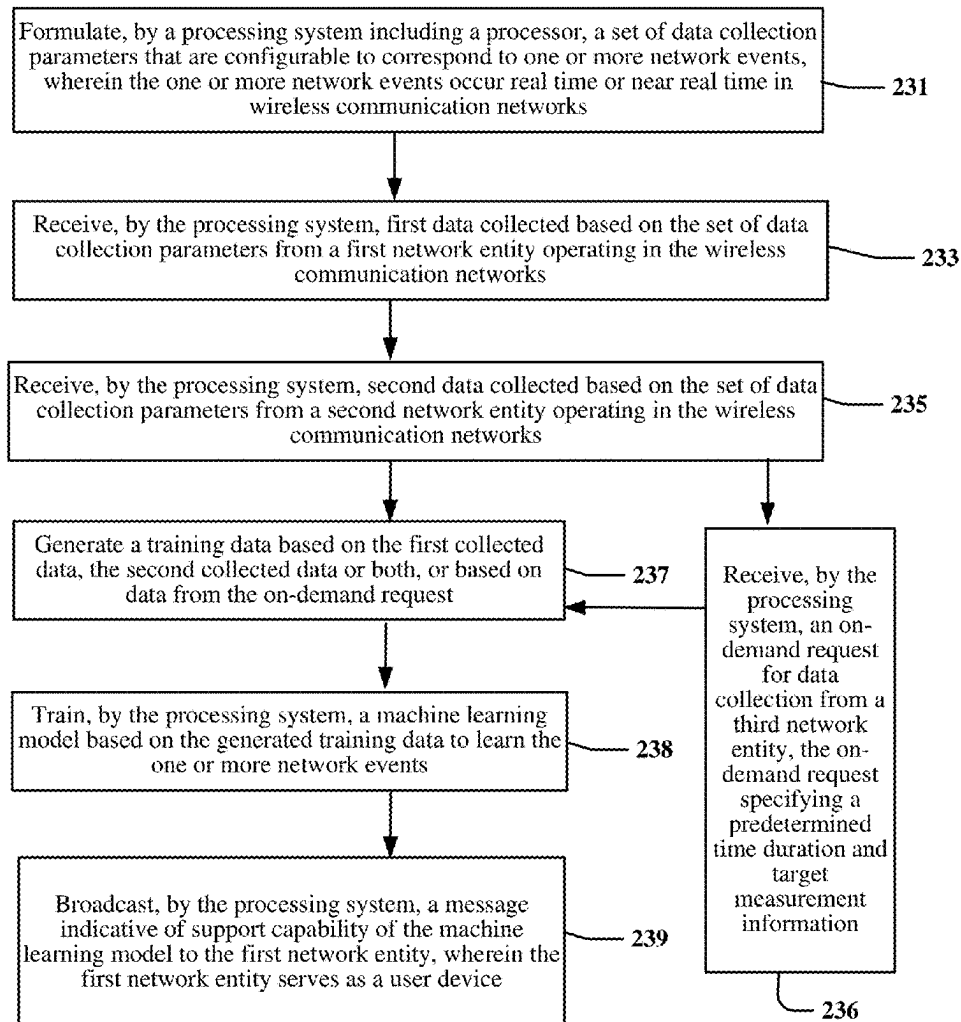


(19) **United States**(12) **Patent Application Publication**
Novlan(10) **Pub. No.: US 2025/0267193 A1**(43) **Pub. Date: Aug. 21, 2025**(54) **PROTOCOL AND SIGNALING
FRAMEWORK ENABLING MACHINE
LEARNING MODELS IN WIRELESS
COMMUNICATION NETWORKS**(52) **U.S. Cl.**
CPC **H04L 67/12** (2013.01); **H04W 28/0215**
(2013.01); **H04W 72/30** (2023.01)(71) Applicant: **AT&T Intellectual Property I, L.P.**,
Atlanta, GA (US)(57) **ABSTRACT**(72) Inventor: **Thomas Novlan**, Jonestown, TX (US)(73) Assignee: **AT&T Intellectual Property I, L.P.**,
Atlanta, GA (US)(21) Appl. No.: **18/443,587**(22) Filed: **Feb. 16, 2024****Publication Classification**(51) **Int. Cl.**
H04L 67/12 (2022.01)
H04W 28/02 (2009.01)
H04W 72/30 (2023.01)

Aspects of the subject disclosure may be directed to, for example, a method including determining a set of data collection parameters that are configurable to collect data indicative of network events occurring in real time or near real time in the wireless communication networks, receiving the collected data based on the set of data collection parameters from a group of network entities operating in the wireless communication networks, based on the received collected data, and generating training data for a machine learning model deployed in the wireless communication networks. Other embodiments are disclosed.

230

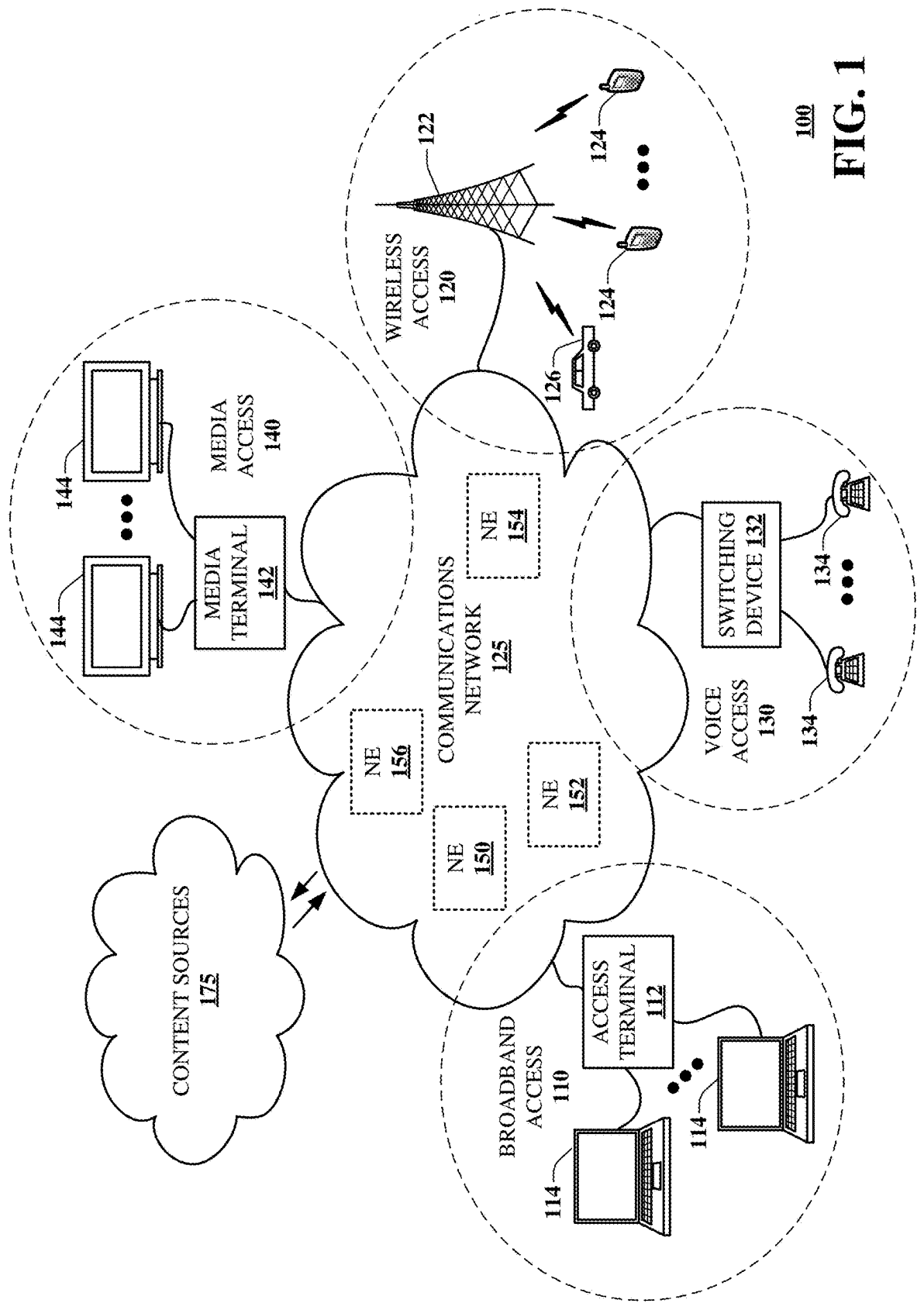


FIG. 1

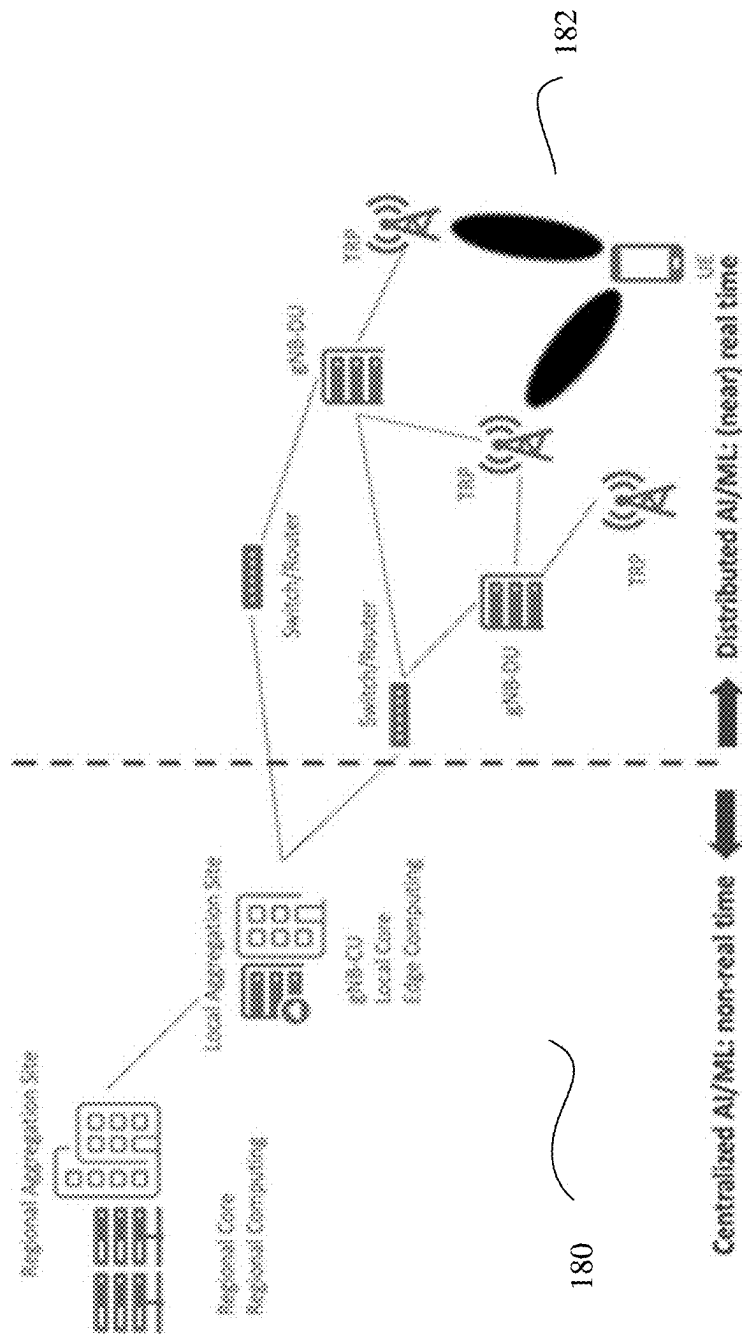
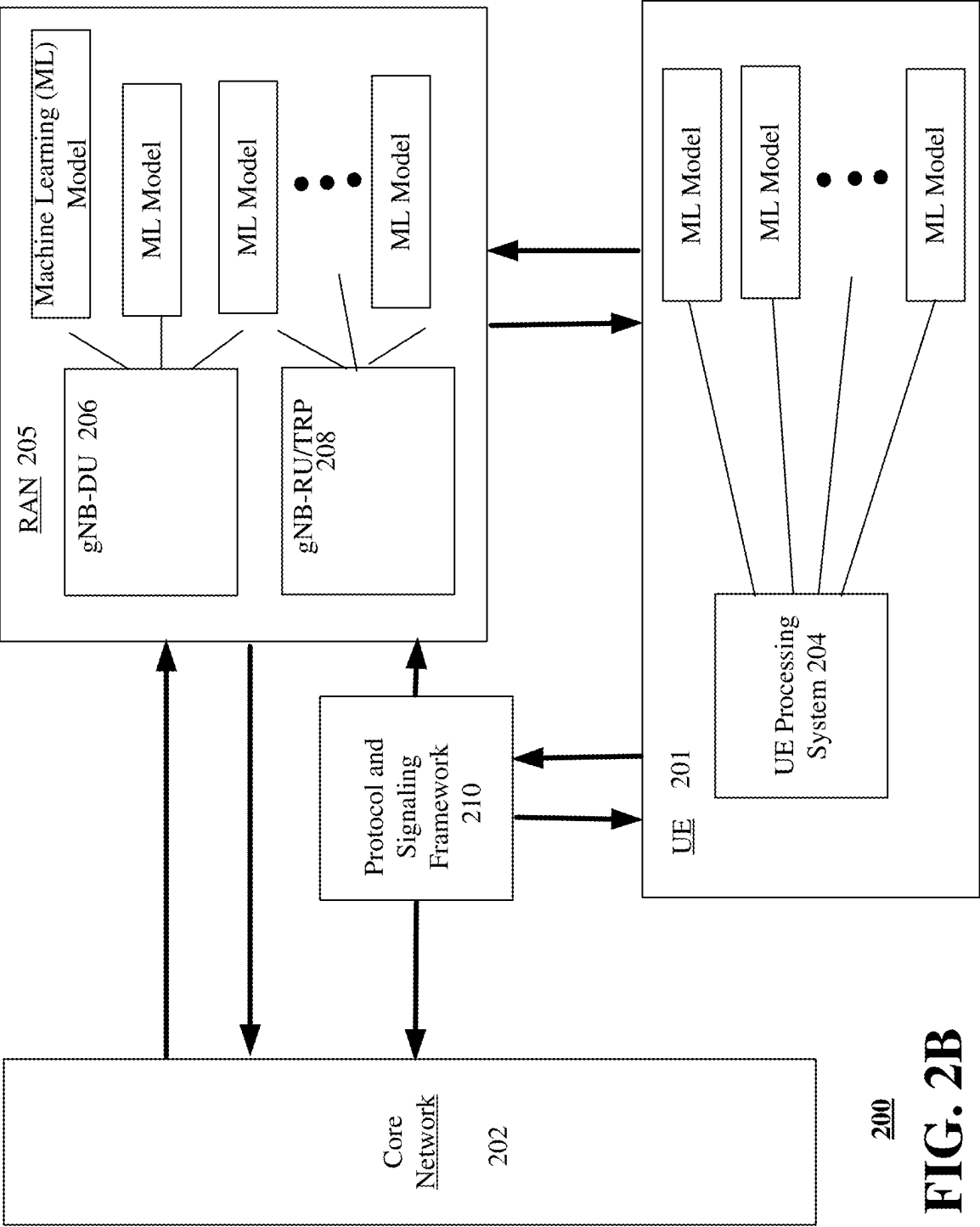
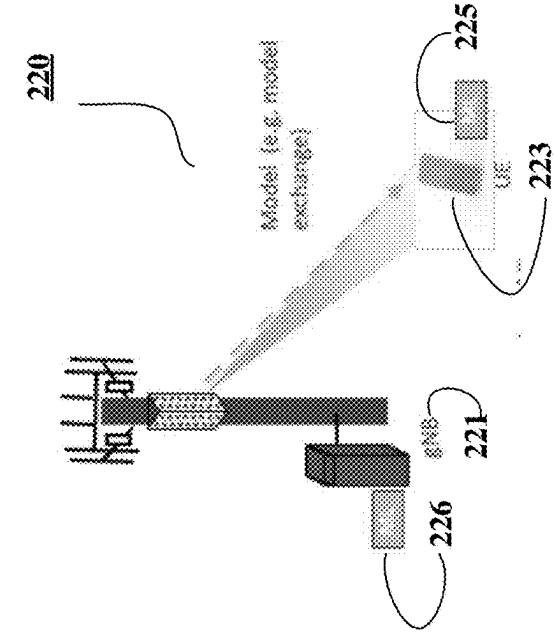
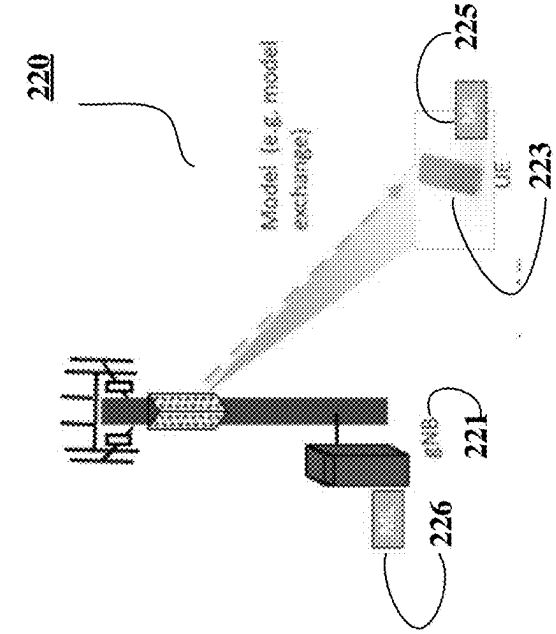
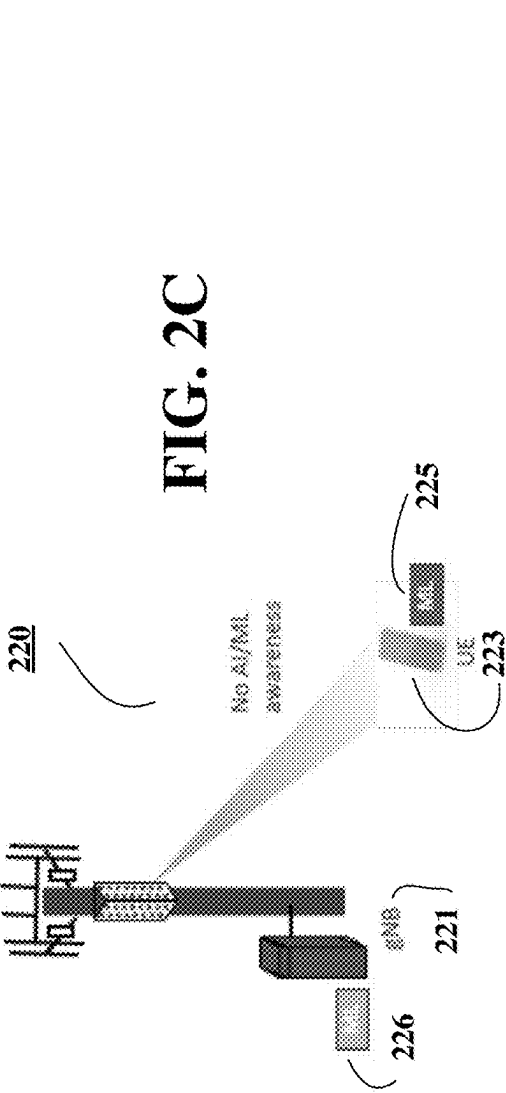


FIG. 2A



200

FIG. 2B



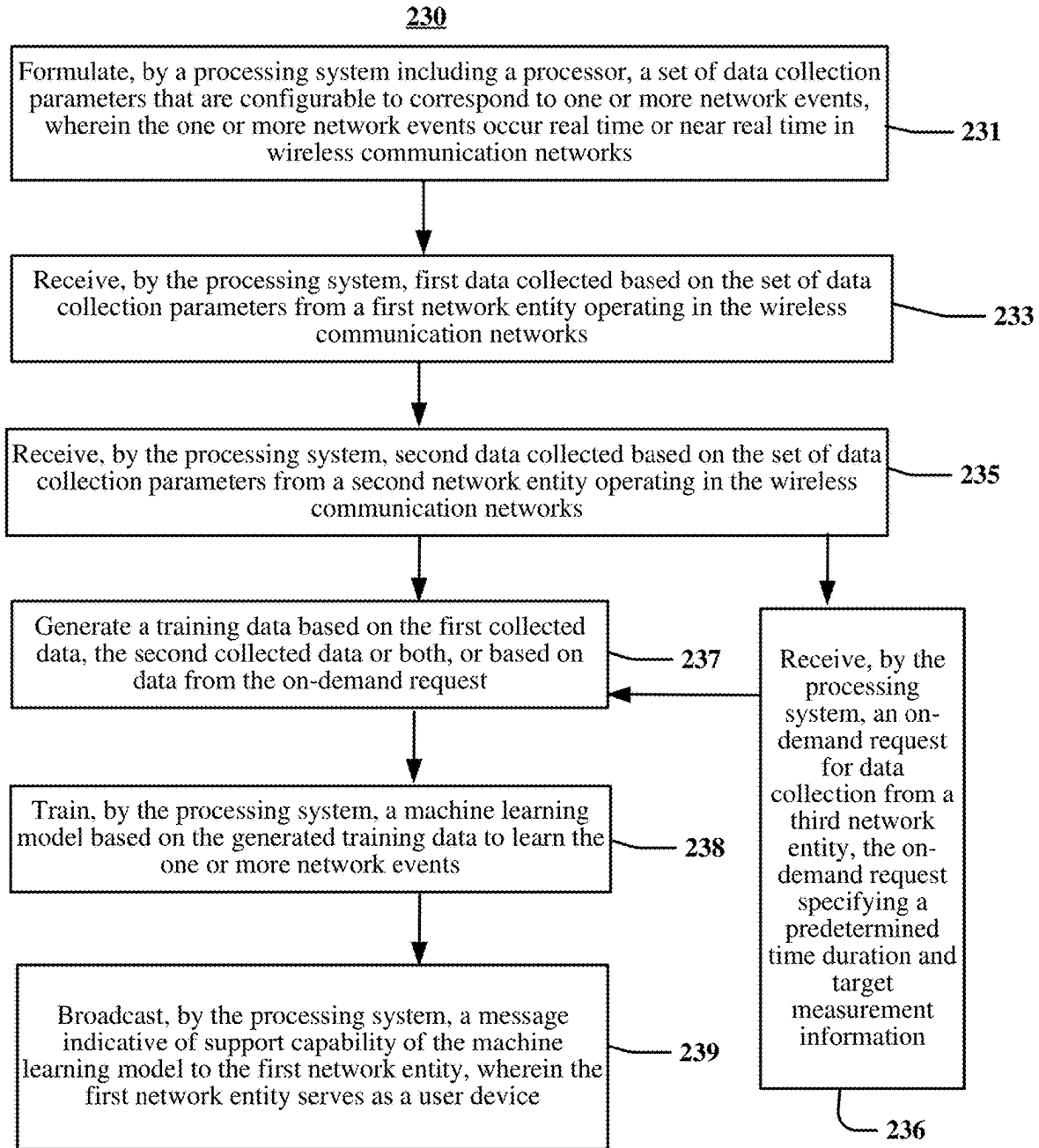
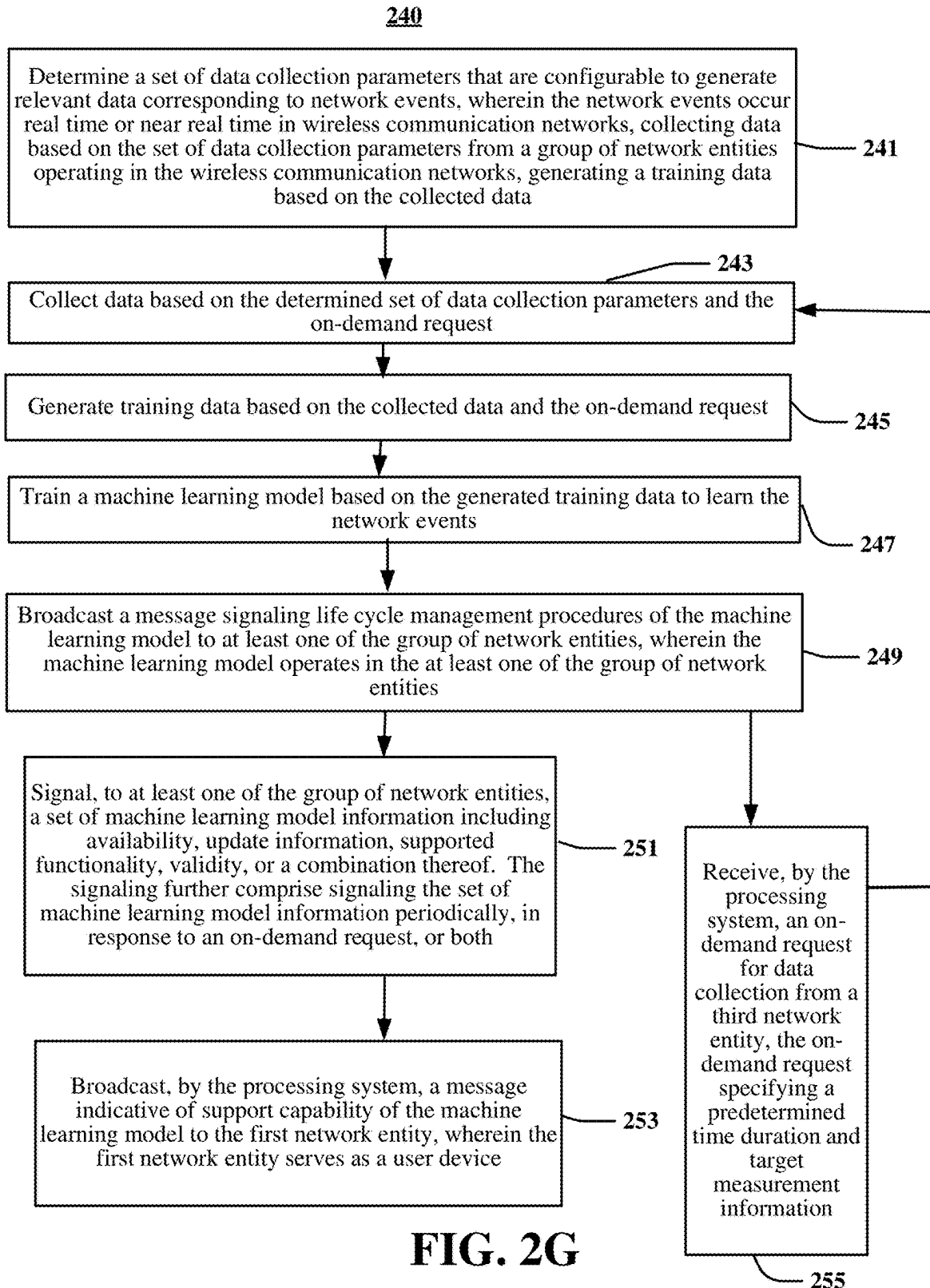


FIG. 2F



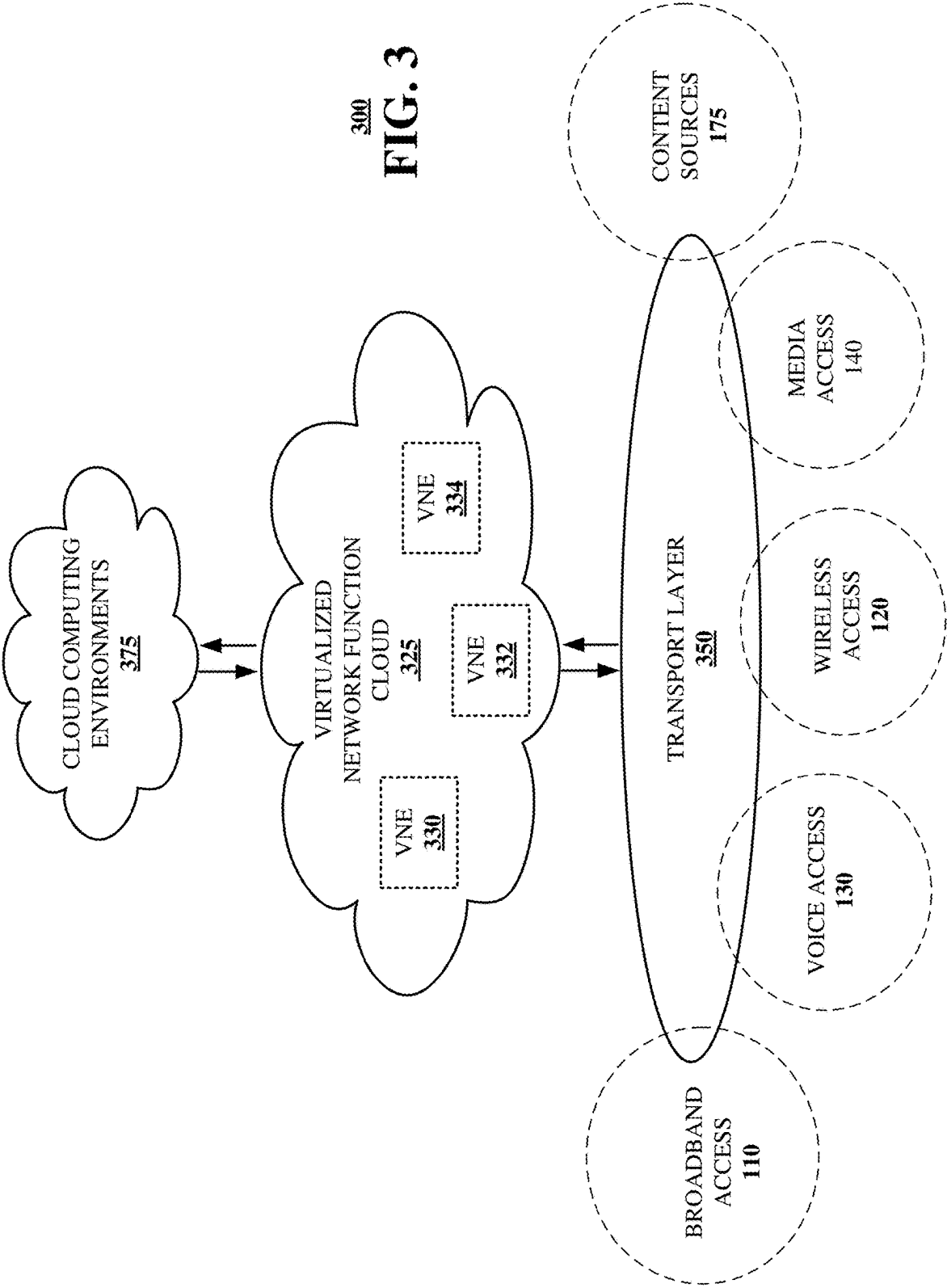


FIG. 3

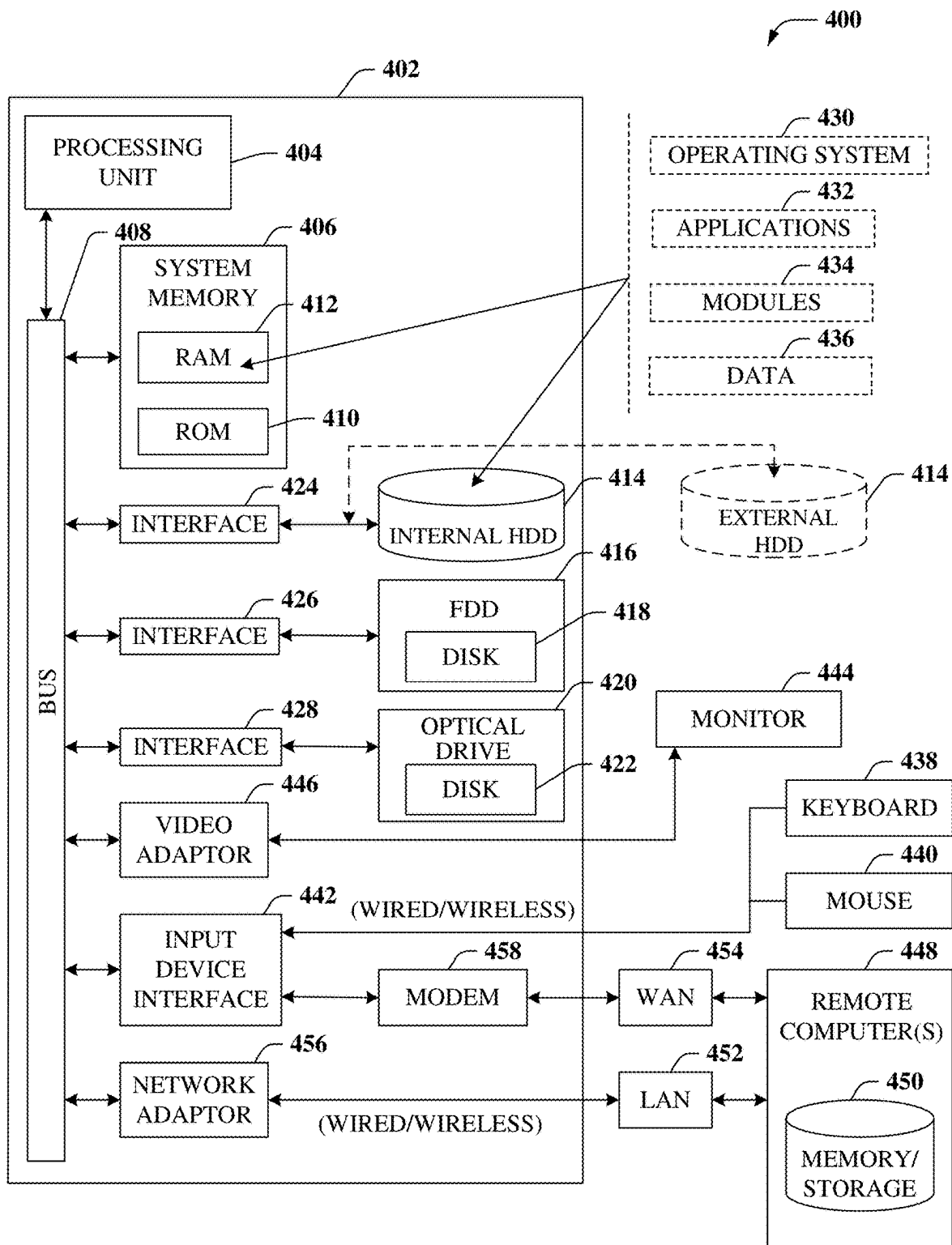


FIG. 4

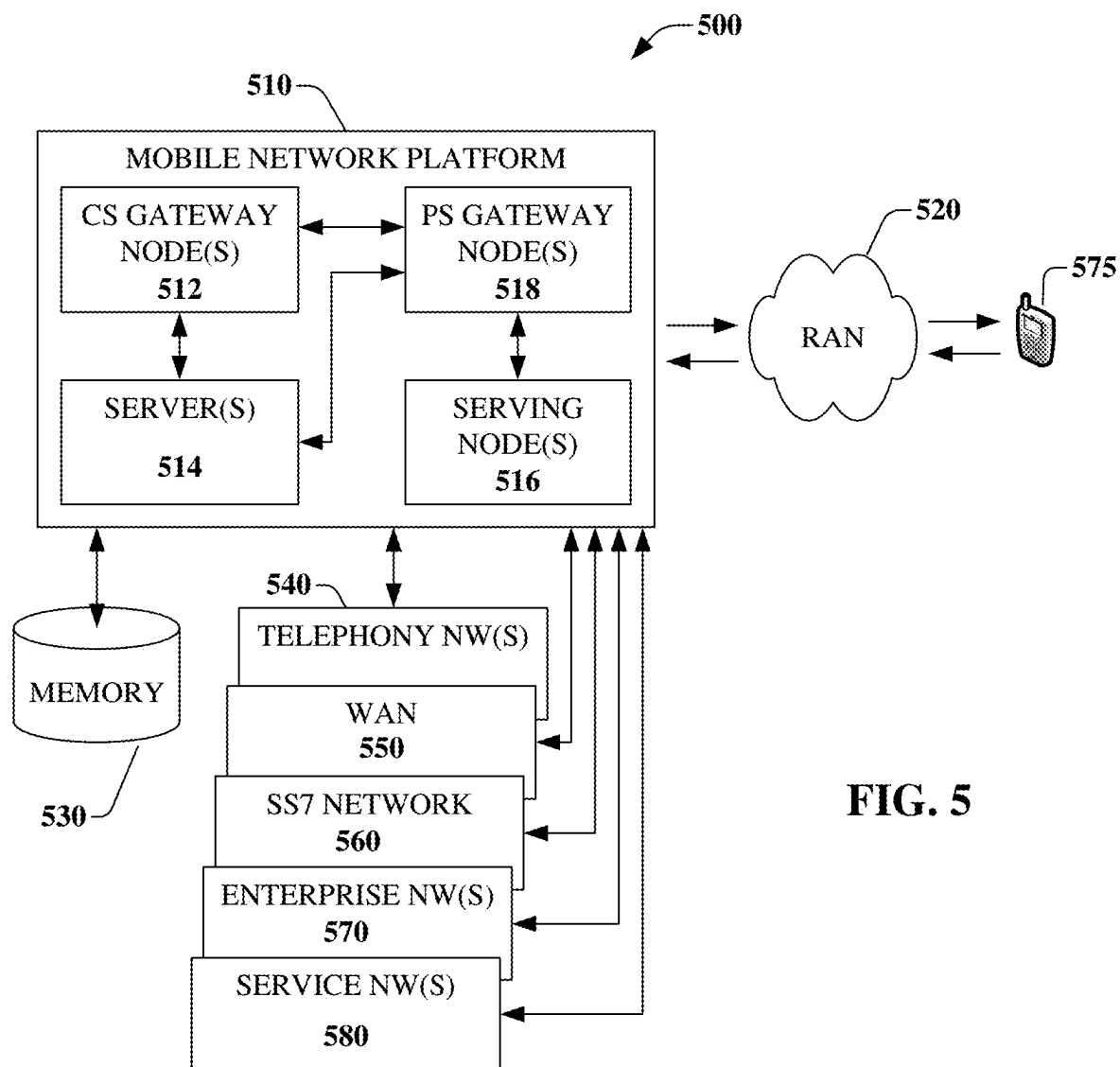
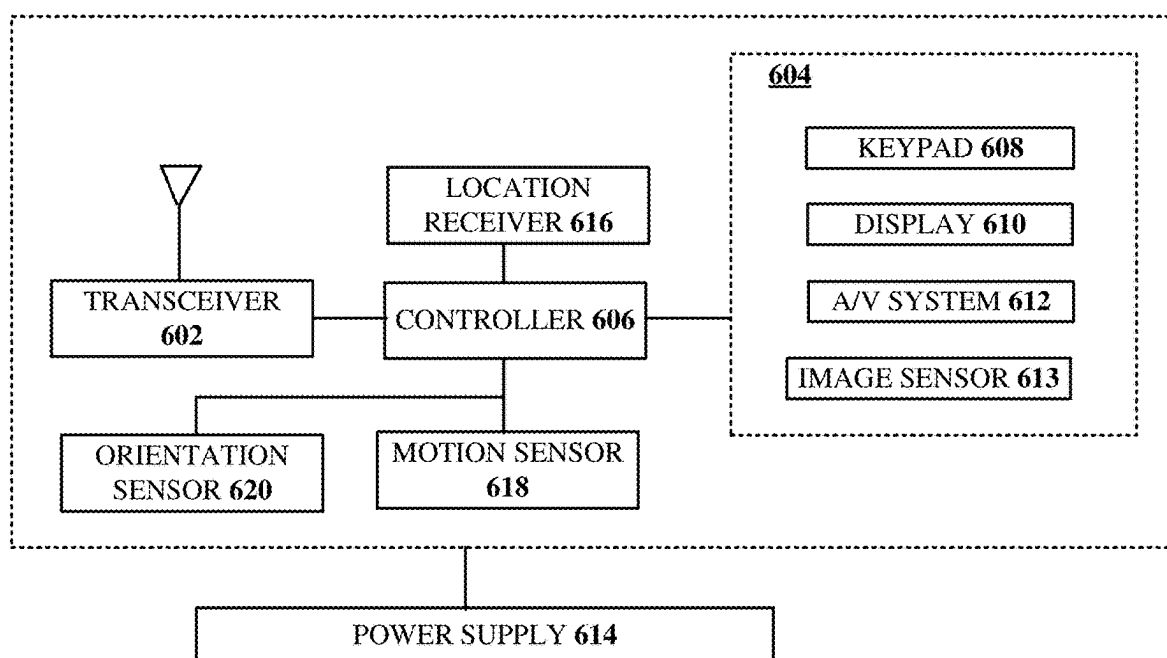


FIG. 5



600
FIG. 6

**PROTOCOL AND SIGNALING
FRAMEWORK ENABLING MACHINE
LEARNING MODELS IN WIRELESS
COMMUNICATION NETWORKS**

FIELD OF THE DISCLOSURE

[0001] The subject disclosure relates to a protocol and signaling framework enabling artificial intelligence or machine learning models in wireless communication networks including a 5G new radio air interface.

BACKGROUND

[0002] Existing air interface measurement collection frameworks such as minimization of drive tests (MDT) may not be optimized for artificial intelligence or machine learning applications. The MDT is a standardized mechanism (i.e., the Third Generation Partnership Project (3GPP), Release 10) which is designed to enable operators to use user equipment (UE) in a network to collect mobile network data. Thus, the MDT can reduce a need for traditional drive tests, which are associated with high costs and time commitments and provide a partial or limited view of a network as testing is limited to locations with a vehicle access. The MDT solution uses the UE to collect field measurements, including radio measurements and location information.

[0003] Conventional frameworks, however, may lack the flexibility to adapt to different scenarios and use cases and require either significant over-collection and subsequent filtering and/or processing at the UE or other data collection entities to extract relevant information. Additionally, the conventional frameworks may require frequent and dynamic parameter reconfigurations to ensure that relevant information is obtained when a scenario of interest is either recreated during testing or observed in the field for model retraining.

BRIEF DESCRIPTION OF THE DRAWINGS

[0004] Reference will now be made to the accompanying drawings, which are not necessarily drawn to scale, and wherein:

[0005] FIG. 1 is a block diagram illustrating an exemplary, non-limiting embodiment of a communications network in accordance with various aspects described herein.

[0006] FIG. 2A is a block diagram illustrating an example, non-limiting embodiment of a wireless communication network.

[0007] FIG. 2B is a block diagram illustrating an example, non-limiting embodiment of a system functioning within the communication network of FIG. 1 in accordance with various aspects described herein.

[0008] FIG. 2C is a block diagram illustrating an example, non-limiting embodiment of a network environment that facilitate a first collaboration level.

[0009] FIG. 2D is a block diagram illustrating an example, non-limiting embodiment of a network environment that facilitate a second collaboration level.

[0010] FIG. 2E is a block diagram illustrating an example, non-limiting embodiment of a network environment that facilitate a third collaboration level.

[0011] FIG. 2F depicts an illustrative embodiment of a method in accordance with various aspects described herein.

[0012] FIG. 2G depicts an illustrative embodiment of a method in accordance with various aspects described herein.

[0013] FIG. 3 is a block diagram illustrating an example, non-limiting embodiment of a virtualized communication network in accordance with various aspects described herein.

[0014] FIG. 4 is a block diagram of an example, non-limiting embodiment of a computing environment in accordance with various aspects described herein.

[0015] FIG. 5 is a block diagram of an example, non-limiting embodiment of a mobile network platform in accordance with various aspects described herein.

[0016] FIG. 6 is a block diagram of an example, non-limiting embodiment of a communication device in accordance with various aspects described herein.

DETAILED DESCRIPTION

[0017] The subject disclosure describes, among other things, illustrative embodiments for a protocol and signaling framework enabling artificial intelligence or machine learning models in wireless communication networks. Other embodiments are described in the subject disclosure.

[0018] One or more aspects of the subject disclosure is directed to a method including, formulating, by a processing system including a processor, a set of data collection parameters that are configurable to collect data corresponding to one or more network events, wherein the one or more network events occur in real time or near real time in wireless communication networks, receiving, by the processing system, first data collected based on the set of data collection parameters from a first network entity operating in the wireless communication networks, receiving, by the processing system, second data collected based on the set of data collection parameters from a second network entity operating in the wireless communication networks, generating a training data based on the first collected data, the second collected data or both, and training, by the processing system, a machine learning model based on the generated training data to learn the one or more network events.

[0019] One or more aspects of the subject disclosure is directed to a device including a processing system including a processor, and a memory that stores executable instructions that, when executed by the processing system, facilitate performance of operations. The operations include determining a set of data collection parameters that are configurable to generate relevant data corresponding to network events, wherein the network events occur in real time or near real time in wireless communication networks, collecting data based on the set of data collection parameters from a group of network entities operating in the wireless communication networks, generating a training data based on the collected data, and training a machine learning model based on the generated training data to learn the network events.

[0020] One or more aspects of the subject disclosure is directed to a non-transitory machine-readable medium, comprising executable instructions that, when executed by a processing system including a processor, facilitate performance of operations. The operations include determining a set of data collection parameters that are configurable to collect data indicative of network events occurring in real time or near real time in the wireless communication networks, receiving the collected data based on the set of data collection parameters from a group of network entities operating in the wireless communication networks, based on the received collected data, generating training data for a

machine learning model deployed in the wireless communication networks, and broadcasting a message signaling life cycle management procedures of the machine learning model to at least one of the group of network entities, wherein the machine learning model is deployed in the at least one of the group of network entities.

[0021] Referring now to FIG. 1, a block diagram is shown illustrating an example, non-limiting embodiment of a system **100** in accordance with various aspects described herein. For example, system **100** can facilitate in whole or in part a protocol and signaling framework enabling ML models in wireless communication networks. In particular, a communications network **125** is presented for providing broadband access **110** to a plurality of data terminals **114** via access terminal **112**, wireless access **120** to a plurality of mobile devices **124** and vehicle **126** via base station or access point **122**, voice access **130** to a plurality of telephony devices **134**, via switching device **132** and/or media access **140** to a plurality of audio/video display devices **144** via media terminal **142**. In addition, communication network **125** is coupled to one or more content sources **175** of audio, video, graphics, text and/or other media. While broadband access **110**, wireless access **120**, voice access **130** and media access **140** are shown separately, one or more of these forms of access can be combined to provide multiple access services to a single client device (e.g., mobile devices **124** can receive media content via media terminal **142**, data terminal **114** can be provided voice access via switching device **132**, and so on).

[0022] The communications network **125** includes a plurality of network elements (NE) **150**, **152**, **154**, **156**, etc. for facilitating the broadband access **110**, wireless access **120**, voice access **130**, media access **140** and/or the distribution of content from content sources **175**. The communications network **125** can include a circuit switched or packet switched network, a voice over Internet protocol (VOIP) network, Internet protocol (IP) network, a cable network, a passive or active optical network, a 4G, 5G, or higher generation wireless access network, WIMAX network, UltraWideband network, personal area network or other wireless access network, a broadcast satellite network and/or other communications network.

[0023] In various embodiments, the access terminal **112** can include a digital subscriber line access multiplexer (DSLAM), cable modem termination system (CMTS), optical line terminal (OLT) and/or other access terminal. The data terminals **114** can include personal computers, laptop computers, netbook computers, tablets or other computing devices along with digital subscriber line (DSL) modems, data over coax service interface specification (DOCSIS) modems or other cable modems, a wireless modem such as a 4G, 5G, or higher generation modem, an optical modem and/or other access devices.

[0024] In various embodiments, the base station or access point **122** can include a 4G, 5G, or higher generation base station, an access point that operates via an 802.11 standard such as 802.11n, 802.11ac or other wireless access terminal. The mobile devices **124** can include mobile phones, e-readers, tablets, phablets, wireless modems, and/or other mobile computing devices.

[0025] In various embodiments, the switching device **132** can include a private branch exchange or central office switch, a media services gateway, VoIP gateway or other gateway device and/or other switching device. The tele-

phony devices **134** can include traditional telephones (with or without a terminal adapter), VOIP telephones and/or other telephony devices.

[0026] In various embodiments, the media terminal **142** can include a cable head-end or other TV head-end, a satellite receiver, gateway or other media terminal **142**. The display devices **144** can include televisions with or without a set top box, personal computers and/or other display devices.

[0027] In various embodiments, the content sources **175** include broadcast television and radio sources, video on demand platforms and streaming video and audio services platforms, one or more content data networks, data servers, web servers and other content servers, and/or other sources of media.

[0028] In various embodiments, the communications network **125** can include wired, optical and/or wireless links and the network elements **150**, **152**, **154**, **156**, etc. can include service switching points, signal transfer points, service control points, network gateways, media distribution hubs, servers, firewalls, routers, edge devices, switches and other network nodes for routing and controlling communications traffic over wired, optical and wireless links as part of the Internet and other public networks as well as one or more private networks, for managing subscriber access, for billing and network management and for supporting other network functions.

[0029] FIG. 2A is a block diagram illustrating an example, non-limiting embodiment of artificial intelligence (AI) models or machine learning (ML) models (hereinafter, collectively referred to as machine learning (ML) models) deployed in wireless communication networks. As depicted in FIG. 2A, the 5G NR overall architecture, as defined in the 3GPP TS 38.300 specification, include a 5G core network **180** and a next generation radio access network (NG-RAN) **182**.

[0030] The 5G industry trends which enable network virtualization and deployment of low-latency and/or high bandwidth services are also making feasible and scalable applications of power artificial intelligence (AI) tools such as machine learning (ML) algorithms to 5G networks (hereinafter, collectively, ML). These ML algorithms rely on historical data for deriving system models and training as well as real-time or near-real-time data collection to adapt to different network conditions. Furthermore, a variety of use cases can be supported by ML techniques including CSI feedback optimization, beam management, and positioning. Different use cases can have vastly different requirements in terms of the impact on network nodes or functionalities. This implies that the appropriate implementation of different ML techniques may involve multiple interfaces, signaling procedures, and processing requirements (including requirements on data aggregation or co-location with different nodes/functions). As shown in FIG. 2A, at the core network and gNB-CU **180**, the disaggregated, heterogeneous and complex network architecture supports ML algorithms which provide non-real time decisions based on centralized data sets. However, at the gNB-DU, a transmission point (e.g., gNB-RU), and user equipment (UE), ML algorithms embedded in functionality of the RAN enable real-time decisions or near real-time decisions with distributed data.

[0031] As part of Release 18 Study Item, the 3GPP started researching a framework for ML models for use in an air-interface corresponding to target use cases considering

aspects such as performance, complexity, and potential specification. The present disclosure is directed to a protocol and signaling framework for ML models which is applied to 5G New Radio (NR) air interfaces, specifically impacting on a Radio Access Network (RAN) architecture and lifecycle management (LCM) of a ML model. The RAN architecture includes an Open Radio Access Network (O-RAN) that allows interoperation and coordination between cellular network elements provided by different industry players.

[0032] FIG. 2B is a block diagram illustrating an example, non-limiting embodiment of a system **200** functioning within the communication network of FIG. **1** in accordance with various aspects described herein. The system **200** includes a user equipment (UE) **201**, a core network **202**, a RAN **204** and a protocol and signaling framework **210**. At the RAN **204**, a plurality of ML models are being deployed or to be deployed. At the UE **201**, one or more ML models are deployed or to be deployed. In other embodiments, the ML models can be deployed at the UE **201** or the RAN **205**. In some embodiments, one or multiple ML models may be deployed. In some embodiments, one ML model can implement one or more network functions, multiple ML models can implement one network function, etc. Additionally or alternatively, one or more ML models can be used for a single use case, or multiple use cases. In some embodiments, one or more ML models, upon deployment, can be activated, deactivated or subject to a fallback as a part of life cycle management (LCM) of such ML models. Additionally, or alternatively, one or more ML models, upon deployment, can be upgraded, transferred, trained, retrained, replaced, etc. as a part of the LCM of such ML models.

[0033] In various embodiments, the protocol and signaling framework **210** includes a processing system including a processor and a memory that stores executable instructions that, when executed by the processing system, facilitate performance of operations of data collection for offline and online training, supporting and facilitating broadcast and on-demand ML model delivery and update of information. The protocol and signaling framework **210** further facilitates and control the LCM of ML models. The protocol and signaling framework **210** may have full visibility into the LCM of ML models at the UE, the gNB, or both. The protocol and signaling framework **210** may control configurations of predetermined thresholds to activate any associated explicit trigger, event or condition-based triggers, etc. in order to perform the LCM of ML models.

[0034] Considering the different architecture options as shown in FIG. **2A** (e.g., centralized v. distributed, non-real time v. near real time, etc.), the protocol and signaling framework **210** operates as a ML model framework applied to the 5G NR air interface which can support configuration, data collection, and collaboration between network entities and/or user devices required by a given ML use cases or approaches. In order to support data collection for ML model training, extensive data from UEs and/or gNBs may be transferred either within the RAN **205** or to a dedicated location for processing and ML model training (e.g., an Operation, Administration, and Maintenance (OAM) system) in the core network **202**. Given that ML models will continue to be refined over time, different approaches may require different types of information with different granularities and/or data collection intervals (even for the same use case). As a result, the protocol and signaling framework **210** facilitates a new data collection framework which is

optimized for current use cases of ML models and can be extended to future use cases. The protocol and signaling framework **210** supports and facilitates ML use cases by serving as a data collection framework for offline and online training of ML models as well as ML model monitoring (e.g., for model LCM including retraining). Additionally, the protocol and signaling framework **210** may be optimized for current use cases of ML models and can be extended to future use cases.

[0035] In various embodiments, the protocol and signaling framework **210** may enable support of broadcast and on-demand ML model delivery and update information, enabling the network to have full visibility into the life cycle management (LCM) of ML models at the UE/gNB and control of configuring any associated explicit triggers and event or condition-based triggers to modify, update or fallback the LCM of ML models (e.g., retraining of ML models).

[0036] In various embodiments, the protocol and signaling framework **210** formulates a set of data collection parameters that are configurable to correspond to one or more network events. The one or more network events occur real time or near real time in wireless communication networks. As one example, the set of data collection parameters specify a predetermined validity time duration. As another example, the set of data collection parameters is configured to maintain a predetermined time interval for periodically collected information until the predetermined time interval is overridden by an updated time interval. As further another example, formulating the set of data collection parameters includes configuring the set of data collection parameters to specify a data type, a data format or both.

[0037] Depending on the deployment characteristics, traffic load, mobility status of UEs, and the gNB ML capabilities, the granularity of the predicted resource status information may be highly variable. For example, in a low load scenario, a predicted channel state or predicted resource status for a given link or cell may be a constant value for a long period of time or at least with low variance under a pre-defined high load threshold. In contrast, in a high load or scenario with lots of dynamic mobility (e.g., during a commuter rush hour), a predicted channel and/or resource status may have a high variance with frequent changes during a short period of time. As a result, a single value corresponding to a fixed time duration may not be feasible to support a large variety of practical scenarios where a load balancing or mobility optimization procedure is performed. Instead, data collected over higher layer signaling (e.g., MAC, RRC, or Xn interfaces) should be associated with validity time duration and/or time interval until a next prediction becomes available.

[0038] In various embodiments, data and/or measurement collection may be requested using an on-demand procedure, where a requesting node may request a time duration for the data and measurement information based on what is useful for a given ML model training and/or inference algorithm associated with the requesting node.

[0039] In various embodiments, network-UE collaboration levels as to ML models are defined as shown in FIGS. **2C** through **2E**, which enable different aspects of the signaling and protocol impacts on ML model approaches applied to air interface use cases. FIG. **2C** is a block diagram illustrating an example, non-limiting embodiment of a network environment that facilitate a first collaboration level.

FIG. 2D is a block diagram illustrating an example, non-limiting embodiment of a network environment that facilitate a second collaboration level. FIG. 2E is a block diagram illustrating an example, non-limiting embodiment of a network environment that facilitate a third collaboration level. FIGS. 2C through 2E illustrate one base station 221 and one user equipment 223 for convenience of description, but the present disclosure is not limited thereto. As depicted in FIGS. 2C through 2E, one or more ML models 225, 226 are deployed in wireless communication networks 220. The wireless communication networks 220 include air interfaces that can be augmented by enabling support for the one or more ML models 225, 226.

[0040] In various embodiments, the collaboration levels as to ML models between the network and the UE can be defined between the UE 223 and the network 221 as follows:

[0041] The First Collaboration Level (FIG. 2C), where no collaboration is assumed between the network 221 and the UE 223;

[0042] The Second Collaboration Level (FIG. 2D), where the collaboration is signaling based with no ML model transfer; and

[0043] The Third Collaboration Level (FIG. 2E), where the collaboration is signaling based with a ML model transfer.

[0044] In some embodiments, differences between the first collaboration level and the second collaboration level and between the first collaboration level and the third collaboration level rest on whether an explicit LCM functionality is supported between the gNB and UE. As described above, the third collaboration level supports an ML model transfer, whereas the second collaboration level does not involve a ML model transfer. In other embodiments, the difference between the second collaboration level and the third collaboration level is whether a model delivery is transparent or explicit via 3GPP defined signaling. The third collaboration level supports the explicit model transfer, whereas the second collaboration level may not support a ML model transfer and the ML model transfer may be transparent to the network or the UE.

[0045] In some embodiment, in case the model delivery is provided explicitly by 3GPP signaling (e.g., use cases which support the third collaboration level), the network should inform UEs whether a model is available for delivery, whether it has been updated if provided previously, etc. Depending on use cases and how frequently training information is updated, this model delivery may be very infrequent, or may be updated based network and user conditions which can dynamically change. In some embodiments, the dynamically changing user conditions may necessitate an on-demand mechanism. For example, ML use case/approaches which support the third collaboration level for model delivery can support both periodic as well as on-demand signaling or broadcasting of a message. As one example, periodic signaling or broadcasting can be System Information Block (SIB)-based. As another example, the on-demand signaling or broadcasting can be requested from UE or via dedicated signaling. Periodic signaling or broadcasting transmit information such as model availability, update information, and other assistance information which indicates supported model functionality and validity of the model within the coverage area or a specific gNB-UE link.

[0046] In either the second collaboration level and the third collaboration level, the protocol and signaling frame-

work 210 may have a full ability to manage the activation/deactivation or model selection/fallback if supported for a given use case. While it is expected that applying a machine learning model at either the UE or gNB may result in improved performance, if the network or environmental conditions change quickly or drift beyond the training data of the model sufficiently, this may not be guaranteed. As a result, it is critical for the protocol and signaling framework 210 to have full visibility into the LCM of ML models at the UE, gNB or both, and control of configuring any associated explicit trigger as well as event or condition-based triggers to perform the LCM of ML models. The event or condition-based triggers of the LCM procedures can be implemented either by pre-configuration thresholds or network-configurable thresholds which are provided to UEs using a broadcast (e.g., SIB) or a dedicated (e.g., RRC) signaling.

[0047] In some embodiments, with respect to ML use cases or approaches which support the second and the third collaboration levels for LCM procedures, the protocol and signaling framework 210 support mechanisms at the gNB and UE for a model selection, an activation, a deactivation, a switching, and a fallback. In other embodiments, the protocol and signaling framework 210 supports network triggered LCM procedures using a broadcast (e.g., System Information Block (SIB)) or a dedicated (e.g., RRC) signaling. SIBs are broadcast messages from base stations (e.g., an gNB) to UEs that contain essential information about the network.

[0048] FIG. 2F depicts an illustrative embodiment of a method 230 in accordance with various aspects described herein. In various embodiments, the method 230 includes formulating, by a processing system including a processor, a set of data collection parameters that are configurable to correspond to one or more network events (Step 231). The one or more network events occur real time or near real time in wireless communication networks (Step 231). In some embodiments, the formulating the set of data collection parameters includes configuring the set of data collection parameters to specify a predetermined validity time duration. In other embodiments, the formulating the set of data collection parameters includes configuring the set of data collection parameters to maintain a predetermined time interval for periodically collected information until the predetermined time interval is overridden by an updated time interval. Additionally or alternatively, the formulating the set of data collection parameters includes configuring the set of data collection parameters to specify the predetermined time duration specified in the on-demand request and the target measurement information. In some embodiments, the formulating the set of data collection parameters includes configuring the set of data collection parameters to specify a data type, a data format or both.

[0049] The method 230 further includes receiving, by the processing system, first data collected based on the set of data collection parameters from a first network entity operating in the wireless communication networks (Step 233) and receiving, by the processing system, second data collected based on the set of data collection parameters from a second network entity operating in the wireless communication networks (Step 235). The method 230 further includes generating a training data based on the first collected data, the second collected data or both (Step 237), and training, by the processing system, a machine learning model based on the generated training data to learn the one or more network

events (Step 238). In some embodiments, the method further includes receiving, by the processing system, an on-demand request for data collection from a third network entity, the on-demand request specifying a predetermined time duration and target measurement information (Step 236). Additionally, or alternatively, the method 230 further includes generating the training data based on the on-demand request (Step 237).

[0050] In various embodiments, the method 230 further includes broadcasting, by the processing system, a message indicative of support capability of the machine learning model to the first network entity, wherein the first network entity serves as a user device (Step 239). In some embodiments, the method 230 further includes broadcasting, by the processing system, a message indicative of validity of the machine learning model to the first network entity, wherein the first network entity serves as a user device. The method 230 further includes broadcasting, periodically, on an on-demand basis, or both, a message indicative of information relevant to the machine learning model to the first network entity, wherein the first network entity serves as a user device.

[0051] FIG. 2G depicts an illustrative embodiment of a method in accordance with various aspects described herein. The method 240 includes determining a set of data collection parameters that are configurable to generate relevant data corresponding to network events, wherein the network events occur real time or near real time in wireless communication networks (Step 241), collecting data based on the set of data collection parameters from a group of network entities operating in the wireless communication networks (Step 243), and generating a training data based on the collected data (Step 245). The method 240 further includes collecting data based on the determined set of data collection parameters (Step 243). The method 240 further includes generating training data based on the collected data and the on-demand request (Step 245). The method 240 further includes training a machine learning model based on the generated training data to learn the network events (Step 247).

[0052] The method 240 further includes broadcasting a message signaling life cycle management procedures of the machine learning model to at least one of the group of network entities, wherein the machine learning model operates in the at least one of the group of network entities (Step 249). Additionally, the method 240 further includes performing, as the life cycle management procedures, an activation or a deactivation of the machine learning model based on a predetermined trigger. In some embodiments, the broadcasting the message signaling the life cycle management procedures further comprises facilitating an activation or deactivation, a switching, a fallback of the machine learning model or a combination thereof, based on a predetermined trigger. The method 240 further includes providing a condition-based life cycle management procedure to a user equipment by using a broadcast or a dedicated signaling. In some embodiments, the method 240 further includes signaling, to at least one of the group of network entities, a set of machine learning model information including availability, update information, supported functionality, validity, or a combination thereof (Step 251). The signaling further comprises signaling the set of machine learning model information periodically, in response to an on-demand request, or both (Step 251).

[0053] In some embodiments, the method 240 further includes receiving an on-demand request for data collection, the on-demand request specifying a predetermined time duration and target measurement (Step 255), and collecting another data based on the predetermined time duration and the target measurement specified in the on-demand request (Step 243).

[0054] While for purposes of simplicity of explanation, the respective processes are shown and described as a series of blocks in FIGS. 2F and 2G, it is to be understood and appreciated that the claimed subject matter is not limited by the order of the blocks, as some blocks may occur in different orders and/or concurrently with other blocks from what is depicted and described herein. Moreover, not all illustrated blocks may be required to implement the methods described herein.

[0055] Referring now to FIG. 3, a block diagram 300 is shown illustrating an example, non-limiting embodiment of a virtualized communication network in accordance with various aspects described herein. In particular a virtualized communication network is presented that can be used to implement some or all of the subsystems and functions of system 100, the subsystems and functions of system 200, and method 230 presented in FIGS. 1, 2A, 2B, 2C, and 3. For example, virtualized communication network 300 can facilitate in whole or in part a protocol and signaling framework enabling ML models in wireless communication networks.

[0056] In particular, a cloud networking architecture is shown that leverages cloud technologies and supports rapid innovation and scalability via a transport layer 350, a virtualized network function cloud 325 and/or one or more cloud computing environments 375. In various embodiments, this cloud networking architecture is an open architecture that leverages application programming interfaces (APIs); reduces complexity from services and operations; supports more nimble business models; and rapidly and seamlessly scales to meet evolving customer requirements including traffic growth, diversity of traffic types, and diversity of performance and reliability expectations.

[0057] In contrast to traditional network elements—which are typically integrated to perform a single function, the virtualized communication network employs virtual network elements (VNEs) 330, 332, 334, etc. that perform some or all of the functions of network elements 150, 152, 154, 156, etc. For example, the network architecture can provide a substrate of networking capability, often called Network Function Virtualization Infrastructure (NFVI) or simply infrastructure that is capable of being directed with software and Software Defined Networking (SDN) protocols to perform a broad variety of network functions and services. This infrastructure can include several types of substrates. The most typical type of substrate being servers that support Network Function Virtualization (NFV), followed by packet forwarding capabilities based on generic computing resources, with specialized network technologies brought to bear when general-purpose processors or general-purpose integrated circuit devices offered by merchants (referred to herein as merchant silicon) are not appropriate. In this case, communication services can be implemented as cloud-centric workloads.

[0058] As an example, a traditional network element 150 (shown in FIG. 1), such as an edge router can be implemented via a VNE 330 composed of NFV software modules,

merchant silicon, and associated controllers. The software can be written so that increasing workload consumes incremental resources from a common resource pool, and moreover so that it is elastic: so, the resources are only consumed when needed. In a similar fashion, other network elements such as other routers, switches, edge caches, and middle boxes are instantiated from the common resource pool. Such sharing of infrastructure across a broad set of uses makes planning and growing infrastructure easier to manage.

[0059] In an embodiment, the transport layer **350** includes fiber, cable, wired and/or wireless transport elements, network elements and interfaces to provide broadband access **110**, wireless access **120**, voice access **130**, media access **140** and/or access to content sources **175** for distribution of content to any or all of the access technologies. In particular, in some cases a network element needs to be positioned at a specific place, and this allows for less sharing of common infrastructure. Other times, the network elements have specific physical layer adapters that cannot be abstracted or virtualized and might require special DSP code and analog front ends (AFEs) that do not lend themselves to implementation as VNEs **330**, **332** or **334**. These network elements can be included in transport layer **350**.

[0060] The virtualized network function cloud **325** interfaces with the transport layer **350** to provide the VNEs **330**, **332**, **334**, etc. to provide specific NFVs. In particular, the virtualized network function cloud **325** leverages cloud operations, applications, and architectures to support networking workloads. The virtualized network elements **330**, **332** and **334** can employ network function software that provides either a one-for-one mapping of traditional network element function or alternately some combination of network functions designed for cloud computing. For example, VNEs **330**, **332** and **334** can include route reflectors, domain name system (DNS) servers, and dynamic host configuration protocol (DHCP) servers, system architecture evolution (SAE) and/or mobility management entity (MME) gateways, broadband network gateways, IP edge routers for IP-VPN, Ethernet and other services, load balancers, distributors and other network elements. Because these elements do not typically need to forward large amounts of traffic, their workload can be distributed across a number of servers—each of which adds a portion of the capability, and which creates an elastic function with higher availability overall than its former monolithic version. These virtual network elements **330**, **332**, **334**, etc. can be instantiated and managed using an orchestration approach similar to those used in cloud compute services.

[0061] The cloud computing environments **375** can interface with the virtualized network function cloud **325** via APIs that expose functional capabilities of the VNEs **330**, **332**, **334**, etc. to provide the flexible and expanded capabilities to the virtualized network function cloud **325**. In particular, network workloads may have applications distributed across the virtualized network function cloud **325** and cloud computing environment **375** and in the commercial cloud or might simply orchestrate workloads supported entirely in NFV infrastructure from these third-party locations.

[0062] Turning now to FIG. 4, there is illustrated a block diagram of a computing environment in accordance with various aspects described herein. In order to provide additional context for various embodiments of the embodiments described herein, FIG. 4 and the following discussion are

intended to provide a brief, general description of a suitable computing environment **400** in which the various embodiments of the subject disclosure can be implemented. In particular, computing environment **400** can be used in the implementation of network elements **150**, **152**, **154**, **156**, access terminal **112**, base station or access point **122**, switching device **132**, media terminal **142**, and/or VNEs **330**, **332**, **334**, etc. Each of these devices can be implemented via computer-executable instructions that can run on one or more computers, and/or in combination with other program modules and/or as a combination of hardware and software. For example, computing environment **400** can facilitate in whole or in part a protocol and signaling framework enabling ML models in wireless communication networks.

[0063] Generally, program modules comprise routines, programs, components, data structures, etc., that perform particular tasks or implement particular abstract data types. Moreover, those skilled in the art will appreciate that the methods can be practiced with other computer system configurations, comprising single-processor or multiprocessor computer systems, minicomputers, mainframe computers, as well as personal computers, hand-held computing devices, microprocessor-based or programmable consumer electronics, and the like, each of which can be operatively coupled to one or more associated devices.

[0064] As used herein, a processing circuit includes one or more processors as well as other application specific circuits such as an application specific integrated circuit, digital logic circuit, state machine, programmable gate array or other circuit that processes input signals or data and that produces output signals or data in response thereto. It should be noted that while any functions and features described herein in association with the operation of a processor could likewise be performed by a processing circuit.

[0065] The illustrated embodiments of the embodiments herein can be also practiced in distributed computing environments where certain tasks are performed by remote processing devices that are linked through a communications network. In a distributed computing environment, program modules can be located in both local and remote memory storage devices.

[0066] Computing devices typically comprise a variety of media, which can comprise computer-readable storage media and/or communications media, which two terms are used herein differently from one another as follows. Computer-readable storage media can be any available storage media that can be accessed by the computer and comprises both volatile and nonvolatile media, removable and non-removable media. By way of example, and not limitation, computer-readable storage media can be implemented in connection with any method or technology for storage of information such as computer-readable instructions, program modules, structured data or unstructured data.

[0067] Computer-readable storage media can comprise, but are not limited to, random access memory (RAM), read only memory (ROM), electrically erasable programmable read only memory (EEPROM), flash memory or other memory technology, compact disk read only memory (CD-ROM), digital versatile disk (DVD) or other optical disk storage, magnetic cassettes, magnetic tape, magnetic disk storage or other magnetic storage devices or other tangible and/or non-transitory media which can be used to store desired information. In this regard, the terms “tangible” or

“non-transitory” herein as applied to storage, memory or computer-readable media, are to be understood to exclude only propagating transitory signals per se as modifiers and do not relinquish rights to all standard storage, memory or computer-readable media that are not only propagating transitory signals per se.

[0068] Computer-readable storage media can be accessed by one or more local or remote computing devices, e.g., via access requests, queries or other data retrieval protocols, for a variety of operations with respect to the information stored by the medium.

[0069] Communications media typically embody computer-readable instructions, data structures, program modules or other structured or unstructured data in a data signal such as a modulated data signal, e.g., a carrier wave or other transport mechanism, and comprises any information delivery or transport media. The term “modulated data signal” or signals refers to a signal that has one or more of its characteristics set or changed in such a manner as to encode information in one or more signals. By way of example, and not limitation, communication media comprise wired media, such as a wired network or direct-wired connection, and wireless media such as acoustic, RF, infrared and other wireless media.

[0070] With reference again to FIG. 4, the example environment can comprise a computer 402, the computer 402 comprising a processing unit 404, a system memory 406 and a system bus 408. The system bus 408 couples system components including, but not limited to, the system memory 406 to the processing unit 404. The processing unit 404 can be any of various commercially available processors. Dual microprocessors and other multiprocessor architectures can also be employed as the processing unit 404.

[0071] The system bus 408 can be any of several types of bus structure that can further interconnect to a memory bus (with or without a memory controller), a peripheral bus, and a local bus using any of a variety of commercially available bus architectures. The system memory 406 comprises ROM 410 and RAM 412. A basic input/output system (BIOS) can be stored in a non-volatile memory such as ROM, erasable programmable read only memory (EPROM), EEPROM, which BIOS contains the basic routines that help to transfer information between elements within the computer 402, such as during startup. The RAM 412 can also comprise a high-speed RAM such as static RAM for caching data.

[0072] The computer 402 further comprises an internal hard disk drive (HDD) 414 (e.g., EIDE, SATA), which internal HDD 414 can also be configured for external use in a suitable chassis (not shown), a magnetic floppy disk drive (FDD) 416, (e.g., to read from or write to a removable diskette 418) and an optical disk drive 420, (e.g., reading a CD-ROM disk 422 or, to read from or write to other high-capacity optical media such as the DVD). The HDD 414, magnetic FDD 416 and optical disk drive 420 can be connected to the system bus 408 by a hard disk drive interface 424, a magnetic disk drive interface 426 and an optical drive interface 428, respectively. The hard disk drive interface 424 for external drive implementations comprises at least one or both of Universal Serial Bus (USB) and Institute of Electrical and Electronics Engineers (IEEE) 1394 interface technologies. Other external drive connection technologies are within contemplation of the embodiments described herein.

[0073] The drives and their associated computer-readable storage media provide nonvolatile storage of data, data structures, computer-executable instructions, and so forth. For the computer 402, the drives and storage media accommodate the storage of any data in a suitable digital format. Although the description of computer-readable storage media above refers to a hard disk drive (HDD), a removable magnetic diskette, and a removable optical media such as a CD or DVD, it should be appreciated by those skilled in the art that other types of storage media which are readable by a computer, such as zip drives, magnetic cassettes, flash memory cards, cartridges, and the like, can also be used in the example operating environment, and further, that any such storage media can contain computer-executable instructions for performing the methods described herein.

[0074] A number of program modules can be stored in the drives and RAM 412, comprising an operating system 430, one or more application programs 432, other program modules 434 and program data 436. All or portions of the operating system, applications, modules, and/or data can also be cached in the RAM 412. The systems and methods described herein can be implemented utilizing various commercially available operating systems or combinations of operating systems.

[0075] A user can enter commands and information into the computer 402 through one or more wired/wireless input devices, e.g., a keyboard 438 and a pointing device, such as a mouse 440. Other input devices (not shown) can comprise a microphone, an infrared (IR) remote control, a joystick, a game pad, a stylus pen, touch screen or the like. These and other input devices are often connected to the processing unit 404 through an input device interface 442 that can be coupled to the system bus 408, but can be connected by other interfaces, such as a parallel port, an IEEE 1394 serial port, a game port, a universal serial bus (USB) port, an IR interface, etc.

[0076] A monitor 444 or other type of display device can be also connected to the system bus 408 via an interface, such as a video adapter 446. It will also be appreciated that in alternative embodiments, a monitor 444 can also be any display device (e.g., another computer having a display, a smart phone, a tablet computer, etc.) for receiving display information associated with computer 402 via any communication means, including via the Internet and cloud-based networks. In addition to the monitor 444, a computer typically comprises other peripheral output devices (not shown), such as speakers, printers, etc.

[0077] The computer 402 can operate in a networked environment using logical connections via wired and/or wireless communications to one or more remote computers, such as a remote computer(s) 448. The remote computer(s) 448 can be a workstation, a server computer, a router, a personal computer, portable computer, microprocessor-based entertainment appliance, a peer device or other common network node, and typically comprises many or all of the elements described relative to the computer 402, although, for purposes of brevity, only a remote memory/storage device 450 is illustrated. The logical connections depicted comprise wired/wireless connectivity to a local area network (LAN) 452 and/or larger networks, e.g., a wide area network (WAN) 454. Such LAN and WAN networking environments are commonplace in offices and companies, and facilitate enterprise-wide computer networks, such as

intranets, all of which can connect to a global communications network, e.g., the Internet.

[0078] When used in a LAN networking environment, the computer **402** can be connected to the LAN **452** through a wired and/or wireless communication network interface or adapter **456**. The adapter **456** can facilitate wired or wireless communication to the LAN **452**, which can also comprise a wireless AP disposed thereon for communicating with the adapter **456**.

[0079] When used in a WAN networking environment, the computer **402** can comprise a modem **458** or can be connected to a communications server on the WAN **454** or has other means for establishing communications over the WAN **454**, such as by way of the Internet. The modem **458**, which can be internal or external and a wired or wireless device, can be connected to the system bus **408** via the input device interface **442**. In a networked environment, program modules depicted relative to the computer **402** or portions thereof, can be stored in the remote memory/storage device **450**. It will be appreciated that the network connections shown are example and other means of establishing a communications link between the computers can be used.

[0080] The computer **402** can be operable to communicate with any wireless devices or entities operatively disposed in wireless communication, e.g., a printer, scanner, desktop and/or portable computer, portable data assistant, communications satellite, any piece of equipment or location associated with a wirelessly detectable tag (e.g., a kiosk, news stand, restroom), and telephone. This can comprise Wireless Fidelity (Wi-Fi) and BLUETOOTH® wireless technologies. Thus, the communication can be a predefined structure as with a conventional network or simply an ad hoc communication between at least two devices.

[0081] Wi-Fi can allow connection to the Internet from a couch at home, a bed in a hotel room or a conference room at work, without wires. Wi-Fi is a wireless technology similar to that used in a cell phone that enables such devices, e.g., computers, to send and receive data indoors and out; anywhere within the range of a base station. Wi-Fi networks use radio technologies called IEEE 802.11 (a, b, g, n, ac, ag, etc.) to provide secure, reliable, fast wireless connectivity. A Wi-Fi network can be used to connect computers to each other, to the Internet, and to wired networks (which can use IEEE 802.3 or Ethernet). Wi-Fi networks operate in the unlicensed 2.4 and 5 GHz radio bands for example or with products that contain both bands (dual band), so the networks can provide real-world performance similar to the basic 10BaseT wired Ethernet networks used in many offices.

[0082] Turning now to FIG. 5, an embodiment **500** of a mobile network platform **510** is shown that is an example of network elements **150**, **152**, **154**, **156**, and/or VNEs **330**, **332**, **334**, etc. For example, platform **510** can facilitate in whole or in part a protocol and signaling framework enabling ML models in wireless communication networks. In one or more embodiments, the mobile network platform **510** can generate and receive signals transmitted and received by base stations or access points such as base station or access point **122**. Generally, mobile network platform **510** can comprise components, e.g., nodes, gateways, interfaces, servers, or disparate platforms, that facilitate both packet-switched (PS) (e.g., internet protocol (IP), frame relay, asynchronous transfer mode (ATM)) and circuit-switched (CS) traffic (e.g., voice and data), as well as

control generation for networked wireless telecommunication. As a non-limiting example, mobile network platform **510** can be included in telecommunications carrier networks and can be considered carrier-side components as discussed elsewhere herein. Mobile network platform **510** comprises CS gateway node(s) **512** which can interface CS traffic received from legacy networks like telephony network(s) **540** (e.g., public switched telephone network (PSTN), or public land mobile network (PLMN)) or a signaling system #7 (SS7) network **560**. CS gateway node(s) **512** can authorize and authenticate traffic (e.g., voice) arising from such networks. Additionally, CS gateway node(s) **512** can access mobility, or roaming, data generated through SS7 network **560**; for instance, mobility data stored in a visited location register (VLR), which can reside in memory **530**. Moreover, CS gateway node(s) **512** interfaces CS-based traffic and signaling and PS gateway node(s) **518**. As an example, in a 3GPP UMTS network, CS gateway node(s) **512** can be realized at least in part in gateway GPRS support node(s) (GGSN). It should be appreciated that functionality and specific operation of CS gateway node(s) **512**, PS gateway node(s) **518**, and serving node(s) **516**, is provided and dictated by radio technology(ies) utilized by mobile network platform **510** for telecommunication over a radio access network **520** with other devices, such as a radiotelephone **575**.

[0083] In addition to receiving and processing CS-switched traffic and signaling, PS gateway node(s) **518** can authorize and authenticate PS-based data sessions with served mobile devices. Data sessions can comprise traffic, or content(s), exchanged with networks external to the mobile network platform **510**, like wide area network(s) (WANs) **550**, enterprise network(s) **570**, and service network(s) **580**, which can be embodied in local area network(s) (LANs), can also be interfaced with mobile network platform **510** through PS gateway node(s) **518**. It is to be noted that WANs **550** and enterprise network(s) **570** can embody, at least in part, a service network(s) like IP multimedia subsystem (IMS). Based on radio technology layer(s) available in technology resource(s) or radio access network **520**, PS gateway node(s) **518** can generate packet data protocol contexts when a data session is established; other data structures that facilitate routing of packetized data also can be generated. To that end, in an aspect, PS gateway node(s) **518** can comprise a tunnel interface (e.g., tunnel termination gateway (TTG) in 3GPP UMTS network(s) (not shown)) which can facilitate packetized communication with disparate wireless network(s), such as Wi-Fi networks.

[0084] In embodiment **500**, mobile network platform **510** also comprises serving node(s) **516** that, based upon available radio technology layer(s) within technology resource(s) in the radio access network **520**, convey the various packetized flows of data streams received through PS gateway node(s) **518**. It is to be noted that for technology resource(s) that rely primarily on CS communication, server node(s) can deliver traffic without reliance on PS gateway node(s) **518**; for example, server node(s) can embody at least in part a mobile switching center. As an example, in a 3GPP UMTS network, serving node(s) **516** can be embodied in serving GPRS support node(s) (SGSN).

[0085] For radio technologies that exploit packetized communication, server(s) **514** in mobile network platform **510** can execute numerous applications that can generate multiple disparate packetized data streams or flows, and manage

(e.g., schedule, queue, format . . .) such flows. Such application(s) can comprise add-on features to standard services (for example, provisioning, billing, customer support . . .) provided by mobile network platform 510. Data streams (e.g., content(s) that are part of a voice call or data session) can be conveyed to PS gateway node(s) 518 for authorization/authentication and initiation of a data session, and to serving node(s) 516 for communication thereafter. In addition to application server, server(s) 514 can comprise utility server(s), a utility server can comprise a provisioning server, an operations and maintenance server, a security server that can implement at least in part a certificate authority and firewalls as well as other security mechanisms, and the like. In an aspect, security server(s) secure communication served through mobile network platform 510 to ensure network's operation and data integrity in addition to authorization and authentication procedures that CS gateway node(s) 512 and PS gateway node(s) 518 can enact. Moreover, provisioning server(s) can provision services from external network(s) like networks operated by a disparate service provider; for instance, WAN 550 or Global Positioning System (GPS) network(s) (not shown). Provisioning server(s) can also provision coverage through networks associated to mobile network platform 510 (e.g., deployed and operated by the same service provider), such as the distributed antennas networks shown in FIG. 1 (s) that enhance wireless service coverage by providing more network coverage.

[0086] It is to be noted that server(s) 514 can comprise one or more processors configured to confer at least in part the functionality of mobile network platform 510. To that end, the one or more processors can execute code instructions stored in memory 530, for example. It should be appreciated that server(s) 514 can comprise a content manager, which operates in substantially the same manner as described hereinbefore.

[0087] In example embodiment 500, memory 530 can store information related to operation of mobile network platform 510. Other operational information can comprise provisioning information of mobile devices served through mobile network platform 510, subscriber databases; application intelligence, pricing schemes, e.g., promotional rates, flat-rate programs, couponing campaigns; technical specification(s) consistent with telecommunication protocols for operation of disparate radio, or wireless, technology layers; and so forth. Memory 530 can also store information from at least one of telephony network(s) 540, WAN 550, SS7 network 560, or enterprise network(s) 570. In an aspect, memory 530 can be, for example, accessed as part of a data store component or as a remotely connected memory store.

[0088] In order to provide a context for the various aspects of the disclosed subject matter, FIG. 5, and the following discussion, are intended to provide a brief, general description of a suitable environment in which the various aspects of the disclosed subject matter can be implemented. While the subject matter has been described above in the general context of computer-executable instructions of a computer program that runs on a computer and/or computers, those skilled in the art will recognize that the disclosed subject matter also can be implemented in combination with other program modules. Generally, program modules comprise routines, programs, components, data structures, etc. that perform particular tasks and/or implement particular abstract data types.

[0089] Turning now to FIG. 6, an illustrative embodiment of a communication device 600 is shown. The communication device 600 can serve as an illustrative embodiment of devices such as data terminals 114, mobile devices 124, vehicle 126, display devices 144 or other client devices for communication via either communications network 125. For example, computing device 600 can facilitate in whole or in part a protocol and signaling framework enabling ML models in wireless communication networks.

[0090] The communication device 600 can comprise a wireline and/or wireless transceiver 602 (herein transceiver 602), a user interface (UI) 604, a power supply 614, a location receiver 616, a motion sensor 618, an orientation sensor 620, and a controller 606 for managing operations thereof. The transceiver 602 can support short-range or long-range wireless access technologies such as Bluetooth®, ZigBee®, Wi-Fi, DECT, or cellular communication technologies, just to mention a few (Bluetooth® and ZigBee® are trademarks registered by the Bluetooth® Special Interest Group and the ZigBee® Alliance, respectively). Cellular technologies can include, for example, CDMA-1X, UMTS/HSDPA, GSM/GPRS, TDMA/EDGE, EV/DO, WiMAX, SDR, LTE, as well as other next generation wireless communication technologies as they arise. The transceiver 602 can also be adapted to support circuit-switched wireline access technologies (such as PSTN), packet-switched wireline access technologies (such as TCP/IP, VoIP, etc.), and combinations thereof.

[0091] The UI 604 can include a depressible or touch-sensitive keypad 608 with a navigation mechanism such as a roller ball, a joystick, a mouse, or a navigation disk for manipulating operations of the communication device 600. The keypad 608 can be an integral part of a housing assembly of the communication device 600 or an independent device operably coupled thereto by a tethered wireline interface (such as a USB cable) or a wireless interface supporting for example Bluetooth®. The keypad 608 can represent a numeric keypad commonly used by phones, and/or a QWERTY keypad with alphanumeric keys. The UI 604 can further include a display 610 such as monochrome or color LCD (Liquid Crystal Display), OLED (Organic Light Emitting Diode) or other suitable display technology for conveying images to an end user of the communication device 600. In an embodiment where the display 610 is touch-sensitive, a portion or all of the keypad 608 can be presented by way of the display 610 with navigation features.

[0092] The display 610 can use touch screen technology to also serve as a user interface for detecting user input. As a touch screen display, the communication device 600 can be adapted to present a user interface having graphical user interface (GUI) elements that can be selected by a user with a touch of a finger. The display 610 can be equipped with capacitive, resistive or other forms of sensing technology to detect how much surface area of a user's finger has been placed on a portion of the touch screen display. This sensing information can be used to control the manipulation of the GUI elements or other functions of the user interface. The display 610 can be an integral part of the housing assembly of the communication device 600 or an independent device communicatively coupled thereto by a tethered wireline interface (such as a cable) or a wireless interface.

[0093] The UI 604 can also include an audio system 612 that utilizes audio technology for conveying low volume

audio (such as audio heard in proximity of a human ear) and high-volume audio (such as speakerphone for hands free operation). The audio system 612 can further include a microphone for receiving audible signals of an end user. The audio system 612 can also be used for voice recognition applications. The UI 604 can further include an image sensor 613 such as a charged coupled device (CCD) camera for capturing still or moving images.

[0094] The power supply 614 can utilize common power management technologies such as replaceable and rechargeable batteries, supply regulation technologies, and/or charging system technologies for supplying energy to the components of the communication device 600 to facilitate long-range or short-range portable communications. Alternatively, or in combination, the charging system can utilize external power sources such as DC power supplied over a physical interface such as a USB port or other suitable tethering technologies.

[0095] The location receiver 616 can utilize location technology such as a global positioning system (GPS) receiver capable of assisted GPS for identifying a location of the communication device 600 based on signals generated by a constellation of GPS satellites, which can be used for facilitating location services such as navigation. The motion sensor 618 can utilize motion sensing technology such as an accelerometer, a gyroscope, or other suitable motion sensing technology to detect motion of the communication device 600 in three-dimensional space. The orientation sensor 620 can utilize orientation sensing technology such as a magnetometer to detect the orientation of the communication device 600 (north, south, west, and east, as well as combined orientations in degrees, minutes, or other suitable orientation metrics).

[0096] The communication device 600 can use the transceiver 602 to also determine a proximity to a cellular, Wi-Fi, Bluetooth®, or other wireless access points by sensing techniques such as utilizing a received signal strength indicator (RSSI) and/or signal time of arrival (TOA) or time of flight (TOF) measurements. The controller 606 can utilize computing technologies such as a microprocessor, a digital signal processor (DSP), programmable gate arrays, application specific integrated circuits, and/or a video processor with associated storage memory such as Flash, ROM, RAM, SRAM, DRAM or other storage technologies for executing computer instructions, controlling, and processing data supplied by the aforementioned components of the communication device 600.

[0097] Other components not shown in FIG. 6 can be used in one or more embodiments of the subject disclosure. For instance, the communication device 600 can include a slot for adding or removing an identity module such as a Subscriber Identity Module (SIM) card or Universal Integrated Circuit Card (UICC). SIM or UICC cards can be used for identifying subscriber services, executing programs, storing subscriber data, and so on.

[0098] The terms “first,” “second,” “third,” and so forth, as used in the claims, unless otherwise clear by context, is for clarity only and does not otherwise indicate or imply any order in time. For instance, “a first determination,” “a second determination,” and “a third determination,” does not indicate or imply that the first determination is to be made before the second determination, or vice versa, etc.

[0099] In the subject specification, terms such as “store,” “storage,” “data store,” “data storage,” “database,” and sub-

stantially any other information storage component relevant to operation and functionality of a component, refer to “memory components,” or entities embodied in a “memory” or components comprising the memory. It will be appreciated that the memory components described herein can be either volatile memory or nonvolatile memory, or can comprise both volatile and nonvolatile memory, by way of illustration, and not limitation, volatile memory, non-volatile memory, disk storage, and memory storage. Further, non-volatile memory can be included in read only memory (ROM), programmable ROM (PROM), electrically programmable ROM (EPROM), electrically erasable ROM (EEPROM), or flash memory. Volatile memory can comprise random access memory (RAM), which acts as external cache memory. By way of illustration and not limitation, RAM is available in many forms such as synchronous RAM (SRAM), dynamic RAM (DRAM), synchronous DRAM (SDRAM), double data rate SDRAM (DDR SDRAM), enhanced SDRAM (ESDRAM), Synchlink DRAM (SL-DRAM), and direct Rambus RAM (DRRAM). Additionally, the disclosed memory components of systems or methods herein are intended to comprise, without being limited to comprising, these and any other suitable types of memory.

[0100] Moreover, it will be noted that the disclosed subject matter can be practiced with other computer system configurations, comprising single-processor or multiprocessor computer systems, mini-computing devices, mainframe computers, as well as personal computers, hand-held computing devices (e.g., PDA, phone, smartphone, watch, tablet computers, netbook computers, etc.), microprocessor-based or programmable consumer or industrial electronics, and the like. The illustrated aspects can also be practiced in distributed computing environments where tasks are performed by remote processing devices that are linked through a communications network; however, some if not all aspects of the subject disclosure can be practiced on stand-alone computers. In a distributed computing environment, program modules can be located in both local and remote memory storage devices.

[0101] In one or more embodiments, information regarding use of services can be generated including services being accessed, media consumption history, user preferences, and so forth. This information can be obtained by various methods including user input, detecting types of communications (e.g., video content vs. audio content), analysis of content streams, sampling, and so forth. The generating, obtaining and/or monitoring of this information can be responsive to an authorization provided by the user. In one or more embodiments, an analysis of data can be subject to authorization from user(s) associated with the data, such as an opt-in, an opt-out, acknowledgement requirements, notifications, selective authorization based on types of data, and so forth.

[0102] Some of the embodiments described herein can also employ artificial intelligence (AI) to facilitate automating one or more features described herein. The embodiments (e.g., in connection with automatically identifying acquired cell sites that provide a maximum value/benefit after addition to an existing communication network) can employ various AI-based schemes for carrying out various embodiments thereof. Moreover, the classifier can be employed to determine a ranking or priority of each cell site of the acquired network. A classifier is a function that maps an input attribute vector, $x=(x_1, x_2, x_3, x_4 \dots x_n)$, to a

confidence that the input belongs to a class, that is, $f(x)$ = confidence (class). Such classification can employ a probabilistic and/or statistical-based analysis (e.g., factoring into the analysis utilities and costs) to determine or infer an action that a user desires to be automatically performed. A support vector machine (SVM) is an example of a classifier that can be employed. The SVM operates by finding a hypersurface in the space of possible inputs, which the hypersurface attempts to split the triggering criteria from the non-triggering events. Intuitively, this makes the classification correct for testing data that is near, but not identical to training data. Other directed and undirected model classification approaches comprise, e.g., naïve Bayes, Bayesian networks, decision trees, neural networks, fuzzy logic models, and probabilistic classification models providing different patterns of independence can be employed. Classification as used herein also is inclusive of statistical regression that is utilized to develop models of priority.

[0103] As will be readily appreciated, one or more of the embodiments can employ classifiers that are explicitly trained (e.g., via a generic training data) as well as implicitly trained (e.g., via observing UE behavior, operator preferences, historical information, receiving extrinsic information). For example, SVMs can be configured via a learning or training phase within a classifier constructor and feature selection module. Thus, the classifier(s) can be used to automatically learn and perform a number of functions, including but not limited to determining according to pre-determined criteria which of the acquired cell sites will benefit a maximum number of subscribers and/or which of the acquired cell sites will add minimum value to the existing communication network coverage, etc.

[0104] As used in some contexts in this application, in some embodiments, the terms “component,” “system” and the like are intended to refer to, or comprise, a computer-related entity or an entity related to an operational apparatus with one or more specific functionalities, wherein the entity can be either hardware, a combination of hardware and software, software, or software in execution. As an example, a component may be, but is not limited to being, a process running on a processor, a processor, an object, an executable, a thread of execution, computer-executable instructions, a program, and/or a computer. By way of illustration and not limitation, both an application running on a server and the server can be a component. One or more components may reside within a process and/or thread of execution and a component may be localized on one computer and/or distributed between two or more computers. In addition, these components can execute from various computer readable media having various data structures stored thereon. The components may communicate via local and/or remote processes such as in accordance with a signal having one or more data packets (e.g., data from one component interacting with another component in a local system, distributed system, and/or across a network such as the Internet with other systems via the signal). As another example, a component can be an apparatus with specific functionality provided by mechanical parts operated by electric or electronic circuitry, which is operated by a software or firmware application executed by a processor, wherein the processor can be internal or external to the apparatus and executes at least a part of the software or firmware application. As yet another example, a component can be an apparatus that provides specific functionality through electronic compo-

nents without mechanical parts, the electronic components can comprise a processor therein to execute software or firmware that confers at least in part the functionality of the electronic components. While various components have been illustrated as separate components, it will be appreciated that multiple components can be implemented as a single component, or a single component can be implemented as multiple components, without departing from example embodiments.

[0105] Further, the various embodiments can be implemented as a method, apparatus or article of manufacture using standard programming and/or engineering techniques to produce software, firmware, hardware or any combination thereof to control a computer to implement the disclosed subject matter. The term “article of manufacture” as used herein is intended to encompass a computer program accessible from any computer-readable device or computer-readable storage/communications media. For example, computer readable storage media can include, but are not limited to, magnetic storage devices (e.g., hard disk, floppy disk, magnetic strips), optical disks (e.g., compact disk (CD), digital versatile disk (DVD)), smart cards, and flash memory devices (e.g., card, stick, key drive). Of course, those skilled in the art will recognize many modifications can be made to this configuration without departing from the scope or spirit of the various embodiments.

[0106] In addition, the words “example” and “exemplary” are used herein to mean serving as an instance or illustration. Any embodiment or design described herein as “example” or “exemplary” is not necessarily to be construed as preferred or advantageous over other embodiments or designs. Rather, use of the word example or exemplary is intended to present concepts in a concrete fashion. As used in this application, the term “or” is intended to mean an inclusive “or” rather than an exclusive “or”. That is, unless specified otherwise or clear from context, “X employs A or B” is intended to mean any of the natural inclusive permutations. That is, if X employs A; X employs B; or X employs both A and B, then “X employs A or B” is satisfied under any of the foregoing instances. In addition, the articles “a” and “an” as used in this application and the appended claims should generally be construed to mean “one or more” unless specified otherwise or clear from context to be directed to a singular form.

[0107] Moreover, terms such as “user equipment,” “mobile station,” “mobile,” subscriber station,” “access terminal,” “terminal,” “handset,” “mobile device” (and/or terms representing similar terminology) can refer to a wireless device utilized by a subscriber or user of a wireless communication service to receive or convey data, control, voice, video, sound, gaming or substantially any data-stream or signaling-stream. The foregoing terms are utilized interchangeably herein and with reference to the related drawings.

[0108] Furthermore, the terms “user,” “subscriber,” “customer,” “consumer” and the like are employed interchangeably throughout, unless context warrants particular distinctions among the terms. It should be appreciated that such terms can refer to human entities or automated components supported through artificial intelligence (e.g., a capacity to make inference based, at least, on complex mathematical formalisms), which can provide simulated vision, sound recognition and so forth.

[0109] As employed herein, the term “processor” can refer to substantially any computing processing unit or device comprising, but not limited to comprising, single-core processors; single-processors with software multithread execution capability; multi-core processors; multi-core processors with software multithread execution capability; multi-core processors with hardware multithread technology; parallel platforms; and parallel platforms with distributed shared memory. Additionally, a processor can refer to an integrated circuit, an application specific integrated circuit (ASIC), a digital signal processor (DSP), a field programmable gate array (FPGA), a programmable logic controller (PLC), a complex programmable logic device (CPLD), a discrete gate or transistor logic, discrete hardware components or any combination thereof designed to perform the functions described herein. Processors can exploit nano-scale architectures such as, but not limited to, molecular and quantum-dot based transistors, switches and gates, in order to optimize space usage or enhance performance of user equipment. A processor can also be implemented as a combination of computing processing units.

[0110] As used herein, terms such as “data storage,” “data storage,” “database,” and substantially any other information storage component relevant to operation and functionality of a component, refer to “memory components,” or entities embodied in a “memory” or components comprising the memory. It will be appreciated that the memory components or computer-readable storage media, described herein can be either volatile memory or nonvolatile memory or can include both volatile and nonvolatile memory.

[0111] What has been described above includes mere examples of various embodiments. It is, of course, not possible to describe every conceivable combination of components or methodologies for purposes of describing these examples, but one of ordinary skill in the art can recognize that many further combinations and permutations of the present embodiments are possible. Accordingly, the embodiments disclosed and/or claimed herein are intended to embrace all such alterations, modifications and variations that fall within the spirit and scope of the appended claims. Furthermore, to the extent that the term “includes” is used in either the detailed description or the claims, such term is intended to be inclusive in a manner similar to the term “comprising” as “comprising” is interpreted when employed as a transitional word in a claim.

[0112] In addition, a flow diagram may include a “start” and/or “continue” indication. The “start” and “continue” indications reflect that the steps presented can optionally be incorporated in or otherwise used in conjunction with other routines. In this context, “start” indicates the beginning of the first step presented and may be preceded by other activities not specifically shown. Further, the “continue” indication reflects that the steps presented may be performed multiple times and/or may be succeeded by other activities not specifically shown. Further, while a flow diagram indicates a particular ordering of steps, other orderings are likewise possible provided that the principles of causality are maintained.

[0113] As may also be used herein, the term(s) “operably coupled to”, “coupled to”, and/or “coupling” includes direct coupling between items and/or indirect coupling between items via one or more intervening items. Such items and intervening items include, but are not limited to, junctions, communication paths, components, circuit elements, cir-

cuits, functional blocks, and/or devices. As an example of indirect coupling, a signal conveyed from a first item to a second item may be modified by one or more intervening items by modifying the form, nature or format of information in a signal, while one or more elements of the information in the signal are nevertheless conveyed in a manner than can be recognized by the second item. In a further example of indirect coupling, an action in a first item can cause a reaction on the second item, as a result of actions and/or reactions in one or more intervening items.

[0114] Although specific embodiments have been illustrated and described herein, it should be appreciated that any arrangement which achieves the same or similar purpose may be substituted for the embodiments described or shown by the subject disclosure. The subject disclosure is intended to cover any and all adaptations or variations of various embodiments. Combinations of the above embodiments, and other embodiments not specifically described herein, can be used in the subject disclosure. For instance, one or more features from one or more embodiments can be combined with one or more features of one or more other embodiments. In one or more embodiments, features that are positively recited can also be negatively recited and excluded from the embodiment with or without replacement by another structural and/or functional feature. The steps or functions described with respect to the embodiments of the subject disclosure can be performed in any order. The steps or functions described with respect to the embodiments of the subject disclosure can be performed alone or in combination with other steps or functions of the subject disclosure, as well as from other embodiments or from other steps that have not been described in the subject disclosure. Further, more than or less than all of the features described with respect to an embodiment can also be utilized.

[0115] One or more of the embodiments described herein can be combined in whole or in part with the embodiments described in co-pending U.S. patent application Ser. No. _____ (having Attorney Docket No. 2023-0349_7785-3348A), entitled “SYSTEMS AND METHODS FOR FACILITATING IDENTIFICATION AND DELIVERY OF MACHINE LEARNING MODELS IN WIRELESS COMMUNICATION NETWORKS,” filed on even date herewith, the disclosure of which is hereby incorporated by reference herein.

[0116] One or more of the embodiments described herein can be combined in whole or in part with the embodiments described in co-pending U.S. patent application Ser. No. _____ (having Attorney Docket No. 2023-0145_7785-3349A), entitled “METHODS, SYSTEMS, AND DEVICES IN SELECTING ARTIFICIAL (AI)/MACHINE LEARNING (ML) MODELS IN RADIO ACCESS NETWORKS,” filed on even date herewith, the disclosure of which is hereby incorporated by reference herein.

What is claimed is:

1. A method, comprising:

formulating, by a processing system including a processor, a set of data collection parameters that are configurable to collect data corresponding to one or more network events, wherein the one or more network events occur in real time or near real time in wireless communication networks;

receiving, by the processing system, first data collected based on the set of data collection parameters from a first network entity operating in the wireless communication networks;

receiving, by the processing system, second data collected based on the set of data collection parameters from a second network entity operating in the wireless communication networks;

generating a training data based on the first collected data, the second collected data or both; and

training, by the processing system, a machine learning model based on the generated training data to learn the one or more network events.

2. The method of claim 1, wherein the formulating the set of data collection parameters comprises configuring the set of data collection parameters to specify a predetermined validity time duration.

3. The method of claim 1, wherein the formulating the set of data collection parameters comprises configuring the set of data collection parameters to maintain a predetermined time interval for periodically collected information until the predetermined time interval is overridden by an updated time interval.

4. The method of claim 1, further comprising receiving, by the processing system, an on-demand request for data collection from a third network entity, the on-demand request specifying a predetermined time duration and target measurement information.

5. The method of claim 4, wherein the formulating the set of data collection parameters comprises configuring the set of data collection parameters to specify the predetermined time duration specified in the on-demand request and the target measurement information.

6. The method of claim 1, wherein the formulating the set of data collection parameters comprises configuring the set of data collection parameters to specify a data type, a data format or both.

7. The method of claim 1, further comprising broadcasting, by the processing system, a message indicative of support capability of the machine learning model to the first network entity, wherein the first network entity serves as a user device.

8. The method of claim 1, further comprising broadcasting, by the processing system, a message indicative of validity of the machine learning model to the first network entity, wherein the first network entity serves as a user device.

9. The method of claim 1, further comprising broadcasting, by the processing system, periodically, on an on-demand basis, or both, a message indicative of information relevant to the machine learning model to the first network entity, wherein the first network entity serves as a user device.

10. A device, comprising:

- a processing system including a processor; and
- a memory that stores executable instructions that, when executed by the processing system, facilitate performance of operations, the operations comprising:

determining a set of data collection parameters that are configurable to generate relevant data corresponding to network events, wherein the network events occur in real time or near real time in wireless communication networks;

collecting data based on the set of data collection parameters from a group of network entities operating in the wireless communication networks;

generating a training data based on the collected data; and

training a machine learning model based on the generated training data to learn the network events.

11. The device of claim 10, wherein the operations further comprise broadcasting a message signaling life cycle management procedures of the machine learning model to at least one of the group of network entities, wherein the machine learning model operates in the at least one of the group of network entities.

12. The device of claim 10, wherein the operations further comprise signaling, to at least one of the group of network entities, a set of machine learning model information including availability, update information, supported functionality, validity, or a combination thereof.

13. The device of claim 12, wherein the signaling further comprise signaling the set of machine learning model information periodically, in response to an on-demand request, or both.

14. The device of claim 10, wherein the operations further comprise performing an activation or a deactivation of the machine learning model based on a predetermined trigger.

15. The device of claim 10, wherein the operations further comprise:

- receiving an on-demand request for data collection, the on-demand request specifying a predetermined time duration and target measurement; and
- collecting another data based on the predetermined time duration and the target measurement specified in the on-demand request.

16. A non-transitory machine-readable medium, comprising executable instructions that, when executed by a processing system including a processor, facilitate performance of operations, the operations comprising:

- determining a set of data collection parameters that are configurable to collect data indicative of network events occurring in real time or near real time in wireless communication networks;

- receiving the collected data based on the set of data collection parameters from a group of network entities operating in the wireless communication networks;

- based on the received collected data, generating training data for a machine learning model deployed in the wireless communication networks; and

- broadcasting a message signaling life cycle management procedures of the machine learning model to at least one of the group of network entities, wherein the machine learning model is deployed in the at least one of the group of network entities.

17. The non-transitory machine-readable medium of claim 16, wherein the operations further comprise training the machine learning model based on the generated training data to learn the network events.

18. The non-transitory machine-readable medium of claim 16, wherein the broadcasting the message further comprises broadcasting the message periodically or on an on-demand basis.

19. The non-transitory machine-readable medium of claim 16, wherein the broadcasting the message signaling the life cycle management procedures further comprises facilitating an activation or deactivation, a switching, a

fallback of the machine learning model or a combination thereof, based on a predetermined trigger.

20. The non-transitory machine-readable medium of claim **16**, wherein the operations further comprise providing a condition-based life cycle management procedure to a user equipment by using a broadcast or a dedicated signaling.

* * * * *