>tend to behave in<br>related pattern  | Find set of critical<br>feature and learn a<br>HMM for each cluster<br>of similar sessions                            | new session is<br>mapped to the most<br>similar session cluster<br>and corresponding<br>HMM are used to<br>predict throughput   |  |

Adapted from source (see details there): M.Wang, et al., "Machine Learning for Networking: Workflow, Advances and Opportunities", IEEE Network Vol.32, Issue 2, March-April 2018





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management
- Intelligent self- assurance and updating methods in IBN II (cont'd)
  - IBN-based closed-loop system is capable of self-assurance and updating
  - The IBS performs monitoring and can use machine learning (ML) to autonomously update the system to dynamic conditions
  - Several features are possible: service assurance, update, failure detection, auto-recovery, etc.
  - Network resource continuous monitoring will support adaptation of service provisioning, according to the system state
  - ML proactive models can be used to observe the resource status and decide to update the system according to predicted future needs
  - The intelligent engine's work may consist in a chain of phases: real-time cloud resource time-series dataset, data processing, data splitting, DL model, and decision engine




- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management
- Intelligent self- assurance and updating methods in IBN
  - Example of ML model appropriate for an IBN engine
  - Deep learning (DL) can solve problems, e.g., prediction, classifications, detection, etc.; Deep Neural Networks (DNN) implement DL
  - DL DNN- process the data in a *feed-forward way* from the input layer to several hidden layers and the final output layer
    - Each node (neuron) is directly connected with other nodes of the next layer
    - The first input layer has multiple input neurons that are further connected with multiple weight factors according to the design of the model and dataset features.
  - Recurrent Neural Network (RNN)- special class DNN :
    - The connections between nodes form a *directed* or *undirected* graph along a **temporal sequence**
    - This allows it to exhibit temporal dynamic behavior
    - RNNs are derived from feedforward neural networks, they can use their internal state (memory) to process variable length sequences of inputs
    - RNN can run arbitrary programs to process arbitrary sequences of inputs





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Example of ML model appropriate for an IBN engine (cont'd)
  - Recurrent Neural Network (RNN)
    - Ordinary Feedforward Neural Networks FNNs are only meant for data points, which are independent of each other; FNN pass information through the network without cycles
    - However, for data in a sequence with dependence between points, a modification in NN is needed to model the dependencies
    - RNNs have 'memory' that helps them store the states or information of previous inputs to generate the next output of the sequence.
    - RRN take the previous output or hidden states as inputs
    - In RNNs, their intermediate values (state) can store information about past inputs for a time that is not fixed a priori
    - RNN has cycles in its graph, and transmits information back into itself





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management (cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Example of ML model appropriate for an IBN engine (cont'd)
  - Recurrent Neural Network (RNN)
  - Advantages: capabilities to
    - handle sequence data; handle inputs of varying lengths; store historical information
  - Disadvantages :
    - The computation can be very slow
    - The basic network does not consider future inputs to make decisions
    - Vanishing or exploding gradient problem
      - The gradients used to compute the weight update may go to zero, preventing the network from learning new weights
        - the deeper the network, the more pronounced is this problem
      - One can avoid this by initializing the weights very carefully
        - If the weights are too small, the gradients shrink exponentially; too big then the gradients grow exponentially
      - Typical FNN can cope with these effects (low number of hidden layers)
    - In an RNN trained on long sequences the gradients can easily explode or vanish
    - RNNs have difficulty dealing with long-range dependencies





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Recurrent Neural Network (RNN)
  - RNN Architectures
    - Bidirectional recurrent neural networks (BRNN)
      - The inputs from future time steps are used to improve the accuracy of the network
      - (It is like having knowledge of the first and last words of a sentence to predict the middle words).
    - Gated Recurrent Units (GRU)
      - Networks designed to handle the vanishing gradient problem
      - They have a reset and update gate
      - These gates determine which information is to be retained for future predictions
      - Long Short-Term Memory (LSTM)
        - Designed to address the RNNs vanishing gradient problem
        - LSTM use three gates called input, output and forget gate
        - Similar to GRU, these gates determine which information to retain





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Recurrent Neural Network (RNN)- details
  - Simple RNN example: feedback loop shown in the gray rectangle can be unrolled in 3 time steps to produce the second network of the above figure



Source: J. Brownlee,

https://machinelearningmastery.com/an-introduction-to-recurrent-neural-networks-and-the-math-that-powers-them/





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Recurrent Neural Network (RNN)-details
  - Notations in the above figure
  - x<sub>t</sub> C R the input at time step t; in a simple case it is a scalar, but one can extend it to a d-dimensional feature vector
  - **y**<sub>t</sub> **C R** the output of the network at time step t; in this example there is one output, but one can produce multiple outputs
  - h<sub>t</sub> C R<sup>m-</sup> vector stores the values of the hidden units/states at time t ( this is called the current context; m is the number of hidden units; h<sub>0</sub> vector is initialized to zero
  - $w_x \in \mathbb{R}^m$  weights associated with inputs in recurrent layer
  - $w_h \in R^{m \times m}$  weights associated with hidden units in recurrent layer
  - $w_y \in \mathbb{R}^m$  weights associated with hidden to output units
  - $\mathbf{b}_{h} \in \mathbf{R}^{m}$  the bias associated with the recurrent layer
  - **b**<sub>v</sub> **C R** <sup>-</sup> the bias associated with the feedforward layer

Source: J. Brownlee,

https://machinelearningmastery.com/an-introduction-to-recurrent-neural-networks-and-the-math-that-powers-them/





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Recurrent Neural Network (RNN)-details

At every time step one can unfold the network for k time steps to get the output at time step k+1

### The unfolded network is very similar to the FFN

The rectangle in the unfolded network shows an operation taking place. Example, with an **activation function f**:

 $\mathbf{h}_{t+1} = \mathbf{f}(\mathbf{x}_{t}, \mathbf{h}_{t}, \mathbf{w}_{x}, \mathbf{w}_{h}, \mathbf{b}_{h}) = \mathbf{f}(\mathbf{w}_{x}\mathbf{x}_{t} + \mathbf{w}_{h}\mathbf{h} + \mathbf{b}_{h})$ 

The output y at time t is

 $y_t = f(h_t, w_y) = f(w_y \cdot h_t + b_y)$ ; the . is the dot product

- In the feedforward pass of a RNN, the network computes the values of the hidden units and the output after k time steps
- The weights associated with the network are shared temporally
- Each recurrent layer has two sets of weights; one for the input and the second one for the hidden unit.
- The last feedforward layer, computes the final output for the kth time step; it is just like an ordinary layer of a FNN





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - Recurrent Neural Network (RNN)- details
  - Activation functions examples (similar to FNN)
    - Sigmoid: σ(x) = 1 / (1 + exp(-x))
    - tanh: tanh(x) = 2σ(2x) 1
    - ReLU: Rectified Linear Unit. f(x) = max(0, x)







- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN

Example: Many

to Many

- Recurrent Neural Network (RNN)- details
- Types of RRNs



- One to Many
- Many to one
- Many to Many



Example: One to One



Source: J. Brownlee,

https://machinelearningmastery.com/an-introduction-to-recurrent-neural-networks-and-the-math-that-powers-them/





- 4.5 Example of Intent Based Management applied in 5G slicing
- Life Cycle Management(cont'd):
- Intelligent self- assurance and updating methods in IBN
  - **Recurrent Neural Network (RNN)**
  - Training the RNN
    - The backpropagation algorithm of an ANN is modified (to include the unfolding in time to train the weights of the network).
    - This algorithm: computing the gradient vector and is called back propagation in time (BPTT)

#### The pseudo-code for training

- The value of k can be selected by the user for training. The p<sub>t</sub> is the target value at time step t:
- 1. Repeat till stopping criterion is met:
  - 1. Set all h to zero.
  - 2. Repeat for t = 0 to n-k
    - 1. Forward propagate the network over the unfolded network for k time steps to compute all h and and y
    - 2. Compute the error as:  $e = y_{t+k} p_{t+k}$
    - 3. Backpropagate the error across the unfolded network and update the weights.





- **1.** Introduction
- 2. Network and services management issues and technologies
- 3. Intent-based Networking (IBN)- concepts and architecture
- 4. IBN approach in 5G slicing life-cycle management and orchestration- examples
- 5. Conclusions





- The Intent Based Networking (IBN) approach and platforms offer automation of management of the network slice lifecycle for single or multi-domain environment
- The user has to define its higher-level service requirements through GUI without dealing with the realisation of them
- IBN system translates the user requirements for a slice into policy configurations/ slice templates through domain policy configurators and forwards them to underlying orchestrators to automatically activate resources over the infrastructure.
- Core controllers (e.g., OSM) and RAN controllers can be used to responsible for handling the core network and RAN slices
- An IBN system can perform the 5G network slicing in single or multi-domain environment and E2E in a flexible way





- The IBN System eliminates the requirement of professional-level expertise for provisioning of the network service.
- Machine learning (ML) algorithms offer support for intent update engines, making IBN an intelligent platform that can update the run-time resources on demand
- The ML updating machine forecasts and predicts the network resources state on runtime, which helps the intent manger in deciding slice admission

### Challenges

- Refining the IBN architecture and systems in the context of multi-tenant, multi-domain, multi-operator, E2E
- Backward compatibility issues
- Initial admission control of new slice requests, to preserve the resources already allocated to other tenants
- Optimization issues related to ML algorithms, real-time monitoring and decisions





Thank you !Questions?





#### References

- 1. Sara Ayoubi, et.al., Machine Learning for Cognitive Network Management, IEEE Comm.Magazine , January 2018, pp.158-165
- 2. J. Strassner, "Policy-Based Network Management", Elsevier.", 2003
- 3. 5GPPP Network Management & Quality of Service Working Group, "Cognitive Network Management for 5G", 2017
- 4. A.Clemm et al., Intent-Based Networking Concepts and Definitions, draft-irtf-nmrg-ibn-concepts-definitions-07, March 2022
- E.Zeydan and Y.Turk "Recent Advances in Intent-Based Networking: A Survey", <u>https://ieeexplore.ieee.org/document/9128422</u>, 2020, Spanish MINECO grant TEC2017-88373-R (5G-REFINE) and Generalitat de Catalunya grant 2017 SGR 1195.
- 6. 5GPPP Architecture Working Group, View on 5G Architecture, Version 2.0, December 2017
- 7. V. P. Kafle, et. al., "Consideration on Automation of 5G Network slicing with Machine Learning", ITU Caleidoscope Santafe 2018
- 8. A.Galis, T. Francesc, C.E,.Rothenberg, S.Clayman, "Slicing 5G Networks: An Architectural Survey",2020, DOI: 10.1002/9781119471509.w5GRef095, <u>https://www.researchgate.net/publication/341437156</u>
- 9. 3GPP- TS 28.530 v 16.5.0 5G; Management and Orchestration; Concepts,Use Cases and Requirements, 2021, https://www.etsi.org/deliver/etsi\_ts/128500\_128599/128530/15.00.00\_60/t%s\_128530v150000p.pdf
- 10. 3GPP- TS 28.530 v 16.5.0 5G; Management and Orchestration; Concepts,Use Cases and Requirements, 2021, https://www.etsi.org/deliver/etsi ts/128500 128599/128530/15.00.00 60/t%s 128530v150000p.pdf
- 11. K. Abbas, T.A.Khan, M.Afaq, and W.C.Song, "Network Slice Lifecycle Management for 5G Mobile Networks: An Intent-Based Networking Approach", IEEE Access,2021
- 12. NX.Foukas, N.Nikaein, M.M. Kassem, M.K.Marina, K Kontovasilis ,"FlexRAN: A Flexible and Programmable Platform for Software-Defined RAN", European FP7 FLEX project (612050), <u>https://dl.acm.org/doi/10.1145/2999572.2999599</u>



### Intent Based Networking in 5G Slicing Life-cycle Management and Orchestration



#### References

- 13. https://opennetworking.org/cord/, https://opennetworking.org/m-cord/
- 14. "Open Source Mano" <u>https://osm-download.etsi.org/ftp/Documentation/201902-osm-scopewhite-%paper/#!02-osm-scope-and-functionality.md</u>
- 15. K.Abbas, M.Afaq, T.A.Khan, A.Rafiq and W-C Song, "Slicing the Core Network and Radio Access Network Domains through Intent-Based Networking for 5G Networks", Electronics 2020, 9, 1710; doi:10.3390/electronics9101710, <a href="http://www.mdpi.com/journal/electronics">www.mdpi.com/journal/electronics</a>
- 16. M.Wang, et al., "Machine Learning for Networking: Workflow, Advances and Opportunities", IEEE Network Vol.32, Issue 2, March-April 2018
- 17. J. Brownlee, <u>https://machinelearningmastery.com/an-introduction-to-recurrent-neural-networks-and-the-math-that-powers-them/</u>
- 18. A.Rafiq, A.Mehmood, T.A.Khan, K.Abbas, M.Afaq and W-C Song, "Intent-Based End-to-End Network Service Orchestration System for Multi-Platforms"Sustainability 2020, 12, 2782; <u>www.mdpi.com/journal/sustainability</u>
- 19. .Perezy, A.Zabala, and A.Banchsyz, "Alviu: An Intent-Based SD-WAN Orchestrator of Network Slices for Enterprise Networks", 2021 IEEE 7th International Conference on Network Softwarization (NetSoft),https://ieeexplore.ieee.org/abstract/document/9492534
- 20. <u>https://www.techtarget.com/searchcloudcomputing/definition/TOSCA-Topology-and-Orchestration-Specification-for-Cloud-Applications</u>
- 21.3GPP TS 28.530 V1.2.1 (2018-07), Management and orchestration of 5G networks; Concepts, use cases and requirements
- 22.3GPP TR 28.801 V15.1.0 (2018-01), Study on management and orchestration of network slicing for next generation network, (Release 15)
- 23. A.Galis, "Network Slicing Management Challenges", IRTF 102 NMRG, 2018
- 24. J. Ordonez-Lucena, P. Ameigeiras, D. Lopez, J.J. Ramos-Munoz, J. Lorca, J. Folgueira, Network "Slicing for 5G with SDN/NFV: Concepts, Architectures and Challenges", IEEE Communications Magazine, 2017, Citation information: DOI 10.1109/MCOM.2017.1600935



### Intent Based Networking in 5G Slicing Life-cycle Management and Orchestration



#### References

- 25. ETSI GS NFV 003 V1.2.1 (2014-12), Network Functions Virtualisation (NFV);Terminology for Main Concepts in NFV, http://www.etsi.org/deliver/etsi\_gs/NFV/001\_099/003/01.02.01\_60/gs\_NFV003v010201p.pdf
- 26. ETSI GR NFV-EVE 012 V3.1.1 (2017-12), Release 3; NFV Evolution and Ecosystem; Report on Network Slicing Support with ETSI NFV Architecture Framework
- 27. ETSI GR NFV-IFA 022 V3.1.1 (2018-04) Network Functions Virtualisation (NFV) Release 3; Management and Orchestration; Report on Management and connectivity for Multi-Site Services
- 28. http://www.5gamericas.org/files/3214/7975/0104/5G\_Americas\_Network\_Slicing\_11.21\_Final.pdf
- 29. X. Foukas, G. Patounas, A.Elmokashfi, and M.K. Marina, Network Slicing in 5G: Survey and Challenges, IEEE Communications Magazine, May 2017, pp.94-100
- 30. G. Nencioni et al., Orchestration and Control in Software-Defined 5G Networks: Research Challenges, Wiley, Wireless Communications and Mobile Computing Volume 2018, Article ID 6923867, pp. 1-19, https://doi.org/10.1155/2018/6923867https://www.hindawi.com/journals/wcmc/2018/6923867/
- 31. Ibrahim AFOLABI, et al., Towards 5G Network Slicing over Multiple-Domains, IEICE Trans. Commun., Vol.E100–B, No.11 Nov. 2017
- 32. A.Galis and K.Makhijani, Network Slicing Landscape: A holistic architectural approach, orchestration and management with applicbility in mobile and fixed networks and clouds, v1.0, Network Slicing Tutorial IEEE NetSoft 2018 Montreal 29th June2018.
- 33. Ibrahim Afolabi, Tarik Taleb, Konstantinos Samdanis, Adlen Ksentini, and Hannu Flinck, Network Slicing and Softwarization: A Survey on Principles, Enabling Technologies, and Solutions IEEE Comm. Surveys & Tutorials, Vol. 20, No. 3, 2018
- 34. T.Taleb, I.Afolabi, K.Samdanis, and F. Z.Yousaf, "On Multi-domain Network Slicing Orchestration Architecture & Federated Resource Control", <u>http://mosaic-lab.org/uploads/papers/3f772f2d-9e0f-4329-9298-aae4ef8ded65.pdf</u>
- 35. 5GPPP Network Management & Quality of Service Working Group, "Cognitive Network Management for 5G", 2017
- 36. "Machine Learning Tutorial for Beginners", https://www.guru99.com/machine-learning-tutorial.html
- 37. Machine Learning Basic Concepts https://courses.edx.org/asset-



### Intent Based Networking in 5G Slicing Life-cycle Management and Orchestration



#### References

- 38. <u>1:ColumbiaX+CSMM.101x+1T2017+type@asset+block@Al\_edx\_ml\_5.1intro.pdf</u>
- 39. Wei-Lun Chao, Machine Learning Tutorial, 2011, http://disp.ee.ntu.edu.tw/~pujols/Machine%20Learning%20Tutorial.pdf
- 40. Jessica Moysen and Lorenza Giupponi, "From 4G to 5G: Self-organized Network Management meets Machine Learning", arXiv:1707.09300v1 [cs.NI] 28 Jul 2017
- 41. S. Shalev-Shwartz and S.Ben-David, Understanding Machine Learning: From Theory to Algorithms 2014, Cambridge University Press
- 42. N.Dupuis, K.Lee, Introduction to Machine Learning, https://www.phusewiki.org/docs/Frankfut%20Connect%202018/ML/Papers/ML17-phuse-paper-ml01-19265.pdf
- 43. J,Quittek, Artificial Intelligence in Network Operations and Management, https://networking.ifip.org/2018/images/2018-IFIP-Networking/Keynote-III-J-Quittek-Slides.pdf
- 44. R.Johansson, Applied Machine Learning Lecture 1: Introduction, Univ. of Gothenburg, 2019, http://www.cse.chalmers.se/~richajo/dit866/lectures/I1/I1.pdf
- 45. M.Wang, et al., "Machine Learning for Networking: Workflow, Advances and Opportunities", IEEE Network Vol.32, Issue 2, March-April 2018
- 46. Gorka Vélez, et al., "Machine Learning for Autonomic Network Management in a Connected Cars Scenario", Communication Technologies for Vehicles, Springer 2016, pp 111-120, <u>https://link.springer.com/chapter/10.1007/978-3-319-38921-9\_12</u>
- 47. R.Li, Z.Zhao, et al., "Deep Reinforcement Learning for Resource Management in Network Slicing", arXiv:1805.06591v3 [cs.NI] 21 Nov 2018
- 48. A.Nowe and T.Brys, "A Gentle Introduction to Reinforcement Learning", Springer International Publishing Switzerland 2016, DOI: 10.1007/978-3-319-45856-4 2
- 49. D.Zeng, et al., "Resource Management at the Network Edge: A Deep Reinforcement Learning Approach", IEEE Network, May/June 2019,pp.26-33

### Machine Learning Techniques in Advanced Network and Services Management



#### List of Acronyms

| 5G CN | Core Network                     |
|-------|----------------------------------|
| 5G-AN | 5G Access Network                |
| 5GS   | 5G System                        |
| AF    | Application Function             |
| AI    | Artificial Intelligence          |
| CC    | Cloud Computing                  |
| CP    | Control Plane                    |
| CRAN  | Cloud based Radio Access Network |
| D2D   | Device to Device communication   |
| DL    | Deep Learning                    |
| DN    | Data Network                     |
| DNN   | Deep Neural Network              |
| DP    | Data Plane (User Plane UP)       |
| DT    | Decision Tree                    |
| IoT   | Internet of Things               |
| IBN   | Intent Based Networking          |
| IBS   | IBN System                       |
| k-NN  | k-Nearest Neighbours             |
| M&C   | Management and Control           |
| ML    | Machine Learning                 |
| NaaS  | Network as a Service             |
| NAI   | Network Access Identifier        |

### Machine Learning Techniques in Advanced Network and Services Management



#### List of Acronyms

| NF      | Network Function                                      |
|---------|---|
| NFV     | Network Function Virtualization                       |
| NN      | Neural Networks                                       |
| NSL     | Network Slice   |
| NSLI    | Network Slice Instance                                |
| NS      | Network Service                                       |
| NSLID   | Network Slice Instance Identifier                     |
| NSSAI   | Network Slice Selection Assistance Information        |
| NSSF    | Network Slice Selection Function                      |
| ONF     | Open Networking Foundation                            |
| PaaS    | Platform as a Service                                 |
| PCF     | Policy Control Function                               |
| QoE     | Quality of Experience                                 |
| RAN     | Radio Access Network                                  |
| RL      | Reinforcement Learning                                |
| SaaS    | Software as a Service                                 |
| SD      | Slice Differentiator                                  |
| SDN     | Software Defined Networking                           |
| SLA     | Service Level Agreement                               |
| SM      | Service Management                                    |
| SMF     | Session Management Function                           |
| SML     | Supervised Machine Learning                           |
| S-NSSAI | Single Network Slice Selection Assistance Information |
| SST     | Slice/Service Type                                    |
| SVM     | Support Vector Machine                                |
| UML     | Unsupervised Machine Learning                         |
| UPF     | User Plane Function                                   |
| VLAN    | Virtual Local Area Network                            |
| VM      | Virtual Machine                                       |