United States Patent Application Publication	20250259722
Kind Code	A1
Publication Date	August 14, 2025
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SYSTEM AND METHOD FOR STEERING CARE PLAN ACTIONS BY DETECTING TONE, EMOTION, AND/OR HEALTH OUTCOME

Abstract

A method for electronically generating a care plan is disclosed herein. The method includes comparing a first data structure with a second data structure. The first data structure includes a set of health artifacts pertaining to a first condition of the patient. The second data structure pertains to the patient and the first condition of the patient, and the second data structure includes a subset of the set of the health artifacts. Responsive to the comparing, the method also includes generating the care plan comprising another subset of the set of health artifacts, and modifying the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.

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Appl. No.: **18/896591**

Filed: **September 25, 2024**

Related U.S. Application Data

parent US continuation 17768111 20220411 ABANDONED US continuation PCT/US2020/054613 20201007 child US 18896591
us-provisional-application US 62914227 20191011

Publication Classification

Int. Cl.: **G16H20/00** (20180101); **G16H10/60** (20180101)

U.S. Cl.:

CPC **G16H20/00** (20180101); **G16H10/60** (20180101);

Background/Summary

CROSS-REFERENCE TO RELATED APPLICATIONS [0001] This application is a continuation of U.S. application Ser. No. 17/768,111 filed on Apr. 11, 2022 titled “System and Method for Steering Care Plan Actions by Detecting Tone, Emotion, and/or Health Outcome,”, which is a U.S. National Phase Entry of PCT Application Serial No. PCT/US2020/054613 filed Oct. 7, 2020 and titled “System and Method for Steering Care Plan Actions by Detecting Tone, Emotion, and/or Health Outcome.” The PCT application claims the benefit of U.S. Provisional Application Ser. No. 62/914,227 filed Oct. 11, 2019 titled “System and Method for Steering Care Plan Actions by Detecting Tone, Emotion, and/or Health Outcome,” which provisional application is incorporated by reference herein as if reproduced in full below.

BACKGROUND
[0002] Population health management entails aggregating patient data across multiple health information technology resources, analyzing the data with reference to a single patient, and generating actionable items through which care providers can improve both clinical and financial outcomes. A population health management service seeks to improve the health outcomes of a group by improving clinical outcomes while lowering costs.
SUMMARY
[0003] Representative embodiments set forth herein disclose various techniques for enabling a system and method for steering care plan actions by detecting tone, emotion, and/or health/medical outcome.
[0004] In some embodiments, a method for electronically generating a care plan is disclosed herein. The method includes comparing a first data structure with a second data structure. The first data structure includes a set of health artifacts pertaining to a first condition of the patient. The second data structure pertains to the patient and the first condition of the patient, and the second data structure includes a subset of the set of the health artifacts. Responsive to the comparing, the method also includes generating the care plan comprising another subset of the set of health artifacts, and modifying the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.
[0005] In some embodiments, a system includes a memory storing instructions and a processor communicatively coupled with the memory. The processor may execute the instructions to perform one or more of the operations described above.
[0006] In some embodiments, a tangible, non-transitory computer-readable medium stores instructions. A process may execute the instructions to perform one or more of the operations described above.

Description

BRIEF DESCRIPTION OF THE DRAWINGS
[0007] For a detailed description of example embodiments, reference will now be made to the accompanying drawings in which:

[0008] FIG. **1** illustrates, in block diagram form, a system architecture **100** that can be configured to provide a population health management service, in accordance with various embodiments.
[0009] FIG. **2** shows additional details of a knowledge cloud, in accordance with various embodiments.
[0010] FIG. **3** shows an example subject matter ontology, in accordance with various embodiments.
[0011] FIG. **4** shows aspects of a conversation, in accordance with various embodiments.
[0012] FIG. **5** shows a cognitive map or "knowledge graph", in accordance with various embodiments.
[0013] FIG. **6** shows a method, in accordance with various embodiments.
[0014] FIGS. **7**A, **7**B, and **7**C show methods, in accordance with various embodiments.
[0015] FIGS. **8**A, **8**B, **8**C, and **8**D show aspects of a user interface, in accordance with various embodiments.
[0016] FIGS. **9**A and **9**B shows aspects of a conversational stream, in accordance with various embodiments.
[0017] FIG. **10** shows aspects of a conversational stream, in accordance with various embodiments.
[0018] FIG. **11** shows aspects of an action calendar, in accordance with various embodiments.
[0019] FIG. **12** shows aspects of a feed, in accordance with various embodiments.
[0020] FIG. **13** shows aspects of a hyper-local community, in accordance with various embodiments.
[0021] FIG. **14** illustrates a detailed view of a computing device that can represent the computing devices of FIG. **1** used to implement the various platforms and techniques described herein, according to some embodiments.
[0022] FIG. **15** shows a method, in accordance with various embodiments.
[0023] FIG. **16** shows a method, in accordance with various embodiments.
[0024] FIG. **17** shows a method, in accordance with various embodiments.
[0025] FIG. **18** shows a therapeutic paradigm logical framework, in accordance with various embodiments
[0026] FIG. **19** shows a method, in accordance with various embodiments.
[0027] FIG. **20** shows a paradigm logical framework, in accordance with various embodiments.
[0028] FIG. **21** shows a method for cognifying unstructured data, in accordance with various embodiments.
[0029] FIG. **22** shows a method for identifying missing information in a corpus of data, in accordance with various embodiments.
[0030] FIG. **23** shows a method for using feedback pertaining to the accuracy of cognified data to update an artificial intelligence engine, in accordance with various embodiments.
[0031] FIG. **24**A shows a block diagram for using a knowledge graph to generate possible health related information, in accordance with various embodiments.
[0032] FIG. **24**B shows a block diagram for using a logical structure to identify structural similarities with known predicates to generate cognified data, in accordance with various embodiments.
[0033] FIG. **25** shows a method for providing first information pertaining to a possible medical condition of a patient to a computing device, in accordance with various embodiments.
[0034] FIG. **26** shows a method for providing second and third information pertaining to a possible medical condition of a patient to a computing device, in accordance with various embodiments.
[0035] FIG. **27** shows a method for providing second information pertaining to a second possible medical condition of the patient, in accordance with various embodiments.
[0036] FIG. **28** shows an example of providing first information of a knowledge graph representing a possible medical condition, in accordance with various embodiments.
[0037] FIG. **29** shows an example of providing second information of the knowledge graph representing the possible medical condition, in accordance with various embodiments.

[0038] FIG. **30** shows an example of providing third information of the knowledge graph representing the possible medical condition, in accordance with various embodiments.
[0039] FIG. **31** shows a method for using cognified data to diagnose a patient, in accordance with various embodiments.
[0040] FIG. **32** shows a method for determining a severity of a medical condition based on a stage and a type of the medical condition, in accordance with various embodiments.
[0041] FIG. **33** shows an example of providing a home user interface for an autonomous multipurpose application, in accordance with various embodiments.
[0042] FIG. **34** shows an example of providing a user interface for selecting which person to schedule an appointment for, in accordance with various embodiments.
[0043] FIG. **35** shows an example of providing a user interface for selecting a specialty for an appointment, in accordance with various embodiments.
[0044] FIG. **36** shows an example of providing a user interface for displaying locations of people and recommended appointment times with the people, in accordance with various embodiments.
[0045] FIG. **37** shows an example of providing a user interface for presenting a profile of a person, in accordance with various embodiments.
[0046] FIG. **38** shows an example of providing a user interface that shows various payment options for the selected appointment, in accordance with various embodiments.
[0047] FIG. **39** shows an example of providing a user interface that shows messages pertaining to appointments for a user, in accordance with various embodiments.
[0048] FIG. **40**A shows an example of a cognitive intelligence platform receiving an image of an insurance card, in accordance with various embodiments.
[0049] FIG. **40**B shows an example of the cognitive intelligence platform extracting insurance plan information and causing it to be presented on a user device, in accordance with various embodiments.
[0050] FIG. **40**C shows an example of the cognitive intelligence platform extracting driver's license information and causing it to be presented on a user device, in accordance with various embodiments.
[0051] FIG. **40**D shows another example of the cognitive intelligence platform extracting insurance plan information and causing it to be presented on a user device, in accordance with various embodiments.
[0052] FIG. **41** shows an example of providing a user interface that shows an appointment has been electronically scheduled, in accordance with various embodiments.
[0053] FIG. **42** shows an example of providing a user interface that shows a user needs financial aid for a particular service, in accordance with various embodiments.
[0054] FIG. **43** shows a method for scheduling an appointment based on whether a user has elected to enable electronic scheduling, in accordance with various embodiments.
[0055] FIG. **44** shows a method for selecting a payment option between a co-pay cost and a self-pay cost, in accordance with various embodiments.
[0056] FIG. **45** shows providing various costs associated with a service to a computing device of a user, in accordance with various embodiments.
[0057] FIG. **46** shows an example of providing a user interface for checking-in a user for service, in accordance with various embodiments.
[0058] FIG. **47** shows an example of providing a user interface that shows additional required information is needed for a check-in document, in accordance with various embodiments.
[0059] FIG. **48**A shows an example of providing a user interface that shows check-in is complete, an estimated wait time, and curated content tailored for a condition of the user, in accordance with various embodiments.
[0060] FIG. **48**B shows an example of providing a user interface that shows an estimated wait time for a scheduled appointment, in accordance with various embodiments.

[0061] FIG. **49** shows an example of providing a user interface that allows searching for content and provides recommended content based on a condition of the user, in accordance with various embodiments.
[0062] FIG. **50** shows an example of providing a user interface to check symptoms, in accordance with various embodiments.
[0063] FIG. **51** shows an example of providing a user interface that provides details about symptoms that have been authored and reviewed by medical doctors, in accordance with various embodiments.
[0064] FIG. **52** shows a method of maintaining and transmitting check-in documents for a user to numerous different computing devices associated with people performing different specialties, in accordance with various embodiments.
[0065] FIG. **53** shows a method of determining whether the user has completed certain check-in documents required for a booked appointment, in accordance with various embodiments.
[0066] FIG. **54** shows a method of providing an estimated wait time to a computing device of the user, in accordance with various embodiments.
[0067] FIG. **55** shows an example of providing a user interface that includes options to select a condition, a number of areas of the condition to manage, and which areas of the condition to manage, in accordance with various embodiments.
[0068] FIG. **56** shows an example of a knowledge graph, a patient graph, and a care plan, in accordance with various embodiments.
[0069] FIGS. **57**A-**57**D show examples for generating a care plan using a knowledge graph and a patient graph, in accordance with various embodiments.
[0070] FIG. **58** shows a method for generating a care plan using a knowledge graph and a patient graph, in accordance with various embodiments.
[0071] FIG. **59** shows a method for updating a patient graph based on an interaction with a health artifact by the patient, in accordance with various embodiments.
[0072] FIG. **60**A-E show examples of modifying a care plan based on a detected emotion of the patient, a detected tone of the patient, a different medical outcome entered by a physician, or some combination thereof, in accordance with various embodiments.
[0073] FIG. **61** shows a method for modifying a care plan based on a detected emotion of the patient, a detected tone of the patient, a different medical outcome entered by a physician, or some combination thereof, in accordance with various embodiments.
[0074] FIG. **62** shows a method for using a net promoter score to update a machine learning model to output different health artifacts, in accordance with various embodiments.

NOTATION AND NOMENCLATURE

[0075] Various terms are used to refer to particular system components. Different companies may refer to a component by different names—this document does not intend to distinguish between components that differ in name but not function. In the following discussion and in the claims, the terms “including” and “comprising” are used in an open-ended fashion, and thus should be interpreted to mean “including, but not limited to . . . ” Also, the term “couple” or “couples” is intended to mean either an indirect or direct connection. Thus, if a first device couples to a second device, that connection may be through a direct connection or through an indirect connection via other devices and connections.

DETAILED DESCRIPTION

[0076] The following discussion is directed to various embodiments of the invention. Although one or more of these embodiments may be preferred, the embodiments disclosed should not be interpreted, or otherwise used, as limiting the scope of the disclosure, including the claims. In addition, one skilled in the art will understand that the following description has broad application, and the discussion of any embodiment is meant only to be exemplary of that embodiment, and not intended to intimate that the scope of the disclosure, including the claims, is limited to that

embodiment.
[0077] According to some embodiments, a cognitive intelligence platform integrates and consolidates data from various sources and entities and provides a population health management service. The cognitive intelligence platform has the ability to extract concepts, relationships, and draw conclusions from a given text posed in natural language (e.g., a passage, a sentence, a phrase, and a question) by performing conversational analysis which includes analyzing conversational context. For example, the cognitive intelligence platform has the ability to identify the relevance of a posed question to another question.
[0078] The benefits provided by the cognitive intelligence platform, in the context of healthcare, include freeing up physicians from focusing on day to day population health management. Thus a physician can focus on her core competency-which includes disease/risk diagnosis and prognosis and patient care. The cognitive intelligence platform provides the functionality of a health coach and includes a physician's directions in accordance with the medical community's recommended care protocols and also builds a systemic knowledge base for health management.
[0079] Accordingly, the cognitive intelligence platform implements an intuitive conversational cognitive agent that engages in a question and answering system that is human-like in tone and response. The described cognitive intelligence platform endeavors to compassionately solve goals, questions and challenges.
[0080] In addition, physicians often generate patient notes before, during, and/or after consultation with a patient. The patient notes may be included in an electronic medical record (EMR). When a patient returns for a subsequent visit, the physician may review numerous EMRs for the patient. Such a review process may be time consuming and inefficient. Insights may be hidden in the various EMRs and may result in the physician making an incorrect diagnosis. Further, it may involve the physician accessing numerous screens and performing multiple queries on a database to obtain the various EMRs. As a result, the computing device of the physician may waste computing resources by loading various screens and sending requests for EMR data to a server. The server that receives the requests may also waste computing resources by processing the numerous requests and transmitting numerous responses. In addition, network resources may be wasted by transmitting the requests and responses between the server and the client.
[0081] Accordingly, some embodiments of the present disclosure address the issues of reviewing the EMRs, by cognifying unstructured data. Unstructured data may include patient notes entered into one or more EMRs by a physician. The patient notes may explain symptoms described by the patient or detected by the physician, vital signs, recommended treatment, risks, prior health conditions, familial health history, and the like. The patient notes may include numerous strings of characters arranged into sentences. The sentences may be organized in one or more paragraphs. The sentences may be parsed and indicia may be identified. The indicia may include predicates, objectives, nouns, verbs, cardinals, ranges, keywords, phrases, numbers, concepts, or some combination thereof.
[0082] The indicia may be compared to one or more knowledge graphs that each represents health related information (e.g., a disease) and various characteristics of the health related information. The knowledge graph may also include how the various diseases are related to one another (e.g., bronchitis can lead to pneumonia). The knowledge graph may represent a model that includes individual elements (nodes) and predicates that describe properties and/or relationships between those individual elements. A logical structure (e.g., Nth order logic) may underlie the knowledge graph that uses the predicates to connect various individual elements. The knowledge graph and the logical structure may combine to form a language that recites facts, concepts, correlations, conclusions, propositions, and the like. The knowledge graph and the logical structure may be generated and updated continuously or on a periodic basis by an artificial intelligence engine with evidence-based guidelines, physician research, patient notes in EMRs, physician feedback, and so forth. The predicates and individual elements may be generated based on data that is input to the

artificial intelligence engine. The data may include evidence-based guidelines that is obtained from a trusted source, such as a physician. The artificial intelligence engine may continuously learn based on input data (e.g., evidence-based guidelines, clinical trials, physician research, electronic medical records, etc.) and modify the individual elements and predicates.

[0083] For example, a physician may indicate that if a person has a blood sugar level of a certain amount and various other symptoms (e.g., unexplained weight loss, sweating, etc.), then that person has type 2 diabetes mellitus. Such a conclusion may be modeled in the knowledge graph and the logical structure as "Type 2 diabetes mellitus has symptoms of a blood sugar level of the certain amount and various other symptoms," where "Type 2 diabetes mellitus," "a blood sugar level of the certain amount," and "various other symptoms" are individual elements in the knowledge graph, and "has symptoms of" is a predicate of the logical structure that relates the individual element "Type 2 diabetes mellitus" to the individual elements of "a blood sugar level of the certain amount" and "various other symptoms".

[0084] The indicia extracted from the unstructured data may be correlated with one or more closely matching knowledge graphs by comparing similarities between the indicia and the individual elements. Tags related to possible health related information may be generated and associated with the indicia in the unstructured data. For example, the tags may specify "A leads to B" (where A is a health related information and B is another health related information), "B causes C" (where C is yet another health related information), "C has complications of D" (where D is yet another health related information), and so forth. These tags associated with the indicia may be correlated with the logical structure (e.g., predicates of the logical structure) based on structural similarity to generate cognified data. For example, if a person exhibits certain symptoms and has certain laboratory tests performed, then that person may have a certain medical condition (e.g., type 2 diabetes mellitus) that is identified in the knowledge graphs using the logical structures.

[0085] A pattern may be detected by identifying structural similarities between the tags and the logical structure in order to generate the cognified data. Cognification may refer to instilling intelligence into something. In the present disclosure, unstructured data may be cognified into cognified data by instilling intelligence into the unstructured data using the knowledge graph and the logical structure. The cognified data may include a summary of a health related condition of a patient, where the summary includes insights, conclusions, recommendations, identified gaps (e.g., in treatment, risk, quality of care, guidelines, etc.), and so forth.

[0086] The cognified data may be presented on a computing device of a physician. Instead of reading pages and pages of digital medical charts (EMRs) for a patient, the physician may read the cognified data that presents pointed summarized information that can be utilized to more efficiently and effectively treat the patient. As a result, computing resources may be saved by preventing numerous searches for EMRs and preventing accessing numerous screens displaying the EMRs. In some embodiments, the physician may submit feedback pertaining to whether or not the cognified data is accurate for the patient. The feedback may be used to update the artificial intelligence engine that uses the knowledge graph and logical structure to generate the cognified data.

[0087] In some embodiments, the cognified data may be used to diagnose a medical condition of the patient. For example, the medical condition may be diagnosed if a threshold criteria is satisfied. The threshold criteria may include matching a certain number of predicates and tags for a particular medical condition represented by a particular knowledge graph. The computing device of the physician and/or the patient may present the diagnosis and a degree of certainty based on the threshold criteria. In some embodiments, the physician may submit feedback pertaining to whether or not the diagnosis is accurate for the patient. The feedback may be used to update the artificial intelligence engine that uses the knowledge graph and logical structure to generate the diagnosis using the cognified data.

[0088] Further, patients may be inundated with information about a particular medical condition with which they are diagnosed and/or inquiring about. The information may not be relevant to a

particular stage of the medical condition. The amount of information may waste memory resources of the computing device of the patient. Also, the user may have a bad experience using the computing device due to the overwhelming amount of information.

[0089] In some embodiments, user experience of using a computing device may be enhanced by running an application that performs various techniques described herein. The user may be interacting with the cognitive agent and the cognitive agent may be steering the conversation as described herein. In some embodiments, the cognitive agent may provide recommendations based on the text entered by the user, and/or patient notes in EMRs, which may be transformed into cognified data. The application may present health related information, such as the cognified data, pertaining to the medical condition to the computing device of the patient and/or the physician.

[0090] Instead of overwhelming the patient with massive amounts of information about the medical condition, the distribution of information may be regulated to the computing device of the patient and/or the physician. For example, if the patient is diagnosed as having type 2 diabetes mellitus, a controlled traversing of the knowledge graph associated with type 2 diabetes mellitus may be performed to provide information to the patient. The traversal may begin at a root node of the knowledge graph and first health related information may be provided to the computing device of the patient at a first time. The first health related information may pertain to a name of the medical condition, a definition of the possible medical condition, or some combination thereof. At a second time, health related information associated with a second node of the knowledge graph may be provided to the computing device of the patient. The second health related information may pertain to how the medical condition affects people, signs and symptoms of the medical condition, a way to treat the medical condition, complications of the medical condition, a progression of the medical condition, or some combination thereof. The health related information associated with the remaining nodes in the knowledge graph may be distributed to the computing device of the patient at different respective times. In some embodiments, the health related information to be provided and/or the times at which the health related information is provided may be selected based on relevancy to a stage of the medical condition of the patient.

[0091] In other scenarios, users (also referred to as patients herein) may use various computing devices (e.g., smartphone, tablet, laptop, etc.) to schedule an appointment with a person (also referred to as care providers herein) having a particular specialty to perform a service. For example, a patient may schedule appointments with care providers to provide one or more services to the patient. A patient may call an office where the care provider having a specialty works and speak to a person who finds an available appointment to book for the care provider and the patient. To book an appointment with another care provider having a different specialty, the patient may call the office of the other care provider having the different specialty to book an available appointment. Further, to book an appointment with a care provider for a dependent (e.g., child), the parent/guardian may contact yet another office where a care provider having yet another specialty (e.g., pediatrician) works to book an appointment. In some instances, the patient may access multiple different websites associated with the care providers to attempt to schedule an appointment. This is inconvenient for the patient and wastes resources by making multiple phone calls or accessing multiple different websites. Switching between websites to find contact information for people having different specialties may cause undesirable network, computing, and/or memory usage to occur. Additionally, typical software applications do not include functionality for scheduling appointments for an entire family (e.g., primary, spouse, dependents (children, senior citizens)) covered by an insurance plan, and/or functionality for scheduling multiple appointments for the same patient and/or different patients.

[0092] When the patient arrives for the scheduled appointments, the patient typically has to fill out paper check-in documents at each office. Even when the information requested by the check-in documents is redundant, such as medical history information, medication information, etc., various offices still request the same information. Part of the issue is a lack of interoperability of electronic

medical records systems. Also, when a computing device is used to complete the check-in documents, the check-in documents are not shared with other systems associated with other specialties, and the user may have to reenter their information using a computing device of another system associated with the other specialties. As such, computing resources of the computing devices may be wasted by running an application to enable entry of information into the check-in documents, instead of just sharing the already completed check-in documents with requesting systems.

[0093] Once check-in is complete, the patient may be presented with paper reading materials in a waiting room. The reading materials may include information (e.g., symptoms, causes, treatments, etc.) pertaining to various different medical conditions. It can oftentimes be overwhelming to a patient to be presented with too much information, especially when the information does not pertain to the condition or conditions for which the patient is seeking treatment. Further, even if the patient knows what he or she is looking for, searching for the paper reading material is inefficient. To that end, even if the user finds reading material that discusses a desired topic, there typically is not a guarantee the reading material was authored/reviewed by a person having proper credentials (e.g., a medical doctor). Educating the patient with pertinent curated content that is tailored for the patient is desired.

[0094] Accordingly, some embodiments of the present disclosure address the above-identified issues, among other things. For example, an autonomous multipurpose application may execute in a cognitive intelligence platform. In some embodiments, the autonomous multipurpose application may be implemented as one or more application programming interfaces (API) executing via one or more computing devices (e.g., servers), as described in more detail below. The term "autonomous" used in conjunction with the "multipurpose application" may refer to the multipurpose application executing a set of operations on behalf of a person or another application with some degree of independence or autonomy in an intelligent manner using knowledge or representation of a user's goals or desires. The terms "autonomous multipurpose application" and "cognitive agent" may be used interchangeably herein.

[0095] In some embodiments, the autonomous multipurpose application may present different user interfaces based on a role associated with a person that logs into the autonomous multipurpose application. The various roles may include a medical personnel (e.g., medical doctor, physician, nurse, dentist, optometrist, psychiatrist, behavioral specialist, physician assistant, and the like), an administrator, a patient/user, and so forth. The user interface presented on a computing device when a person having the medical personnel role is logged in may be referred to as "clinic viewer" herein. The user interface presented on a computing device when a person having the administrator role is logged in may be referred to as "administrator viewer" herein. The user interface presented on a computing device when a person having the patient/user role may be referred to as "patient viewer" herein.

[0096] The autonomous multipurpose application may perform numerous operations pertaining to scheduling appointments for patients, checking-in patients for scheduled appointments, educating the patients about medical conditions, and/or searching for content based on search queries, among other things. For scheduling purposes, the autonomous multipurpose application may be communicatively coupled with computing devices of care providers (e.g., medical personnel) and/or electronic medical record (EMR) systems used by the care providers (e.g., medical personnel). These computing devices and/or electronic medical record systems may execute patient management systems or scheduling management systems that maintain schedules of appointments for the care providers. For example, a schedule for a care provider may show which appointments are scheduled or booked and which appointments are available by date and time.

[0097] The autonomous multipurpose application may obtain the schedules for people having a desired specialty within a certain geographic location (e.g., within a radius of a geolocation of a computing device of the user, within a radius of an entered address, etc.). A user may elect to

enable electronic scheduling. If an available appointment is found within the certain geographic region, and the user is available at the same date and time as the available appointment, the autonomous multipurpose application may electronically schedule the available appointment as a booked appointment. If the user has not enabled electronic scheduling, the autonomous multipurpose application may recommend one or more available appointments to the computing device of the user for presentation.

[0098] The autonomous multipurpose application may enable a user to schedule numerous appointments for himself or herself with people having different specialties via a single user interface. For example, the specialties may include a medical doctor (physician), a dentist, an optometrist, a physician's assistant, a chiropractor, a behavioral specialist, a lab technician, a masseuse, a barber, an orthodontist, a dermatologist, and the like. Also, the autonomous multipurpose application may enable the user to schedule appointments for dependents (e.g., children, spouse, senior citizen, etc.) of an insurance plan.

[0099] In some embodiments, the autonomous multipurpose application may provide service cost transparency. For example, the autonomous multipurpose application may use the insurance plan information extracted from an insurance card and/or provided by a user to determine what a service may cost. The autonomous multipurpose application may determine a co-pay cost based on the deductible of the insurance plan. The autonomous multipurpose application may determine a self-pay cost without considering the insurance plan. The co-pay cost and the self-pay cost may be presented on the computing device of the user, administrator, or person having a specialty. In some embodiments, if electronic scheduling is enabled, the autonomous multipurpose application may electronically select the cost that is the lowest.

[0100] Further, the autonomous multipurpose application may function as a centralized manager and repository for documents pertaining to the user and the dependents of the user. For example, when a user checks-in using a computing device (e.g., kiosk) executing the autonomous multipurpose application at a clinic, check-in documents pertaining to the user stored in a database may be checked to determine whether the check-in documents are complete. The check-in documents may refer to consent forms, medical history documents, health information release authorization forms, new patient sheets, massage client intake forms, mental health intake forms, consent treatment for minor child forms, doctor referral forms, adult health history forms, school physical forms, insurance verification sheets, medical reports, therapy intake forms, initial exam reports, pain assessment sheets, and the like. In some embodiments, the autonomous multipurpose application may communicate with external systems, such as EMR systems, to request the documents for the user from those systems. For example, if the user checked-in for another appointment with a different physician, the user may have already completed the various check-in documents and the autonomous multipurpose application may retrieve those completed check-in documents and store them for future reference. The autonomous multipurpose application may transmit the completed check-in documents to the EMR system associated with the person with which the user has an appointment.

[0101] If the check-in documents are partially complete, the autonomous multipurpose application may cause the portions of information that are missing to be presented for completion. If the check-in documents are incomplete, the autonomous multipurpose application may cause the check-in documents to be presented on a computing device for completion by the user, an administrator, a person having a specialty, or the like.

[0102] The autonomous multipurpose application may also manage and store other information for the users. For example, the user may capture an image of their driver's license, insurance card, and the like, and transmit the image to the autonomous multipurpose application. The autonomous multipurpose application may analyze the image (e.g., using machine learning and/or optical character recognition) to extract information from the image. For example, the autonomous multipurpose application may extract a picture of the user from a driver's license, a name of the

user, a birthdate of the user, an address of the user, an identification number, an insurance plan number, a type of insurance, an expiration date of the user's driver's license, an expiration date of the user's insurance plan, and the like. The autonomous multipurpose application may electronically fill information in corresponding documents based on the extracted information. Further, the autonomous multipurpose application may perform logic based on the extracted information. For example, if the user's insurance is about to expire, the autonomous multipurpose application may transmit a message (e.g., email, text message, phone call, onscreen notification, etc.) to the user to renew their insurance. Similar types of information may be managed and stored for each person in a family. The information may be disbursed to a requesting client, such as an EMR system used by an entity at which the users make appointments.

[0103] The autonomous multipurpose application may communicate with a knowledge cloud that includes knowledge graphs that each pertain to a respective medical condition. For example, each knowledge graph may include individual elements (e.g., health artifacts) and predicates that describe relationships between the individual elements in a logical structure. Each knowledge graph may include nodes representing the individual elements and branches representing the predicates that connect the nodes. Each knowledge graph may begin at a root node that includes a type or name of the medical condition, for example. One knowledge graph may include a root node representing “Diabetes”. A predicate may represent “is caused by” branch that connects to another node “high blood sugar”. The logical structure may be formulated as “Diabetes is caused by high blood sugar”.

[0104] When a user successfully checks-in for a scheduled appointment, the autonomous multipurpose application may access the knowledge cloud to obtain curated content pertaining to one or more conditions of the user. For example, the user may specify the condition for which the user is seeking treatment, and educational curated content about that condition may be recommended and/or provided to the computing device of the user. The autonomous multipurpose application may also recommend other curated content to the user for the conditions of the user that are known by the autonomous multipurpose application. Each time a user has an appointment, the autonomous multipurpose application may update information pertaining to the user to keep knowledge about the user up to date.

[0105] In addition, when the user is checked-in, a wait time estimator model may be used by the autonomous multipurpose application to provide an estimated wait time. For example, the wait time estimator may be a machine learning model that is trained using data representing an average amount of time it takes a person having a specialty to perform a service. The training data may be specific for each different person and the amount of time it takes that person to perform the service. The wait time estimator may use training data pertaining to each patient. For example, if John Smith is at an appointment in the doctor's office immediately before Jane Doe, the average time that John Smith stays in the office may be used to estimate the wait time for Jane Doe. The wait times from different offices and/or clinics may be aggregated for each specialty in that office and/or for each person having the specialties to perform the service associated with the specialties.

[0106] Various timestamps associated with interactions between the user and the person having the specialty may be obtained from a system (e.g., EMR) used by the person having the specialty. For example, a timestamp of when the user checked-in for a scheduled appointment may be obtained, a timestamp of how long it took for the user to be called back to the doctor's office may be obtained, a timestamp of how long the user waited in the doctor's office prior to the doctor entering, a timestamp of any patient notes made by the doctor, a timestamp of any patient notes made by a nurse, a timestamp of when the doctor leaves after performing a service, a timestamp of when the user pays, or some combination thereof. The timestamps may be used to estimate wait times for users that have appointments scheduled with that doctor.

[0107] The autonomous multipurpose application may provide natural language searching for content. For example, the user may search “information about Diabetes” and the autonomous

multipurpose application may return curated content pertaining to Diabetes to the computing device of the user.

[0108] The disclosed autonomous multipurpose application may provide an enhanced experience for users by improving scheduling, check-in, wait time estimation, cost transparency, and/or content distribution, among other things. The autonomous multipurpose application may use artificial intelligence to make decisions and perform actions.

[0109] In addition, the cognitive intelligence platform may use a knowledge graph pertaining to a condition of a user and a data structure (e.g., a patient graph) corresponding to the condition and the user to electronically generate a care plan for the condition of the user. The patient graph may include elements (e.g., health artifacts) and branches representing relationships between the elements. The elements may be represented as nodes in the patient graph. The elements may represent interactions and/or actions the user has had and/or performed pertaining to the condition. For example, if the condition is diabetes and the user has already performed a blood glucose test, then the user may have a patient graph corresponding to diabetes that includes an element for the blood glucose test. The element may include one or more associated information, such as a timestamp of when the blood glucose test was taken, if it was performed at-home or at a care provider, a result of the blood glucose test, and so forth.

[0110] The autonomous multipurpose application may cause the patient viewer to be presented on the computing device of the user, and the patient viewer may present the various conditions of the user. Further, the patient viewer may ask the user to specify a number of areas of the condition the user would like to manage, and to select which areas of the condition the user would like to manage.

[0111] The patient graph for the condition of the user may be compared (e.g., projected on) to the knowledge graph for the condition of the user to generate a care plan. The cognitive intelligence platform may generate the care plan based on the areas of the condition the user specified to manage, based on areas of the condition on which the user has not taken action and/or interacted with in view of the knowledge graph and patient graph, based on a detected emotion of the user, based on a detected tone of the user, based on a medical outcome selected by a medical personnel, or some combination thereof. For example, the cognitive intelligence platform may determine that the user currently is prescribed medication A for diabetes based on the user's patient graph for diabetes, but medication A is ineffective for the user. The cognitive intelligence platform may compare the patient graph to the knowledge graph pertaining to diabetes to determine that medication B can be prescribed to treat diabetes for the user. The care plan may include an action instruction that instructs the medical personnel to prescribe medication B and/or discuss information pertaining to medication A and/or medication B. The care plan may be transmitted to the user device for presentation in the patient viewer, the clinic viewer, and/or the administrator viewer.

[0112] The patient graph for each condition may also include an engagement profile that may be used to determine a compliance of the user with the care plan. The engagement profile may store information at a meta data level that corresponds to the actions and/or interactions the user performs pertaining to the care plan for the condition. In some embodiments, activity of the user on the computing device may be tracked; medical records may be obtained from EMR systems, claims systems, clinical systems, and the like; and so forth. For example, if the care plan recommends the user read a certain article pertaining to diabetes, and the user selects the article, the engagement profile may store information related to the user selecting the article, how long the user read the article, if the user finished the article, and so forth. Further, if the medical records indicate the user had a blood glucose test performed, the engagement profile may store information pertaining to the blood glucose test being performed.

[0113] The patient graph for the diabetes of the user may be updated based on the information stored in the engagement profile. For example, if information in the engagement profile indicates

the user completes performance of a blood glucose test, an element pertaining to the blood glucose test may be added to a section of the patient graph of the user corresponding to diabetes. In some embodiments, certain conditions may specify the same elements as each other. For example, two conditions may include knowledge graphs that both include elements for testing for the condition using a blood glucose test. If the patient performs the blood glucose test for one of the conditions, the patient graphs for both conditions may be updated to include the information for the blood glucose test at the appropriate elements. As a result, if a knowledge graph for one condition includes an element for a test, and the user has already performed the test for another condition, as represented in the patient graph for the other condition, the cognitive intelligence platform may not include an action instruction to perform the test in the care plan for the user for the one condition. In this way, the care plans may be not include redundant data and/or action instructions.

[0114] In some embodiments, the patient graph may represent a checklist of items (e.g., elements, actions, interactions, content, etc.) pertaining to the condition that the user performed. The knowledge graph may represent a superset of items pertaining to the condition, and if the user complies with the superset of items (e.g., completes a care plan for a condition), the user may be managing the condition in a desired manner (e.g., the user is taking medications on a specified basis, the values of certain tests for the user are within a desired range, the user has been informed by the recommended content, etc.). The compliance with the care plan may be determined based on the engagement profile and/or the patient graph.

[0115] In some embodiments, the patient graph for a condition may be compared (e.g., projected on) to the knowledge graph for the condition, and if the patient graph includes each element of the knowledge graph, then a determination may be made that the user is managing the condition in a desired manner. In some embodiments, a notification may be presented on the patient viewer, the clinic viewer, and/or the administrator viewer indicating the same. If some of the elements of the knowledge graph are missing in the patient graph, the cognitive intelligence platform may provide a care plan including action instructions pertaining to those missing elements. Based on the engagement profile, if certain elements are partially completed, performed, and/or interacted with, the cognitive intelligence platform may provide a care plan including action instructions pertaining to those partially completed, performed, and/or interact with elements.

[0116] In some embodiments, an emotion of the user, a tone of the user, and/or a medical outcome desired by a medical personnel may be used to modify the care plan presented to the user. For example, data (e.g., video, image, text, etc.) may be received by the cognitive intelligence platform from a computing device of the user while the user is interacting with the patient viewer and/or interacting with the computing device of the user. The cognitive intelligence platform may perform certain emotion detecting and/or tone detecting techniques using the data. For example, facial recognition techniques may be performed to determine an emotion the user is experiencing. Such a determination may be made in response to the care plan presented to the user, content presented to the user, responses provided by the cognitive intelligence platform, or the like. Further, a tone and/or emotion of the user may be determined using text input by the user while interacting with the patient viewer and/or interacting with the computing device of the user. In addition, the cognitive intelligence platform may receive a desired medical outcome input by a medical personnel using the clinic viewer.

[0117] The cognitive intelligence platform may modify the care plan based on the detected emotion, detected tone, and/or the desired medical outcome. The modified care plan may be presented in the patient viewer, the clinic viewer, and/or the administrator viewer.

[0118] The described methods and systems are described as occurring in the healthcare space, though other areas are also contemplated, such as finance, career, etc.

[0119] FIG. **1** shows a system architecture **100** that can be configured to provide a population health management service, in accordance with various embodiments. Specifically, FIG. **1** illustrates a high-level overview of an overall architecture that includes a cognitive intelligence

platform **102** communicably coupled to a user device **104**. The cognitive intelligence platform **102** includes several computing devices, where each computing device, respectively, includes at least one processor, at least one memory, and at least one storage (e.g., a hard drive, a solid-state storage device, a mass storage device, and a remote storage device). The individual computing devices can represent any form of a computing device such as a desktop computing device, a rack-mounted computing device, and a server device. The foregoing example computing devices are not meant to be limiting. On the contrary, individual computing devices implementing the cognitive intelligence platform **102** can represent any form of computing device without departing from the scope of this disclosure.

[0120] The several computing devices work in conjunction to implement components of the cognitive intelligence platform **102** including: a knowledge cloud **106**; a critical thinking engine **108**; a natural language database **122**; and a cognitive agent **110**. The cognitive intelligence platform **102** is not limited to implementing only these components, or in the manner described in FIG. **1**. That is, other system architectures can be implemented, with different or additional components, without departing from the scope of this disclosure. The example system architecture **100** illustrates one way to implement the methods and techniques described herein.

[0121] The knowledge cloud **106** represents a set of instructions executing within the cognitive intelligence platform **102** that implement a database configured to receive inputs from several sources and entities. For example, some of the sources and entities include a service provider **112**, a facility **114**, and a microsurvey **116**—each described further below.

[0122] The critical thinking engine **108** represents a set of instructions executing within the cognitive intelligence platform **102** that execute tasks using artificial intelligence, such as recognizing and interpreting natural language (e.g., performing conversational analysis), and making decisions in a linear manner (e.g., in a manner similar to how the human left brain processes information). Specifically, an ability of the cognitive intelligence platform **102** to understand natural language is powered by the critical thinking engine **108**. In various embodiments, the critical thinking engine **108** includes a natural language database **122**. The natural language database **122** includes data curated over at least thirty years by linguists and computer data scientists, including data related to speech patterns, speech equivalents, and algorithms directed to parsing sentence structure.

[0123] Furthermore, the critical thinking engine **108** is configured to deduce causal relationships given a particular set of data, where the critical thinking engine **108** is capable of taking the individual data in the particular set, arranging the individual data in a logical order, deducing a causal relationship between each of the data, and drawing a conclusion. The ability to deduce a causal relationship and draw a conclusion (referred to herein as a “causal” analysis) is in direct contrast to other implementations of artificial intelligence that mimic the human left brain processes. For example, the other implementations can take the individual data and analyze the data to deduce properties of the data or statistics associated with the data (referred to herein as an “analytical” analysis). However, these other implementations are unable to perform a causal analysis—that is, deduce a causal relationship and draw a conclusion from the particular set of data. As described further below—the critical thinking engine **108** is capable of performing both types of analysis: causal and analytical.

[0124] In some embodiments, the critical thinking engine **108** includes an artificial intelligence engine **109** (“AI Engine” in FIG. **1**) that uses one or more machine learning models. The one or more machine learning models may be generated by a training engine and may be implemented in computer instructions that are executable by one or more processing device of the training engine, the artificial intelligence engine **109**, another server, and/or the user device **104**. To generate the one or more machine learning models, the training engine may train, test, and validate the one or more machine learning models. The training engine may be a rackmount server, a router computer, a personal computer, a portable digital assistant, a smartphone, a laptop computer, a tablet

computer, a camera, a video camera, a netbook, a desktop computer, a media center, or any combination of the above. The one or more machine learning models may refer to model artifacts that are created by the training engine using training data that includes training inputs and corresponding target outputs. The training engine may find patterns in the training data that map the training input to the target output, and generate the machine learning models that capture these patterns.

[0125] The one or more machine learning models may be trained to generate one or more knowledge graphs each pertaining to a particular medical condition. The knowledge graphs may include individual elements (nodes) that are linked via predicates of a logical structure. The logical structure may use any suitable order of logic (e.g., higher order logic and/or Nth order logic). Higher order logic may be used to admit quantification over sets that are nested arbitrarily deep. Higher order logic may refer to a union of first-, second-, third, . . . , Nth order logic. Clinical-based evidence, clinical trials, physician research, and the like that includes various information (e.g., knowledge) pertaining to different medical conditions may be input as training data to the one or more machine learning models. The information may pertain to facts, properties, attributes, concepts, conclusions, risks, correlations, complications, etc. of the medical conditions. Keywords, phrases, sentences, cardinals, numbers, values, objectives, nouns, verbs, concepts, and so forth may be specified (e.g., labeled) in the information such that the machine learning models learn which ones are associated with the medical conditions. The information may specify predicates that correlates the information in a logical structure such that the machine learning models learn the logical structure associated with the medical conditions.

[0126] In some embodiments, the one or more machine learning models may be trained to transform input unstructured data (e.g., patient notes) into cognified data using the knowledge graph and the logical structure. The machine learning models may identify indicia in the unstructured data and compare the indicia to the knowledge graphs to generate possible health related information (e.g., tags) pertaining to the patient. The possible health related information may be associated with the indicia in the unstructured data. The one or more machine learning models may also identify, using the logical structure, a structural similarity of the possible health related information and a known predicate in the logical structure. The structural similarity between the possible health related information and the known predicate may enable identifying a pattern (e.g., treatment patterns, education and content patterns, order patterns, referral patterns, quality of care patterns, risk adjustment patterns, etc.). The one or more machine learning models may generate the cognified data based on the structural similarity and/or the pattern identified. Accordingly, the machine learning models may use a combination of knowledge graphs, logical structures, structural similarity comparison mechanisms, and/or pattern recognition to generate the cognified data. The cognified data may be output by the one or more trained machine learning models.

[0127] The cognified data may provide a summary of the medical condition of the patient. A diagnosis of the patient may be generated based on the cognified data. The summary of the medical condition may include one or more insights not present in the unstructured data. The summary may identify gaps in the unstructured data, such as treatment gaps (e.g., should prescribe medication, should provide different medication, should change dosage of medication, etc.), risk gaps (e.g., the patient is at risk for cancer based on familial history and certain lifestyle behaviors), quality of care gaps (e.g., need to check-in with the patient more frequently), and so forth. The summary of the medical condition may include one or more conclusions, recommendations, complications, risks, statements, causes, symptoms, etc. pertaining to the medical condition. In some embodiments, the summary of the medical condition may indicate another medical condition that the medical condition can lead to. Accordingly, the cognified data represents intelligence, knowledge, and logic cognified from unstructured data.

[0128] In some embodiments, the cognified data may be reviewed by physicians and the physicians

may provide feedback pertaining to whether or not the cognified data is accurate. Also, the physicians may provide feedback pertaining to whether or not the diagnosis generated using the cognified data is accurate. This feedback may be used to update the one or more machine learning models to improve their accuracy.

[0129] The AI engine **109** may include machine learning models that are trained to schedule appointments for users, recommend appointments to users, determine costs of services, manage documents for users, extract data from images, provide curated content tailored for users, estimate wait times, perform natural language searching of curated content, and so forth.

[0130] The cognitive agent **110** represents a set of instructions executing within the cognitive intelligence platform **102** that implement a client-facing component of the cognitive intelligence platform **102**. The cognitive agent **110** may be referred to as the autonomous multipurpose application interchangeably herein. The cognitive agent **110** is an interface between the cognitive intelligence platform **102** and the user device **104**. And in some embodiments, the cognitive agent **110** includes a conversation orchestrator **124** that determines pieces of communication that are presented to the user device **104** (and the user). When a user of the user device **104** interacts with the cognitive intelligence platform **102**, the user interacts with the cognitive agent **110**. In some embodiments, the user of the user device **104** may be a patient. The several references herein, to the cognitive agent **110** performing a method, can implicate actions performed by the critical thinking engine **108**, which accesses data in the knowledge cloud **106** and the natural language database **122**.

[0131] Various user interfaces may be provided to computing devices communicating with the cognitive agent **110** executing in the cognitive intelligence platform **102**. The user interfaces may be presented in a standalone application executing on the devices or in a web browser as website pages. In some embodiments, the cognitive agent **110** may be installed on a device of the user, the service provider **112**, and/or the facility **114**. In some embodiments, the devices of the user, the service provider **112**, and/or the facility **114** may communicate with cognitive intelligence platform **102** in a client-server architecture. In some embodiments, the cognitive agent **110** may be implemented as computer instructions as an application programming interface.

[0132] In various embodiments, the several computing devices executing within the cognitive intelligence platform are communicably coupled by way of a network/bus interface. Furthermore, the various components (e.g., the knowledge cloud **106**, the critical thinking engine **108**, and the cognitive agent **110**), are communicably coupled by one or more inter-host communication protocols **118**. In one example, the knowledge cloud **106** is implemented using a first computing device, the critical thinking engine **108** is implemented using a second computing device, and the cognitive agent **110** is implemented using a third computing device, where each of the computing devices are coupled by way of the inter-host communication protocol **118**. Although in this example, the individual components are described as executing on separate computing devices this example is not meant to be limiting, the components can be implemented on the same computing device, or partially on the same computing device, without departing from the scope of this disclosure.

[0133] The user device **104** represents any form of a computing device, or network of computing devices, e.g., a personal computing device, a smart phone, a tablet, a wearable computing device, a notebook computer, a media player device, and a desktop computing device. The user device **104** includes a processor, at least one memory, and at least one storage. A user uses the user device **104** to input a given text posed in natural language (e.g., typed on a physical keyboard, spoken into a microphone, typed on a touch screen, or combinations thereof) and interacts with the cognitive intelligence platform **102**, by way of the cognitive agent **110**.

[0134] The architecture **100** includes a network **120** that communicatively couples various devices, including the cognitive intelligence platform **102** and the user device **104**. The network **120** can include local area network (LAN) and wide area networks (WAN). The network **102** can include

wired technologies (e.g., Ethernet®) and wireless technologies (e.g., Wi-Fi®, code division multiple access (CDMA), global system for mobile (GSM), universal mobile telephone service (UMTS), Bluetooth®, and ZigBee®. For example, the user device **104** can use a wired connection or a wireless technology (e.g., Wi-Fi®) to transmit and receive data over the network **120**.

[0135] Still referring to FIG. **1**, the knowledge cloud **106** is configured to receive data from various sources and entities and integrate the data in a database. An example source that provides data to the knowledge could **106** is the service provider **112**, an entity that provides a type of service to a user. For example, the service provider **112** can be a health service provider (e.g., a doctor's office, a physical therapist's office, a nurse's office, or a clinical social worker's office), and a financial service provider (e.g., an accountant's office). For purposes of this discussion, the cognitive intelligence platform **102** provides services in the health industry, thus the examples discussed herein are associated with the health industry. However, any service industry can benefit from the disclosure herein, and thus the examples associated with the health industry are not meant to be limiting.

[0136] Throughout the course of a relationship between the service provider **112** and a user (e.g., the service provider **112** provides healthcare to a patient), the service provider **112** collects and generates data associated with the patient or the user, including health records that include doctor's notes about the patient and prescriptions, billing records, and insurance records. The service provider **112**, using a computing device (e.g., a desktop computer or a tablet), provides the data associated with the user to the cognitive intelligence platform **102**, and more specifically the knowledge cloud **106**.

[0137] Another example source that provides data to the knowledge cloud **106** is the facility **114**. The facility **114** represents a location owned, operated, or associated with any entity including the service provider **112**. As used herein, an entity represents an individual or a collective with a distinct and independent existence. An entity can be legally recognized (e.g., a sole proprietorship, a partnership, a corporation) or less formally recognized in a community. For example, the entity can include a company that owns or operates a gym (facility). Additional examples of the facility **114** include, but is not limited to, a hospital, a trauma center, a clinic, a dentist's office, a pharmacy, a store (including brick and mortar stores and online retailers), an out-patient care center, a specialized care center, a birthing center, a gym, a cafeteria, and a psychiatric care center.

[0138] As the facility **114** represents a large number of types of locations, for purposes of this discussion and to orient the reader by way of example, the facility **114** represents the doctor's office or a gym. The facility **114** generates additional data associated with the user such as appointment times, an attendance record (e.g., how often the user goes to the gym), a medical record, a billing record, a purchase record, an order history, and an insurance record. The facility **114**, using a computing device (e.g., a desktop computer or a tablet), provides the data associated with the user to the cognitive intelligence platform **102**, and more specifically the knowledge cloud **106**.

[0139] An additional example source that provides data to the knowledge cloud **106** is the microsurvey **116**. The microsurvey **116** represents a tool created by the cognitive intelligence platform **102** that enables the knowledge cloud **106** to collect additional data associated with the user. The microsurvey **116** is originally provided by the cognitive intelligence platform **102** (by way of the cognitive agent **110**) and the user provides data responsive to the microsurvey **116** using the user device **104**. Additional details of the microsurvey **116** are described below.

[0140] Yet another example source that provides data to the knowledge cloud **106**, is the cognitive intelligence platform **102**, itself. In order to address the care needs and well-being of the user, the cognitive intelligence platform **102** collects, analyzes, and processes information from the user, healthcare providers, and other eco-system participants, and consolidates and integrates the information into knowledge. For example, clinical-based evidence and guidelines may be obtained by the cognitive intelligence platform **102** and used as knowledge. The knowledge can be shared with the user and stored in the knowledge cloud **106**.

[0141] In various embodiments, the computing devices used by the service provider **112** and the facility **114** are communicatively coupled to the cognitive intelligence platform **102**, by way of the network **120**. While data is used individually by various entities including: a hospital, practice group, facility, or provider, the data is less frequently integrated and seamlessly shared between the various entities in the current art. The cognitive intelligence platform **102** provides a solution that integrates data from the various entities. That is, the cognitive intelligence platform **102** ingests, processes, and disseminates data and knowledge in an accessible fashion, where the reason for a particular answer or dissemination of data is accessible by a user.
[0142] In particular, the cognitive intelligence platform **102** (e.g., by way of the cognitive agent **110** interacting with the user) holistically manages and executes a health plan for durational care and wellness of the user (e.g., a patient or consumer). The health plan includes various aspects of durational management that is coordinated through a care continuum.
[0143] The cognitive agent **110** can implement various personas that are customizable. For example, the personas can include knowledgeable (sage), advocate (coach), and witty friend (jester). And in various embodiments, the cognitive agent **110** persists with a user across various interactions (e.g., conversations streams), instead of being transactional or transient. Thus, the cognitive agent **110** engages in dynamic conversations with the user, where the cognitive intelligence platform **102** continuously deciphers topics that a user wants to talk about. The cognitive intelligence platform **102** has relevant conversations with the user by ascertaining topics of interest from a given text posed in a natural language input by the user. Additionally the cognitive agent **110** connects the user to healthcare service providers, hyperlocal health communities, and a variety of services and tools/devices, based on an assessed interest of the user.
[0144] As the cognitive agent **110** persists with the user, the cognitive agent **110** can also act as a coach and advocate while delivering pieces of information to the user based on tonal knowledge, human-like empathies, and motivational dialog within a respective conversational stream, where the conversational stream is a technical discussion focused on a specific topic. Overall, in response to a question—e.g., posed by the user in natural language—the cognitive intelligence platform **102** consumes data from and related to the user and computes an answer. The answer is generated using a rationale that makes use of common sense knowledge, domain knowledge, evidence-based medicine guidelines, clinical ontologies, and curated medical advice. Thus, the content displayed by the cognitive intelligence platform **102** (by way of the cognitive agent **110**) is customized based on the language used to communicate with the user, as well as factors such as a tone, goal, and depth of topic to be discussed.
[0145] Overall, the cognitive intelligence platform **102** is accessible to a user, a hospital system, and physician. Additionally, the cognitive intelligence platform **102** is accessible to paying entities interested in user behavior—e.g., the outcome of physician-consumer interactions in the context of disease or the progress of risk management. Additionally, entities that provides specialized services such as tests, therapies, and clinical processes that need risk based interactions can also receive filtered leads from the cognitive intelligence platform **102** for potential clients.

Conversational Analysis

[0146] In various embodiments, the cognitive intelligence platform **102** is configured to perform conversational analysis in a general setting. The topics covered in the general setting is driven by the combination of agents (e.g., cognitive agent **110**) selected by a user. In some embodiments, the cognitive intelligence platform **102** uses conversational analysis to identify the intent of the user (e.g., find data, ask a question, search for facts, find references, and find products) and a respective micro-theory in which the intent is logical.
[0147] For example, the cognitive intelligence platform **102** applies conversational analysis to decode what the user is asking or stated, where the question or statement is in free form language (e.g., natural language). Prior to determining and sharing knowledge (e.g., with the user or the knowledge cloud **106**), using conversational analysis, the cognitive intelligence platform **102**

identifies an intent of the user and overall conversational focus.
[0148] The cognitive intelligence platform **102** responds to a statement or question according to the conversational focus and steers away from another detected conversational focus so as to focus on a goal defined by the cognitive agent **110**. Given an example statement of a user, "I want to fly out tomorrow," the cognitive intelligence platform **102** uses conversational analysis to determine an intent of the statement. Is the user aspiring to be bird-like or does he want to travel? In the former case, the micro-theory is that of human emotions whereas in the latter case, the micro-theory is the world of travel. Answers are provided to the statement depending on the micro-theory in which the intent logically falls.
[0149] The cognitive intelligence platform **102** utilize a combination of linguistics, artificial intelligence, and decision trees to decode what a user is asking or stating. The discussion includes methods and system design considerations and results from an existing embodiment. Additional details related to conversational analysis are discussed next.
Analyzing Conversational Context As Part of Conversational Analysis
[0150] For purposes of this discussion, the concept of analyzing conversational context as part of conversational analysis is now described. To analyze conversational context, the following steps are taken: 1) obtain text (e.g., receive a question) and perform translations; 2) understand concepts, entities, intents, and micro-theory; 3) relate and search; 4) ascertain the existence of related concepts; 5) logically frame concepts or needs; 6) understand the questions that can be answered from available data; and 7) answer the question. Each of the foregoing steps is discussed next, in turn.
Step 1: Obtain Text/Question and Perform Translations
[0151] In various embodiments, the cognitive intelligence platform **102** (FIG. **1**) receives a text or question and performs translations as appropriate. The cognitive intelligence platform **102** supports various methods of input including text received from a touch interface (e.g., options presented in a microsurvey), text input through a microphone (e.g., words spoken into the user device), and text typed on a keyboard or on a graphical user interface. Additionally, the cognitive intelligence platform **102** supports multiple languages and auto translation (e.g., from English to Traditional/Simplified Chinese or vice versa).
[0152] The example text below is used to described methods in accordance with various embodiments herein: [0153] "One day in January 1913. G. H. Hardy, a famous Cambridge University mathematician received a letter from an Indian named Srinivasa Ramanujan asking him for his opinion of **120** mathematical theorems that Ramanujan said he had discovered. To Hardy, many of the theorems made no sense. Of the others, one or two were already well-known. Ramanujan must be some kind of trickplayer, Hardy decided, and put the letter aside. But all that day the letter kept hanging round Hardy. Might there by something in those wild-looking theorems? [0154] That evening Hardy invited another brilliant Cambridge mathematician, J. E. Littlewood, and the two men set out to assess the Indian's worth. That incident was a turning point in the history of mathematics. [0155] At the time, Ramanujan was an obscure Madras Port Trust clerk. A little more than a year later, he was at Cambridge University, and beginning to be recognized as one of the most amazing mathematicians the world has ever known. Though he died in **1920**, much of his work was so far in advance of his time that only in recent years is it beginning to be properly understood. [0156] Indeed, his results are helping solve today's problems in computer science and physics, problems that he could have had no notion of. [0157] For Indians, moreover, Ramanujan has a special significance. Ramanujan, through born in poor and ill-paid accountant's family 100 years ago, has inspired many Indians to adopt mathematics as career. [0158] Much of Ramanujan's work is in number theory, a branch of mathematics that deals with the subtle laws and relationships that govern numbers. Mathematicians describe his results as elegant and beautiful but they are much too complex to be appreciated by laymen. [0159] His life, though, is full of drama and sorrow. It is one of the great romantic stories of mathematics, a distressing

reminder that genius can surface and rise in the most unpromising circumstances."
[0160] The cognitive intelligence platform **102** analyzes the example text above to detect structural elements within the example text (e.g., paragraphs, sentences, and phrases). In some embodiments, the example text is compared to other sources of text such as dictionaries, and other general fact databases (e.g., Wikipedia) to detect synonyms and common phrases present within the example text.
Step 2: Understand Concept, Entity, Intent, and Micro-Theory
[0161] In step 2, the cognitive intelligence platform **102** parses the text to ascertain concepts, entities, intents, and micro-theories. An example output after the cognitive intelligence platform **102** initially parses the text is shown below, where concepts, and entities are shown in bold. [0162] "One day in January 1913. G. H. Hardy, a famous Cambridge University mathematician received a letter from an Indian named Srinivasa Ramanujan asking him for his opinion of **120** mathematical theorems that Ramanujan said he had discovered. To Hardy, many of the theorems made no sense. Of the others, one or two were already well-known. Ramanujan must be some kind of trickplayer, Hardy decided, and put the letter aside. But all that day the letter kept hanging round Hardy. Might there by something in those wild-looking theorems?
[0163] That evening Hardy invited another brilliant Cambridge mathematician, J. E. Littlewood, and the two men set out to assess the Indian's worth. That incident was a turning point in the history of mathematics. [0164] At the time, Ramanujan was an obscure Madras Port Trust clerk. A little more than a year later, he was at Cambridge University, and beginning to be recognized as one of the most amazing mathematicians the world has ever known. Though he died in **1920**, much of his work was so far in advance of his time that only in recent years is it beginning to be properly understood. [0165] Indeed, his results are helping solve today's problems in computer science and physics, problems that he could have had no notion of. [0166] For Indians, moreover, Ramanujan has a special significance. Ramanujan, through born in poor and ill-paid accountant's family 100 years ago, has inspired many Indians to adopt mathematics as career. [0167] Much of Ramanujan's work is in number theory, a branch of mathematics that deals with the subtle laws and relationships that govern numbers. [0168] Mathematicians describe his results as elegant and beautiful but they are much too complex to be appreciated by laymen. [0169] His life, though, is full of drama and sorrow. It is one of the great romantic stories of mathematics, a distressing reminder that genius can surface and rise in the most unpromising circumstances."
[0170] For example, the cognitive intelligence platform **102** ascertains that Cambridge is a university-which is a full understanding of the concept. The cognitive intelligence platform (e.g., the cognitive agent **110**) understands what humans do in Cambridge, and an example is described below in which the cognitive intelligence platform **102** performs steps to understand a concept.
[0171] For example, in the context of the above example, the cognitive agent **110** understands the following concepts and relationships: [0172] Cambridge employed John Edensor Littlewood (1) [0173] Cambridge has the position Ramanujan's position at Cambridge University (2) [0174] Cambridge employed G. H. Hardy. (3)
[0175] The cognitive agent **110** also assimilates other understandings to enhance the concepts, such as: [0176] Cambridge has Trinity College as a suborganization. (4) [0177] Cambride is located in Cambridge. (5) [0178] Alan Turing is previously enrolled at Cambridge. (6) [0179] Stephen Hawking attended Cambridge. (7)
[0180] The statements (1)-(7) are not picked at random. Instead the cognitive agent **110** dynamically constructs the statements (1)-(7) from logic or logical inferences based on the example text above. Formally, the example statements (1)-(7) are captured as follows: [0181] (#$subOrganizations #$UniversityOfCambridge #$TrinityCollege-Cambridge-England) (8) [0182] (#$placeInCity #$UniversityOfCambridge #$Cityof CambridgeEngland) (9) [0183] (#$schooling #$AlanTuring #$UniversityOfCambridge #$PreviouslyEnrolled) (10) [0184] (#$hasAlumni #$UniversityOfCambridge #$StephenHawking) (11)

Step 3: Relate and Search

[0185] Next, in step 3, the cognitive agent **110** relates various entities and topics and follows the progression of topics in the example text. Relating includes the cognitive agent **110** understanding the different instances of Hardy are all the same person, and the instances of Hardy are different from the instances of Littlewood. The cognitive agent **110** also understands that the instances Hardy and Littlewood share some similarities—e.g., both are mathematicians and they did some work together at Cambridge on Number Theory. The ability to track this across the example text is referred to as following the topic progression with a context.

Step 4: Ascertain the Existence of Related Concepts

[0186] Next, in Step 4, the cognitive agent **110** asserts non-existent concepts or relations to form new knowledge. Step 4 is an optional step for analyzing conversational context. Step 4 enhances the degree to which relationships are understood or different parts of the example text are understood together. If two concepts appear to be separate—e.g., a relationship cannot be graphically drawn or logically expressed between enough sets of concepts—there is a barrier to understanding. The barriers are overcome by expressing additional relationships. The additional relationships can be discovered using strategies like adding common sense or general knowledge sources (e.g., using the common sense data **208**) or adding in other sources including a lexical variant database, a dictionary, and a thesaurus.

[0187] One example of concept progression from the example text is as follows: the cognitive agent **110** ascertains the phrase “theorems that Ramanujan said he had discovered” is related to the phrase “his results”, which is related to “Ramanujan's work is in number theory, a branch of mathematics that deals with the subtle laws and relationships that govern numbers.”

Step 5: Logically Frame Concepts or Needs

[0188] In Step 5, the cognitive agent **110** determines missing parameters-which can include for example, missing entities, missing elements, and missing nodes-in the logical framework (e.g., with a respective micro-theory). The cognitive agent **110** determines sources of data that can inform the missing parameters. Step 5 can also include the cognitive agent **110** adding common sense reasoning and finding logical paths to solutions.

[0189] With regards to the example text, some common sense concepts include: [0190] Mathematicians develop Theorems. (12) [0191] Theorems are hard to comprehend. (13) [0192] Interpretations are not apparent for years. (14) [0193] Applications are developed over time. (15) [0194] Mathematicians collaborate and assess work. (16)

[0195] With regards to the example text, some passage concepts include: [0196] Ramanujan did Theorems in Early 20th Century. (17) [0197] Hardy assessed Ramanujan's Theorems. (18) [0198] Hardy collaborated with Littlewood. (19) [0199] Hardy and Littlewood assessed Ramanujan's work (20)

Within the micro-theory of the passage analysis, the cognitive agent **110** understands and catalogs available paths to answer questions. In Step 5, the cognitive agent **110** makes the case that the concepts (**12**)-(**20**) are expressed together.

Step 6: Understand the Questions that can be Answered from Available Data

[0200] In Step 6, the cognitive agent **110** parses sub-intents and entities. Given the example text, the following questions are answerable from the cognitive agent's developed understanding of the example text, where the understanding was developed using information and context ascertained from the example text as well as the common sense data **208** (FIG. **2**): [0201] What situation causally contributed to Ramanujan's position at Cambridge? (21) [0202] Does the author of the passage regret that Ramanujan died prematurely? (22) [0203] Does the author of the passage believe that Ramanujan is a mathematical genius? (23) Based on the information that is understood by the cognitive agent **110**, the questions (21)-(23) can be answered.

[0204] By using an exploration method such as random walks, the cognitive agent **110** makes a determination as the paths that are plausible and reachable with the context (e.g., micro-theory) of

the example text. Upon explorations, the cognitive agent **110** catalogs a set of meaningful questions. The set of meaningful questions are not asked, but instead explored based on the cognitive agent's understanding of the example text.
[0205] Given the example text, an example of exploration that yields a positive result is: “a situation X that caused Ramanujan's position.” In contrast, an example of exploration that causes irrelevant results is: “a situation Y that caused Cambridge.” The cognitive agent **110** is able to deduce that the latter exploration is meaningless, in the context of a micro-theory, because situations do not cause universities. Thus the cognitive agent **110** is able to deduce, there are no answers to Y, but there are answers to X.
Step 7: Answer the Question
[0206] In Step 7, the cognitive agent **110** provides a precise answer to a question. For an example question such as: “What situation causally contributed to Ramanujan's position at Cambridge?” the cognitive agent **110** generates a precise answer using the example reasoning: [0207] HardyandLittlewoodsEvaluatingOfRamanujansWork (24) [0208] HardyBeliefThatRamanujanlsAnExpertInMathematics (25) [0209] HardysBeliefThatRamanujanlsAnExpertInMathematicsAndAGenius (26)
[0210] In order to generate the above reasoning statements (24)-(26), the cognitive agent **110** utilizes a solver or prover in the context of the example text's micro-theory—and associated facts, logical entities, relations, and assertions. As an additional example, the cognitive agent **110** uses a reasoning library that is optimized for drawing the example conclusions above within the fact, knowledge, and inference space (e.g., work space) that the cognitive agent **110** maintains.
[0211] By implementing the steps 1-7, the cognitive agent **110** analyzes conversational context. The described method for analyzing conversation context can also be used for recommending items in conversations streams. A conversational stream is defined herein as a technical discussion focused on specific topics. As related to described examples herein, the specific topics relate to health (e.g., diabetes). Throughout the lifetime of a conversational stream, a cognitive agent **110** collect information over may channels such as chat, voice, specialized applications, web browsers, contact centers, and the like.
[0212] By implementing the methods to analyze conversational context, the cognitive agent **110** can recommend a variety of topics and items throughout the lifetime of the conversational stream. Examples of items that can be recommended by the cognitive agent **110** include: surveys, topics of interest, local events, devices or gadgets, dynamically adapted health assessments, nutritional tips, reminders from a health events calendar, and the like.
[0213] Accordingly, the cognitive intelligence platform **102** provides a platform that codifies and takes into consideration a set of allowed actions and a set of desired outcomes. The cognitive intelligence platform **102** relates actions, the sequences of subsequent actions (and reactions), desired sub-outcomes, and outcomes, in a way that is transparent and logical (e.g., explainable). The cognitive intelligence platform **102** can plot a next best action sequence and a planning basis (e.g., health care plan template, or a financial goal achievement template), also in a manner that is explainable. The cognitive intelligence platform **102** can utilize a critical thinking engine **108** and a natural language database **122** (e.g., a linguistics and natural language understanding system) to relate conversation material to actions.
[0214] For purposes of this discussion, several examples are discussed in which conversational analysis is applied within the field of durational and whole-health management for a user. The discussed embodiments holistically address the care needs and well-being of the user during the course of his life. The methods and systems described herein can also be used in fields outside of whole-health management, including: phone companies that benefits from a cognitive agent; hospital systems or physicians groups that want to coach and educate patients; entities interested in user behavior and the outcome of physician-consumer interactions in terms of a progress of disease or risk management; entities that provide specialized services (e.g., test, therapies, clinical

processes) to filter leads; and sellers, merchants, stores and big box retailers that want to understand which product to sell.

[0215] In addition, the conversational analysis may include cognifying the text input by the user. For example, if the user states (e.g., text, voice) they have various symptoms, the cornification techniques disclosed herein may be performed to construct cognified data using the text input. The user may input text specifying that they have a level of 5.7 mmol/L blood sugar. The cognitive intelligence platform **102** may cognify the text to output that the level of blood sugar is within acceptable limits, and that blood sugar testing was used to measure the blood sugar level. In some embodiments, the cognification techniques may be performed to generate a diagnosis of a medical condition of the patient. Further, the cognitive intelligence platform **102** may provide information to the user pertaining to the medical condition at a regulated pace.

[0216] FIG. **2** shows additional details of a knowledge cloud, in accordance with various embodiments. In particular, FIG. **2** illustrates various types of data received from various sources, including service provider data **202**, facility data **204**, microsurvey data **206**, commonsense data **208**, domain data **210**, evidence-based guidelines **212**, subject matter ontology data **214**, and curated advice **216**. The types of data represented by the service provider data **202** and the facility data **204** include any type of data generated by the service provider **112** and the facility **114**, and the above examples are not meant to be limiting. Thus, the example types of data are not meant to be limiting and other types of data can also be stored within the knowledge cloud **106** without departing from the scope of this disclosure.

[0217] The service provider data **202** is data provided by the service provider **112** (described in FIG. **1**) and the facility data **204** is data provided by the facility **114** (described in FIG. **1**). For example, the service provider data **202** includes medical records of a respective patient of a service provider **112** that is a doctor. In another example, the facility data **204** includes an attendance record of the respective patient, where the facility **114** is a gym. The microsurvey data **206** is data provided by the user device **104** responsive to questions presented in the microsurvey **116** (FIG. **1**).

[0218] Common sense data **208** is data that has been identified as “common sense”, and can include rules that govern a respective concept and used as glue to understand other concepts.

[0219] Domain data **210** is data that is specific to a certain domain or subject area. The source of the domain data **210** can include digital libraries. In the healthcare industry, for example, the domain data **210** can include data specific to the various specialties within healthcare such as, obstetrics, anesthesiology, and dermatology, to name a few examples. In the example described herein, the evidence-based guidelines **212** include systematically developed statements to assist practitioner and patient decisions about appropriate health care for specific clinical circumstances.

[0220] Curated advice **214** includes advice from experts in a subject matter. The curated advice **214** can include peer-reviewed subject matter, and expert opinions. Subject matter ontology data **216** includes a set of concepts and categories in a subject matter or domain, where the set of concepts and categories capture properties and relationships between the concepts and categories.

[0221] In particular, FIG. **3** illustrates an example subject matter ontology **300** that is included as part of the subject matter ontology data **216**.

[0222] FIG. **4** illustrates aspects of a conversation **400** between a user and the cognitive intelligence platform **102**, and more specifically the cognitive agent **110**. For purposes of this discussion, the user **401** is a patient of the service provider **112**. The user interacts with the cognitive agent **110** using a computing device, a smart phone, or any other device configured to communicate with the cognitive agent **110** (e.g., the user device **104** in FIG. **1**). The user can enter text into the device using any known means of input including a keyboard, a touchscreen, and a microphone. The conversation **400** represents an example graphical user interface (GUI) presented to the user **401** on a screen of his computing device.

[0223] Initially, the user asks a general question, which is treated by the cognitive agent **110** as an “originating question.” The originating question is classified into any number of potential questions

("pursuable questions") that are pursued during the course of a subsequent conversation. In some embodiments, the pursuable questions are identified based on a subject matter domain or goal. In some embodiments, classification techniques are used to analyze language (e.g., such as those outlined in HPS ID20180901-01_method for conversational analysis). Any known text classification technique can be used to analyze language and the originating question. For example, in line **402**, the user enters an originating question about a subject matter (e.g., blood sugar) such as: "Is a blood sugar of 90 normal"? |

[0224] In response to receiving an originating question, the cognitive intelligence platform **102** (e.g., the cognitive agent **110** operating in conjunction with the critical thinking engine **108**) performs a first round of analysis (e.g., which includes conversational analysis) of the originating question and, in response to the first round of analysis, creates a workspace and determines a first set of follow up questions.

[0225] In various embodiments, the cognitive agent **110** may go through several rounds of analysis executing within the workspace, where a round of analysis includes: identifying parameters, retrieving answers, and consolidating the answers. The created workspace can represent a space where the cognitive agent **110** gathers data and information during the processes of answering the originating question. In various embodiments, each originating question corresponds to a respective workspace. The conversation orchestrator **124** can assess data present within the workspace and query the cognitive agent **110** to determine if additional data or analysis should be performed.

[0226] In particular, the first round of analysis is performed at different levels, including analyzing natural language of the text, and analyzing what specifically is being asked about the subject matter (e.g., analyzing conversational context). The first round of analysis is not based solely on a subject matter category within which the originating question is classified. For example, the cognitive intelligence platform **102** does not simply retrieve a predefined list of questions in response to a question that falls within a particular subject matter, e.g., blood sugar. That is, the cognitive intelligence platform **102** does not provide the same list of questions for all questions related to the particular subject matter. Instead, for example, the cognitive intelligence platform **102** creates dynamically formulated questions, curated based on the first round of analysis of the originating question.

[0227] In particular, during the first round of analysis, the cognitive agent **110** parses aspects of the originating question into associated parameters. The parameters represent variables useful for answering the originating question. For example, the question "is a blood sugar of 90 normal" may be parsed and associated parameters may include, an age of the inquirer, the source of the value **90** (e.g., in home test or a clinical test), a weight of the inquirer, and a digestive state of the user when the test was taken (e.g., fasting or recently eaten). The parameters identify possible variables that can impact, inform, or direct an answer to the originating question.

[0228] For purposes of the example illustrated in FIG. **4**, in the first round of analysis, the cognitive intelligence platform **102** inserts each parameter into the workspace associated with the originating question (line **402**). Additionally, based on the identified parameters, the cognitive intelligence platform **102** identifies a customized set of follow up questions ("a first set of follow-up questions). The cognitive intelligence platform **102** inserts first set of follow-up questions in the workspace associated with the originating question.

[0229] The follow up questions are based on the identified parameters, which in turn are based on the specifics of the originating question (e.g., related to an identified micro-theory). Thus the first set of follow-up questions identified in response to, if a blood sugar is normal, will be different from a second set of follow up questions identified in response to a question about how to maintain a steady blood sugar.

[0230] After identifying the first set of follow up questions, in this example first round of analysis, the cognitive intelligence platform **102** determines which follow up question can be answered using available data and which follow-up question to present to the user. As described over the next few

paragraphs, eventually, the first set of follow-up questions is reduced to a subset (“a second set of follow-up questions”) that includes the follow-up questions to present to the user.
[0231] In various embodiments, available data is sourced from various locations, including a user account, the knowledge cloud **106**, and other sources. Other sources can include a service that supplies identifying information of the user, where the information can include demographics or other characteristics of the user (e.g., a medical condition, a lifestyle). For example, the service can include a doctor's office or a physical therapist's office.
[0232] Another example of available data includes the user account. For example, the cognitive intelligence platform **102** determines if the user asking the originating question, is identified. A user can be identified if the user is logged into an account associated with the cognitive intelligence platform **102**. User information from the account is a source of available data. The available data is inserted into the workspace of the cognitive agent **110** as a first data.
[0233] Another example of available data includes the data stored within the knowledge cloud **106**. For example, the available data includes the service provider data **202** (FIG. **2**), the facility data **204**, the microsurvey data **206**, the common sense data **208**, the domain data **210**, the evidence-based guidelines **212**, the curated advice **214**, and the subject matter ontology data **216**. Additionally data stored within the knowledge cloud **106** includes data generated by the cognitive intelligence platform **102**, itself.
[0234] Follow up questions presented to the user (the second set of follow-up questions) are asked using natural language and are specifically formulated (“dynamically formulated question”) to elicit a response that will inform or fulfill an identified parameter. Each dynamically formulated question can target one parameter at a time. When answers are received from the user in response to a dynamically formulated question, the cognitive intelligence platform **102** inserts the answer into the workspace. In some embodiments, each of the answers received from the user and in response to a dynamically formulated question, is stored in a list of facts. Thus the list of facts include information specifically received from the user, and the list of facts is referred to herein as the second data.
[0235] With regards to the second set of follow-up questions (or any set of follow-up questions), the cognitive intelligence platform **102** calculates a relevance index, where the relevance index provides a ranking of the questions in the second set of follow-up questions. The ranking provides values indicative of how relevant a respective follow-up question is to the originating question. To calculate the relevance index, the cognitive intelligence platform **102** can use conversations analysis techniques described in HPS ID20180901-01_method. In some embodiments, the first set or second set of follow up questions is presented to the user in the form of the microsurvey **116**.
[0236] In this first round of analysis, the cognitive intelligence platform **102** consolidates the first and second data in the workspace and determines if additional parameters need to be identified, or if sufficient information is present in the workspace to answer the originating question. In some embodiments, the cognitive agent **110** (FIG. **1**) assesses the data in the workspace and queries the cognitive agent **110** to determine if the cognitive agent **110** needs more data in order to answer the originating question. The conversation orchestrator **124** executes as an interface
[0237] For a complex originating question, the cognitive intelligence platform **102** can go through several rounds of analysis. For example, in a first round of analysis the cognitive intelligence platform **102** parses the originating question. In a subsequent round of analysis, the cognitive intelligence platform **102** can create a sub question, which is subsequently parsed into parameters in the subsequent round of analysis. The cognitive intelligence platform **102** is smart enough to figure out when all information is present to answer an originating question without explicitly programming or pre-programming the sequence of parameters that need to be asked about.
[0238] In some embodiments, the cognitive agent **110** is configured to process two or more conflicting pieces of information or streams of logic. That is, the cognitive agent **110**, for a given originating question can create a first chain of logic and a second chain of logic that leads to

different answers. The cognitive agent **110** has the capability to assess each chain of logic and provide only one answer. That is, the cognitive agent **110** has the ability to process conflicting information received during a round of analysis.

[0239] Additionally, at any given time, the cognitive agent **110** has the ability to share its reasoning (chain of logic) to the user. If the user does not agree with an aspect of the reasoning, the user can provide that feedback which results in affecting change in a way the critical thinking engine **108** analyzed future questions and problems.

[0240] Subsequent to determining enough information is present in the workspace to answer the originating question, the cognitive agent **110** answers the question, and additionally can suggest a recommendation or a recommendation (e.g., line **418**). The cognitive agent **110** suggests the reference or the recommendation based on the context and questions being discussed in the conversation (e.g., conversation **400**). The reference or recommendation serves as additional handout material to the user and is provided for informational purposes. The reference or recommendation often educates the user about the overall topic related to the originating question.

[0241] In the example illustrated in FIG. **4**, in response to receiving the originating questions (line **402**), the cognitive intelligence platform **102** (e.g., the cognitive agent **110** in conjunction with the critical thinking engine **108**) parses the originating question to determine at least one parameter: location. The cognitive intelligence platform **102** categorizes this parameter, and a corresponding dynamically formulated question in the second set of follow-up questions. Accordingly, in lines **404** and **406**, the cognitive agent **110** responds by notifying the user "I can certainly check this . . . " and asking the dynamically formulated question "I need some additional information in order to answer this question, was this an in-home glucose test or was it done by a lab or testing service?"

[0242] The user **401** enters his answer in line **408**: "It was an in-home test," which the cognitive agent **110** further analyzes to determine additional parameters: e.g., a digestive state, where the additional parameter and a corresponding dynamically formulated question as an additional second set of follow-up questions. Accordingly, the cognitive agent **110** poses the additional dynamically formulated question in lines **410** and **412**: "One other question . . . " and "How long before you took that in-home glucose test did you have a meal?" The user provides additional information in response "it was about an hour" (line **414**).

[0243] The cognitive agent **110** consolidates all the received responses using the critical thinking engine **108** and the knowledge cloud **106** and determines an answer to the initial question posed in line **402** and proceeds to follow up with a final question to verify the user's initial question was answered. For example, in line **416**, the cognitive agent **110** responds: "It looks like the results of your test are at the upper end of the normal range of values for a glucose test given that you had a meal around an hour before the test." The cognitive agent **110** provides additional information (e.g., provided as a link): "Here is something you could refer," (line **418**), and follows up with a question "Did that answer your question?" (line **420**).

[0244] As described above, due to the natural language database **108**, in various embodiments, the cognitive agent **110** is able to analyze and respond to questions and statements made by a user **401** in natural language. That is, the user **401** is not restricted to using certain phrases in order for the cognitive agent **110** to understand what a user **401** is saying. Any phrasing, similar to how the user would speak naturally can be input by the user and the cognitive agent **110** has the ability to understand the user.

[0245] FIG. **5** illustrates a cognitive map or "knowledge graph" **500**, in accordance with various embodiments. In particular, the knowledge graph represents a graph traversed by the cognitive intelligence platform **102**, when assessing questions from a user with Type 2 diabetes. Individual nodes in the knowledge graph **500** represent a health artifact (health related information) or relationship (predicate) that is gleaned from direct interrogation or indirect interactions with the user (by way of the user device **104**).

[0246] In one embodiment, the cognitive intelligence platform **102** identified parameters for an

originating question based on a knowledge graph illustrated in FIG. **5**. For example, the cognitive intelligence platform **102** parses the originating question to determine which parameters are present for the originating question. In some embodiments, the cognitive intelligence platform **102** infers the logical structure of the parameters by traversing the knowledge graph **500**, and additionally, knowing the logical structure enables the cognitive agent **110** to formulate an explanation as to why the cognitive agent **110** is asking a particular dynamically formulated question.

[0247] In some embodiments, the individual elements or nodes are generated by the artificial intelligence engine based on input data (e.g., evidence-based guidelines, patient notes, clinical trials, physician research or the like). The artificial intelligence engine may parse the input data and construct the relationships between the health artifacts.

[0248] For example, a root node may be associated with a first health related information "Type 2 Diabetes Mellitus", which is a name of a medical condition. In some embodiments, the root node may also be associated with a definition of the medical condition. An example predicate, "has symptom", is represented by an individual node connected to the root node, and another health related information, "High Blood Sugar", is represented by an individual node connected to the individual node representing the predicate. A logical structure may be represented by these three nodes, and the logical structure may indicate that "Type 2 Diabetes Mellitus has symptom High Blood Sugar".

[0249] In some embodiments, the health related information may correspond to known facts, concepts, and/or any suitable health related information that are discovered or provided by a trusted source (e.g., a physician having a medical license and/or a certified/accredited healthcare organization), such as evidence-based guidelines, clinical trials, physician research, patient notes entered by physicians, and the like. The predicates may be part of a logical structure (e.g., sentence) such as a form of subject-predicate-direct object, subject-predicate-indirect object-direct object, subject-predicate-subject complement, or any suitable simple, compound, complex, and/or compound/complex logical structure. The subject may be a person, place, thing, health artifact, etc. The predicate may express an action or being within the logical structure and may be a verb, modifying words, phrases, and/or clauses. For example, one logical structure may be the subject-predicate-direct object form, such as "A has B" (where A is the subject and may be a noun or a health artifact, "has" is the predicate, and B is the direct object and may be a health artifact).

[0250] The various logical structures in the depicted knowledge graph may include the following: "Type 2 Diabetes Mellitus has symptom High Blood Sugar"; "Type 2 Diabetes Mellitus has complication Stroke"; "Type 2 Diabetes Mellitus has complication Coronary Artery Disease"; "Type 2 Diabetes Mellitus has complication Diabetes Foot Problems"; "Type 2 Diabetes Mellitus has complication Diabetic Neuropathy"; "Type 2 Diabetes Mellitus has complication Diabetic Retinopathy"; "Type 2 Diabetes Mellitus diagnosed or monitored using Blood Glucose Test"; just to name a few examples. It should be understood that there are other logical structures and represented in the knowledge graph **500**.

[0251] In some embodiments, the information depicted in the knowledge graph may be represented as a matrix. The health artifacts may be represented as quantities and the predicates may be represented as expressions in a rectangular array in rows and columns of the matrix. The matrix may be treated as a single entity and manipulated according to particular rules.

[0252] The knowledge graph **500** or the matrix may be generated for each known medical condition and stored by the cognitive intelligence platform **102**. The knowledge graphs and/or matrices may be updated continuously or on a periodic basis using subject data pertaining to the medical conditions received from the trusted sources. For example, additional clinical trials may lead to new discoveries about particular medical condition treatments, which may be used to update the knowledge graphs and/or matrices.

[0253] The knowledge graph **500** including the logical structures may be used to transform unstructured data (patient notes in an EMR entered by a physician) into cognified data. The

cognified data may be used to generate a diagnosis of the patient. Also, the cognified data may be used to determine which information pertaining to the medical condition to provide to the patient and when to provide the information to the patient to improve the user experience using the computing device. The disclosed techniques may also save computing resources by providing the cognified data to the physician to review, improve diagnosis accuracy, and/or regulate the amount of information provided to the patient.

[0254] FIG. **6** shows a method, in accordance with various embodiments. The method is performed at a user device (e.g., the user device **102**) and in particular, the method is performed by an application executing on the user device **102**. The method begins with initiating a user registration process (block **602**). The user registration can include tasks such as displaying a GUI asking the user to enter in personal information such as his name and contact information.

[0255] Next, the method includes prompting the user to build his profile (block **604**). In various embodiments, building his profile includes displaying a GUI asking the user to enter in additional information, such as age, weight, height, and health concerns. In various embodiments, the steps of building a user profile is progressive, where building the user profile takes place over time. In some embodiments, the process of building the user profile is presented as a game. Where a user is presented with a ladder approach to create a “star profile”. Aspects of a graphical user interface presented during the profile building step are additionally discussed in FIGS. **8**A-**8**B.

[0256] The method contemplates the build profile (block **604**) method step is optional. For example, the user may complete building his profile at this method step 604, the user may complete his profile at a later time, or the cognitive intelligence platform **102** builds the user profile over time as more data about the user is received and processed. For example, the user is prompted to build his profile, however, the user fails to enter in information or skips the step. The method proceeds to prompting a user to complete a microsurvey (block **606**). In some embodiments, the cognitive agent **110** uses answers received in response to the microsurvey to build the profile of the user. Overall, the data collected through the user registration process is stored and used later as available data to inform answers to missing parameters.

[0257] Next, the cognitive agent **110** proceeds to scheduling a service (block **608**). The service can be scheduled such that it aligns with a health plan of the user or a protocol that results in a therapeutic goal. Next, the cognitive agent **110** proceeds to reaching agreement on a care plan (block **610**).

[0258] FIGS. **7**A, **7**B, and **7**C, show methods, in accordance with various embodiments. The methods are performed at the cognitive intelligence platform. In particular, in FIG. **7**A, the method begins with receiving a first data including user registration data (block **702**); and providing a health assessment and receiving second data including health assessment answers (block **704**). In various embodiments, the health assessment is a micro-survey with dynamically formulated questions presented to the user.

[0259] Next the method determine if the user provided data to build a profile (decision block **706**). If the user did not provide data to build the profile, the method proceeds to building profile based on first and second data (block **708**). If the user provided data to build the profile, the method proceeds to block **710**.

[0260] At block **710**, the method **700** proceeds to receiving an originating question about a specific subject matter, where the originating question is entered using natural language, and next the method proceeds to performing a round of analysis (block **712**). Next, the method determines if sufficient data is present to answer originating questions (decision block **714**). If no, the method proceeds to block **712** and the method performs another round of analysis. If yes, the method proceeds to setting goals (block **716**), then tracking progress (block **718**), and then providing updates in a news feed (block **720**).

[0261] In FIG. **7**B, a method **730** of performing a round of analysis is illustrated. The method begins with parsing the originating question into parameters (block **732**); fulfilling the parameters

from available data (block **734**); inserting available data (first data) into a working space (block **736**); creating a dynamically formulated question to fulfill a parameter (block **738**); and inserting an answer to the dynamically formulated question into the working space (block **740**).
[0262] In FIG. **7**C, a method **750** is performed at the cognitive intelligence platform. The method begins with receiving a health plan (block **752**); accessing the knowledge cloud and retrieving first data relevant to the subject matter (block **754**); and engaging in conversation with the user using natural language to general second data (block **756**). In various embodiments, the second data can include information such as a user's scheduling preferences, lifestyle choices, and education level. During the process of engaging in conversation, the method includes educating and informing the user (block **758**). Next, the method includes defining an action plan based, at least in part, on the first and second data (block **760**); setting goals (block **762**); and tracking progress (block **764**).
[0263] FIGS. **8**A, **8**B, **8**C, and **8**D illustrate aspects of interactions between a user and the cognitive intelligence platform **102**, in accordance with various embodiments. As a user interacts with the GUI, the cognitive intelligence platform **102** continues to build a database of knowledge about the user based on questions asked by the user as well as answers provided by the user (e.g., available data as described in FIG. **4**). In particular, FIG. **8**A displays a particular screen shot **801** of the user device **104** at a particular instance in time. The screen shot **801** displays a graphical user interface (GUI) with menu items associated with a user's (e.g., Nathan) profile including Messages from the doctor (element **804**), Goals (element **806**), Trackers (element **808**), Health Record (element **810**), and Health Plans & Assessments (element **812**). The menu item Health Plans & Assessments (element **812**), additionally include child menu items: Health Assessments (element **812***a*), Health plans (**812***b*).
[0264] The screen shot **803** displays the same GUI as in the screen shot **801**, however, the user has scrolled down the menu, such that additional menu items below Health Plans & Assessments (element **812**) are shown. The additional menu items include Reports (element **814**), Health Team (element **816**), and Purchases and Services (Element **818**). Furthermore, additional menu items include Add your Health Team (element **820**) and Read about improving your A1C levels (element **822**).
[0265] For purposes of the example in FIG. **8**A, the user selects the menu item Health Plans (element **812***b*). Accordingly, in response to the receiving the selection of the menu item Health Plans, types of health plans are shown, as illustrated in screen shot **805**. The types of health plans shown with respect to Nathan's profile include: Diabetes (element **824**), Cardiovascular, Asthma, and Back Pain. Each type of health plan leads to separate displays. For purposes of this example in FIG. **8**A, the user selects the Diabetes (element **824**) health plan.
[0266] In FIG. **8**B, the screenshot **851** is seen in response to the user's selection of Diabetes (element **824**). Example elements displayed in screenshot **851** include: Know How YOUR Body Works (element **852**); Know the Current Standards of Care (element **864**); Expertise: Self-Assessment (element **866**); Expertise: Self-Care/Treatment (element **868**); and Managing with Lifestyle (element **870**). Managing with Lifestyle (element **870**) focuses and tracks actions and lifestyle actions that a user can engage in. As a user's daily routine helps to manage diabetes, managing the user's lifestyle is important. The cognitive agent **110** can align a user's respective health plan based on a health assessment at enrollment. In various embodiments, the cognitive agent **110** aligns the respective health plan with an interest of the user, a goal and priority of the user, and lifestyle factors of the user-including exercise, diet and nutrition, and stress reduction.
[0267] Each of these elements **852**, **864**, **866**, **868**, and **870** can display additional sub-elements depending on a selection of the user. For example, as shown in the screen shot **851**, Know How YOUR Body Works (element **852**) includes additional sub-elements: Diabetes Personal Assessment (**854**); and Functional Changes (**856**). Additional sub-elements under Functional Changes (**856**) include: Blood Sugar Processing (**858**) and Manageable Risks (**860**). Finally, the sub-element Manageable Risks (**860**) includes an additional sub-element Complications (**862**). For purposes of

this example, the user selects the Diabetes Personal Assessment (**854**) and the screen shot **853** shows a GUI (**872**) associated with the Diabetes Personal Assessment.
[0268] The Diabetes Personal Assessment includes questions such as “Approximately what year was your Diabetes diagnosed” and corresponding elements a user can select to answer including “Year” and “Can't remember” (element **874**). Additional questions include “Is your Diabetes Type 1 or Type 2” and corresponding answers selectable by a user include “Type 1,” “Type 2,” and “Not sure” (element **876**). Another question includes “Do you take medication to manage your blood sugar” and corresponding answers selectable by a user include “Yes” and “No” (element **878**). An additional question asks “Do you have a healthcare professional that works with you to manage your Diabetes” and corresponding answers selectable by the user include “Yes” and “No” (element **880**).
[0269] In various embodiments, the cognitive intelligence platform **102** collects information about the user based on responses provided by the user or questions asked by the user as the user interacts with the GUI. For example, as the user views the screen shot **851**, if the user asks if diabetes is curable, this question provides information about the user such as a level of education of the user.
[0270] FIG. **8**C illustrates aspects of an additional tool—e.g., a microsurvey-provided to the user that helps gather additional information about the user (e.g., available data). In various embodiments, a micro-survey represent a short targeted survey, where the questions presented in the survey are limited to a respective micro-theory. A microsurvey can be created by the cognitive intelligence platform **102** for several different purposes, including: completing a user profile, and informing a missing parameter during the process of answering an originating question.
[0271] In FIG. **8**C, the microsurvey **882** gathers information related to health history, such as “when did you last see a doctor or other health professional to evaluate your health” where corresponding answers selectable by the user include specifying a month and year, “don't recall,” and “haven't had an appointment” (element **884**). An additional question asks “Which listed characteristics or conditions are true for you now? In the past?” where corresponding answers selectable by the user include “Diabetes during pregnancy,” “Over Weight,” “Insomnia,” and “Allergies” (element **886**). Each of the corresponding answer in element **886** also includes the option to indicate whether the characteristics or conditions are true for the user “Now”, “Past,” or “Current Treatment.”
[0272] In FIG. **8**D, aspects of educating a user are shown in the screen shot **890**. The screen shot displays an article titled “Diabetes: Preventing High Blood Sugar Emergencies,” and proceeds to describe when high blood sugar occurs and other information related to high blood sugar. The content displayed in the screen shot **890** is searchable and hearable as a podcast.
[0273] Accordingly, the cognitive agent **110** can answer a library of questions and provide content for many questions a user has as it related to diabetes. The information provided for purposes of educating a user is based on an overall health plan of the user, which is based on meta data analysis of interactions with the user, and an analysis of the education level of the user.
[0274] FIGS. **9**A-**9**B illustrate aspects of a conversational stream, in accordance with various embodiments. In particular, FIG. **9**A displays an example conversational stream between a user and the cognitive agent **110**. The screen shot **902** is an example of a dialogue that unfolds between a user and the cognitive agent **110**, after the user has registered with the cognitive intelligence platform **102**. In the screen shot **902**, the cognitive agent **110** begins by stating “Welcome, would you like to watch a video to help you better understand my capabilities” (element **904**). The cognitive agent provides an option to watch the video (element **906**). In response, the user inputs text “that's quite impressive” (element **908**). In various embodiments, the user inputs text using the input box **916**, which instructs the user to “Talk to me or type your question”.
[0275] Next, the cognitive agent **110** says “Thank you. I look forward to helping you meet your health goals!” (element **910**). At this point, the cognitive agent **110** can probe the user for additional data by offering a health assessment survey (e.g., a microsurvey) (element **914**). The cognitive

agent **110** prompts the user to fill out the health assessment by stating: “To help further personalize your health improvement experience, I would like to start by getting to know you and your health priorities. The assessment will take about 10 minutes. Let's get started!” (element **912**).

[0276] In FIG. **9**B, an additional conversational stream between the user and the cognitive agent **110** is shown. In this example conversational stream, the user previously completed a health assessment survey. The conversational stream can follow the example conversational stream discussed in FIG. **9**A.

[0277] In the screen shot **918**, the cognitive agent acknowledges the user's completion of the health assessment survey (element **920**) and provides additional resources to the user (element **922**). In element **920**, the cognitive agent states: “Congrats on taking the first step toward better health! Based upon your interest, I have some recommended health improvement initiatives for you to consider,” and presents the health improvement initiatives. In the example conversational stream, the user gets curious about a particular aspect of his health and states: “While I finished my health assessment, it made me remember that a doctor I saw before moving here told me that my blood sugar test was higher than normal.” (element **924**). After receiving the statement in element **924**, the cognitive agent **110** treats the statement as an originating question and undergoes an initial round of analysis (and additional rounds of analysis as needed) as described above.

[0278] The cognitive agent **110** presents an answer as shown in screen shot **926**. For example, the cognitive agent **110** states: “You mentioned in your health assessment that you have been diagnosed with Diabetes, and my health plan can help assure your overall compliance” (element **928**). The cognitive agent further adds: “The following provides you a view of our health plan which builds upon your level of understanding as well as additional recommendations to assist in monitoring your blood sugar levels” (element **930**). The cognitive agent **110** provides the user with the option to view his Diabetes Health Plan (element **932**).

[0279] The user responds “That would be great, how do we get started” (element **934**). The cognitive agent **110** receives the user's response as another originated question and undergoes an initial round of analysis (and additional rounds of analysis as needed) as described above. In the example screen shot **926**, the cognitive agent **110** determines additional information is needed and prompts the user for additional information.

[0280] FIG. **10** illustrates an additional conversational stream, in accordance with various embodiments. In particular, in the screen shot **1000**, the cognitive agent **110** elicit feedback (element **1002**) to determine whether the information provided to the user was useful to the user.

[0281] FIG. **11** illustrates aspects of an action calendar, in accordance with various embodiments. The action calendar is managed through the conversational stream between the cognitive agent **110** and the user. The action calendar aligns to care and wellness protocols, which are personalized to the risk condition or wellness needs of the user. The action calendar is also contextually aligned (e.g., what is being required or searched by the user) and hyper local (e.g., aligned to events and services provided in the local community specific to the user).

[0282] FIG. **12** illustrates aspects of a feed, in accordance with various embodiments. The feed allows a user to explore new opportunities and celebrate achieving goals (e.g., therapeutic or wellness goals). The feed provides a searchable interface (element **1202**).

[0283] The feed provides an interface where the user accesses a personal log of activities the user is involved in. The personal log is searchable. For example, if the user reads an article recommended by the cognitive agent **110** and highlights passages, the highlighted passages are accessible through the search. Additionally, the cognitive agent **110** can initiate a conversational stream focused on subject matter related to the highlighted passages.

[0284] The feed provides an interface to celebrate mini achievements and successes in the user's personal goals (e.g., therapeutic or wellness goals). In the feed, the cognitive agent **110** is still available (ribbon **1204**) to help search, guide, or steer the user toward a therapeutic or wellness goal.

[0285] FIG. **13** illustrates aspects of a hyper-local community, in accordance with various embodiments. A hyper-local community is a digital community that is health and wellness focused and encourages the user to find opportunities for themselves and get involved in a community that is physically close to the user. The hyper-local community allows a user to access a variety of care and wellness resources within his community and example recommendations include: Nutrition; Physical Activities; Healthcare Providers; Educations; Local Events; Services; Deals and Stores; Charities; and Products offered within the community. The cognitive agent **110** optimizes suggestions which help the user progress towards a goal as opposed to providing open ended access to hyper-local assets. The recommendations are curated and monitored for relevance to the user, based on the user's goals and interactions between the user and the cognitive agent **110**.
[0286] Accordingly, the cognitive intelligence platform provides several core features including:
[0287] 1) the ability to identify an appropriate action plan using narrative style interactions that generates data that includes intent and causation and using narrative style interactions; [0288] 2) monitoring: integration of offline to online clinical results across the functional medicine clinical standards; [0289] 3) the knowledge cloud that includes a comprehensive knowledge base of thousands of health related topics, an educational guide to better health aligned to western and eastern culture; [0290] 4) coaching using artificial intelligence; and [0291] 5) profile and health store that offers a holistic profile of each consumers health risks and interactions, combined with a repository of services, products, lab tests, devices, deals, supplements, pharmacy & telemedicine.
[0292] FIG. **14** illustrates a detailed view of a computing device **1400** that can be used to implement the various components described herein, according to some embodiments. In particular, the detailed view illustrates various components that can be included in the user device **104** illustrated in FIG. **1**, as well as the several computing devices implementing the cognitive intelligence platform **102**. As shown in FIG. **14**, the computing device **1400** can include a processor **1402** that represents a microprocessor or controller for controlling the overall operation of the computing device **1400**. The computing device **1400** can also include a user input device **1408** that allows a user of the computing device **1400** to interact with the computing device **1400**. For example, the user input device **1408** can take a variety of forms, such as a button, keypad, dial, touch screen, audio input interface, visual/image capture input interface, input in the form of sensor data, and so on. Still further, the computing device **1400** can include a display **1410** that can be controlled by the processor **1402** to display information to the user. A data bus **1416** can facilitate data transfer between at least a storage device **1440**, the processor **1402**, and a controller **1413**. The controller **1413** can be used to interface with and control different equipment through an equipment control bus **1414**. The computing device **1400** can also include a network/bus interface **1411** that couples to a data link **1412**. In the case of a wireless connection, the network/bus interface **1411** can include a wireless transceiver.
[0293] As noted above, the computing device **1400** also includes the storage device **1440**, which can comprise a single disk or a collection of disks (e.g., hard drives), and includes a storage management module that manages one or more partitions within the storage device **1440**. In some embodiments, storage device **1440** can include flash memory, semiconductor (solid-state) memory or the like. The computing device **1400** can also include a Random-Access Memory (RAM) **1420** and a Read-Only Memory (ROM) **1422**. The ROM **1422** can store programs, utilities or processes to be executed in a non-volatile manner. The RAM **1420** can provide volatile data storage, and stores instructions related to the operation of processes and applications executing on the computing device.
[0294] FIG. **15** shows a method (**1500**), in accordance with various embodiments, for answering a user-generated natural language medical information query based on a diagnostic conversational template.
[0295] In the method as shown in FIG. **15**, an artificial intelligence-based diagnostic conversation agent receives a user-generated natural language medical information query as entered by a user

through a user interface on a computer device (FIG. **15**, block **1502**). In some embodiments, the artificial intelligence-based diagnostic conversation agent is the conversation agent **110** of FIG. **1**. In some embodiments the computer device is the mobile device **104** of FIG. **1**. One example of a user-generated natural language medical information query as entered by a user through a user interface is the question "Is a blood sugar of 90 normal?" as shown in line **402** of FIG. **4**. In some embodiments, receiving a user-generated natural language medical information query as entered by a user through a user interface on a computer device (FIG. **15**, block **1502**) is Step 1 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0296] In response to the user-generated natural language medical information query, the artificial intelligence-based diagnostic conversation agent selects a diagnostic fact variable set relevant to generating a medical advice query answer for the user-generated natural language medical information query by classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications associated with respective diagnostic fact variable sets (FIG. **15**, block **1504**). In some embodiments, the artificial intelligence-based diagnostic conversation agent selecting a diagnostic fact variable set relevant to generating a medical advice query answer for the user-generated natural language medical information query by classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications associated with respective diagnostic fact variable sets (FIG. **15**, block **1504**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0297] FIG. **15** further shows compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set (FIG. **15**, block **1506**). Compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set (FIG. **15**, block **1506**) may include one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0298] In response to the user-specific medical fact variable values, the artificial intelligence-based diagnostic conversation agent generates a medical advice query answer in response to the user-generated natural language medical information query (FIG. **15**, block **1508**). In some embodiments, this is Step 7 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0299] In some embodiments, compiling user-specific medical fact variable values (FIG. **15**, block **1506**) includes extracting a first set of user-specific medical fact variable values from a local user medical information profile associated with the user-generated natural language medical information query and requesting a second set of user specific medical fact variable values through natural-language questions sent to the user interface on the mobile device (e.g. the microsurvey data **206** of FIG. **2** that came from the microsurvey **116** of FIG. **1**). The local user medical information profile can be the profile as generated in FIG. **7**A at block **708**.

[0300] In some embodiments, compiling user-specific medical fact variable values (FIG. **15**, block **1506**) includes extracting a third set of user-specific medical fact variable values that are lab result values from the local user medical information profile associated with the user generated natural language medical information query. The local user medical information profile can be the profile as generated in FIG. **7**A at block **708**.

[0301] In some embodiments, compiling user-specific medical fact variable values (FIG. **15**, block **1506**) includes extracting a fourth set of user-specific medical variable values from a remote medical data service profile associated with the local user medical information profile. The remote medical data service profile can be the service provider data **202** of FIG. **2**, which can come from the service provider **112** of FIG. **1**. The local user medical information profile can be the profile as

generated in FIG. **7**A at block **708**.
[0302] In some embodiments, compiling user-specific medical fact variable values (FIG. **15**, block **1506**) includes extracting a fifth set of user-specific medical variable values from demographic characterizations provided by a remote data service analysis of the local user medical information profile. The remote demographic characterizations can be the service provider data **202** of FIG. **2**, which can come from the service provider **112** of FIG. **1**. The local user medical information profile can be the profile as generated in FIG. **7**A at block **708**.
[0303] In some embodiments, generating the medical advice query answer (FIG. **15**, block **1508**) includes providing a treatment action-item recommendation in response to user-specific medical fact values that may be non-responsive to the medical question presented in the user-generated natural language medical information query. Such an action could define an action plan based on the data compiled (FIG. **15**, block **1506**), as shown in FIG. **7**C, block **758**.
[0304] In some embodiments, generating the medical advice query answer (FIG. **15**, block **1506**) includes providing a medical education media resource in response to user-specific medical fact variable values that may be non-responsive to the medical question presented in the user-generated natural language medical information query. Such an action could serve to educate and inform the user, as in block **758** of FIG. **7**C.
[0305] In some embodiments, selecting a diagnostic fact variable set relevant to generating a medical advice query answer for the user-generated natural language medical information query by classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications associated with respective diagnostic fact variable sets (FIG. **15**, block **1504**) includes classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications based on relevance to the local user medical information profile associated with the user-generated natural language medical information query. The local user medical information profile can be the profile as generated in FIG. **7**A at block **708**.
[0306] In some embodiments, the method (**1500**) for answering a user-generated natural language medical information query based on a diagnostic conversational template is implemented as a computer program product in a computer-readable medium.
[0307] In some embodiments, the system and method **1500** shown in FIG. **15** and described above is implemented on the computing device **1400** shown in FIG. **14**.
[0308] FIG. **16** shows a method (**1600**), in accordance with various embodiments, for answering a user-generated natural language query based on a conversational template.
[0309] In the method as shown in FIG. **16**, an artificial intelligence-based conversation agent receives a user-generated natural language query as entered by a user through a user interface (FIG. **16**, block **1602**). In some embodiments, the artificial intelligence-based conversation agent is the conversation agent **110** of FIG. **1**. In some embodiments, the user interface is on a computer device. In some embodiments the computer device is the mobile device **104** of FIG. **1**. One example of a user-generated natural language query as entered by a user through a user interface is the question "Is a blood sugar of 90 normal?" as shown in line **402** of FIG. **4**. In some embodiments, receiving a user-generated natural language query as entered by a user through a user interface on a computer device (FIG. **16**, block **1602**) is Step 1 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0310] In response to the user-generated natural language query, the artificial intelligence-based conversation agent selects a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets (FIG. **16**, block **1604**). In some embodiments, the artificial intelligence-based conversation agent selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of

domain-directed query classifications associated with respective fact variable sets (FIG. **16**, block **1604**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0311] FIG. **16** further shows compiling user-specific variable values for one or more respective fact variables of the fact variable set (FIG. **16**, block **1606**). Compiling user-specific fact variable values for one or more respective fact variables of the fact variable set (FIG. **16**, block **1606**) may include one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0312] In response to the user-specific fact variable values, the artificial intelligence-based conversation agent generates a query answer in response to the user-generated natural language query (FIG. **16**, block **1608**). In some embodiments, this is Step 7 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0313] In some embodiments, compiling user-specific fact variable values (FIG. **16**, block **1606**) includes extracting a first set of user-specific fact variable values from a local user profile associated with the user-generated natural language query and requesting a second set of user specific variable values through natural-language questions sent to the user interface on the mobile device (e.g. the microsurvey data **206** of FIG. **2** that came from the microsurvey **116** of FIG. **1**). The local user profile can be the profile as generated in FIG. **7**A at block **708**. In some embodiments, the natural language questions sent to the user interface on the mobile device can be a part of a conversation template.

[0314] In some embodiments, compiling user-specific fact variable values (FIG. **16**, block **1606**) includes extracting a third set of user-specific fact variable values that are test result values from the local user profile associated with the user generated natural language query. The local user profile can be the profile as generated in FIG. **7**A at block **708**. In some embodiments, compiling user-specific fact variable values (FIG. **16**, block **1606**) includes extracting a fourth set of user-specific variable values from a remote data service profile associated with the local user profile. The remote data service profile can be the service provider data **202** of FIG. **2**, which can come from the service provider **112** of FIG. **1**. The local user profile can be the profile as generated in FIG. **7**A at block **708**.

[0315] In some embodiments, compiling user-specific fact variable values (FIG. **16**, block **1606**) includes extracting a fifth set of user-specific variable values from demographic characterizations provided by a remote data service analysis of the local user profile. The remote demographic characterizations can be the service provider data **202** of FIG. **2**, which can come from the service provider **112** of FIG. **1**. The local user profile can be the profile as generated in FIG. **7**A at block **708**.

[0316] In some embodiments, generating the query answer (FIG. **16**, block **1608**) includes providing an action-item recommendation in response to user-specific fact values that may be non-responsive to the question presented in the user-generated natural language query. Such an action could define an action plan based on the data compiled (FIG. **16**, block **1606**), as shown in FIG. **7**C, block **758**.

[0317] In some embodiments, generating the advice query answer (FIG. **16**, block **1606**) includes providing an education media resource in response to user-specific fact variable values that may be non-responsive to the question presented in the user-generated natural language query. Such an action could serve to educate and inform the user, as in block **758** of FIG. **7**C.

[0318] In some embodiments, selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets (FIG. **16**, block **1604**) includes classifying the user-generated natural language query into one of a set of domain-directed query classifications based on relevance to the local user profile associated with the user-generated natural language query. The local user profile can be the profile

as generated in FIG. **7**A at block **708**.
[0319] In some embodiments, the method (**1600**) for answering a user-generated natural language query based on a conversational template is implemented as a computer program product in a computer-readable medium.
[0320] In some embodiments, the system and method shown in FIG. **16** and described above is implemented in the cognitive intelligence platform **102** shown in FIG. **1**.
[0321] In the cognitive intelligence platform **102**, a cognitive agent **110** is configured for receiving a user-generated natural language query at an artificial intelligence-based conversation agent from a user interface on a user device **104** (FIG. **16**, block **1602**).
[0322] A critical thinking engine **108** is configured for, responsive to content of the user-generated natural language query, selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets (FIG. **16**, block **1604**).
[0323] Included is a knowledge cloud **106** that compiles user-specific fact variable values for one or more respective fact variables of the fact variable set (FIG. **16**, block **1606**).
[0324] Responsive to the fact variable values, the cognitive agent **110** is further configured for generating the query answer in response to the user-generated natural language query (FIG. **16**, block **1606**).
[0325] In some embodiments, the system and method **1600** shown in FIG. **16** and described above is implemented on the computing device **1400** shown in FIG. **14**.
[0326] FIG. **17** shows a computer-implemented method **1700** for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system. In some embodiments, the method **1700** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**.
[0327] The method **1700** involves receiving a user-generated natural language medical information query from a medical conversational user interface at an artificial intelligence-based medical conversation cognitive agent (block **1702**). In some embodiments, receiving a user-generated natural language medical information query from a medical conversational user interface at an artificial intelligence-based medical conversation cognitive agent (block **1702**) is performed by a cognitive agent that is a part of the cognitive intelligence platform and is configured for this purpose. In some embodiments, the artificial intelligence-based diagnostic conversation agent is the conversation agent **110** of FIG. **1**. One example of a user-generated natural language medical information query is “Is a blood sugar of 90 normal?” as shown in line **402** of FIG. **4**. In some embodiments, the user interface is on the mobile device **104** of FIG. **1**. In some embodiments, receiving a user-generated natural language medical information query from a medical conversational user interface at an artificial intelligence-based medical conversation cognitive agent (block **1702**) is Step 1 as earlier discussed in the context of “Analyzing Conversational Context As Part of Conversational Analysis”.
[0328] The method **1700** further includes extracting a medical question from a user of the medical conversational user interface from the user-generated natural language medical information query (block **1704**). In some embodiments, extracting a medical question from a user of the medical conversational user interface from the user-generated natural language medical information query (block **1704**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, extracting a medical question from a user of the medical conversational user interface from the user-generated natural language medical information query (block **1704**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of “Analyzing

Conversational Context As Part of Conversational Analysis".
[0329] The method **1700** includes compiling a medical conversation language sample (block **1706**). In some embodiments, compiling a medical conversation language sample (block **1706**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The medical conversation language sample can include items of health-information-related-text derived from a health-related conversation between the artificial intelligence-based medical conversation cognitive agent and the user. In some embodiments compiling a medical conversation language sample (block **1706**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0330] The method **1700** involves extracting internal medical concepts and medical data entities from the medical conversation language sample (block **1708**). In some embodiments, extracting internal medical concepts and medical data entities from the medical conversation language sample (block **1708**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The internal medical concepts can include descriptions of medical attributes of the medical data entities. In some embodiments, extracting internal medical concepts and medical data entities from the medical conversation language sample (block **1708**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0331] The method **1700** involves inferring a therapeutic intent of the user from the internal medical concepts and the medical data entities (block **1710**). In some embodiments, inferring a therapeutic intent of the user from the internal medical concepts and the medical data entities (block **1710**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, inferring a therapeutic intent of the user from the internal medical concepts and the medical data entities (block **1710**) is accomplished as in Step 2 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0332] The method **1700** includes generating a therapeutic paradigm logical framework **1800** for interpreting of the medical question (block **1712**). In some embodiments, generating a therapeutic paradigm logical framework **1800** for interpreting of the medical question (block **1712**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, generating a therapeutic paradigm logical framework **1800** for interpreting of the medical question (block **1712**) is accomplished as in Step 5 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0333] FIG. **18** shows an example therapeutic paradigm logical framework **1800**. The therapeutic paradigm logical framework **1800** includes a catalog **1802** of medical logical progression paths **1804** from the medical question **1806** to respective therapeutic answers **1810**.
[0334] Each of the medical logical progression paths **1804** can include one or more medical logical linkages **1808** from the medical question **1806** to a therapeutic path-specific answer **1810**.
[0335] The medical logical linkages **1808** can include the internal medical concepts **1812** and external therapeutic paradigm concepts **1814** derived from a store of medical subject matter ontology data **1816**. In some embodiments, the store of subject matter ontology data **1816** is contained in a knowledge cloud. In some embodiments, the knowledge cloud is the knowledge cloud **102** of FIGS. **1** and **2**. In some embodiments, the subject matter ontology data **1816** is the subject matter ontology data **216** of FIG. **2**. In some embodiments, the subject matter ontology data **1816** includes the subject matter ontology **300** of FIG. **3**.
[0336] The method **1700** shown in FIG. **17** further includes selecting a likely medical information path from among the medical logical progression paths **1804** to a likely path-dependent medical

information answer based at least in part upon the therapeutic intent of the user (block **1714**). In some embodiments, selecting a likely medical information path from among the medical logical progression paths **1804** to a likely path-dependent medical information answer based at least in part upon the therapeutic intent of the user (block **1714** is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The selection can also be based in part upon the sufficiency of medical diagnostic data to complete the medical logical linkages **1808**. In some embodiments, selection can also be based in part upon the sufficiency of medical diagnostic data to complete the medical logical linkages **1808** can be performed by a critical thinking engine that is further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The medical diagnostic data can include user-specific medical diagnostic data. The selection can also be based in part upon treatment sub-intents including tactical constituents related to the therapeutic intent of the user by the store of medical subject matter ontology data **1816**. In some embodiments, selection based in part upon treatment sub-intents including tactical constituents related to the therapeutic intent of the user by the store of medical subject matter ontology data **1816** can be performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The selection can further occur after requesting additional medical diagnostic data from the user. An example of requesting additional medical diagnostic data from the user is shown in FIG. **4** on line **406** “I need some additional information in order to answer this question, was this an in-home glucose test or was it done by a lab or testing service”. In some embodiments, the process of selection after requesting additional medical diagnostic data from the user can be performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, selecting a likely medical information path from among the medical logical progression paths **1804** to a likely path-dependent medical information answer based at least in part upon the therapeutic intent of the user (block **1714**) is accomplished through one or more of Steps 5-6 as earlier discussed in the context of “Analyzing Conversational Context As Part of Conversational Analysis”.

[0337] The method **1700** involves answering the medical question by following the likely medical information path to the likely path-dependent medical information answer (block **1716**). In some embodiments, answering the medical question by following the likely medical information path to the likely path-dependent medical information answer (block **1716**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, answering the medical question by following the likely medical information path to the likely path-dependent medical information answer (block **1716**) is accomplished as in Step 7 as earlier discussed in the context of “Analyzing Conversational Context As Part of Conversational Analysis”.

[0338] The method **1700** can further include relating medical inference groups of the internal medical concepts. In some embodiments, relating medical inference groups of the internal medical concepts is performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. Relating medical inference groups of the internal medical concepts can be based at least in part on shared medical data entities for which each internal medical concept of a medical inference group of internal medical concepts describes a respective medical data attribute. In some embodiments, relating medical inference groups of the internal medical concepts based at least in part on shared medical data entities for which each internal medical concept of a medical inference group of internal medical concepts describes a respective medical data attribute can be performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**.

[0339] In some embodiments, the method **1700** of FIG. **17** is implemented as a computer program product in a computer-readable medium.
[0340] FIG. **19** shows a computer-implemented method **1900** for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system. In some embodiments, the method **1900** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**.
[0341] The method **1900** involves receiving a user-generated natural language query at an artificial intelligence-based conversation agent (block **1902**). In some embodiments, receiving a user-generated natural language query from a conversational user interface at an artificial intelligence-based conversation cognitive agent (block **1902**) is performed by a cognitive agent that is a part of the cognitive intelligence platform and is configured for this purpose. In some embodiments, the artificial intelligence-based conversation agent is the conversation agent **110** of FIG. **1**. One example of a user-generated natural language query is "Is a blood sugar of 90 normal?" as shown in line **402** of FIG. **4**. In some embodiments, the user interface is on the mobile device **104** of FIG. **1**. In some embodiments, receiving a user-generated natural language query from a conversational user interface at an artificial intelligence-based conversation cognitive agent (block **1902**) is Step 1 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0342] The method **1900** further includes extracting a question from a user of the conversational user interface from the user-generated natural language query (block **1904**). In some embodiments, extracting a question from a user of the conversational user interface from the user-generated natural language query (block **1904**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, extracting a question from a user of the conversational user interface from the user-generated natural language query (block **1904**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0343] The method **1900** includes compiling a language sample (block **1906**). In some embodiments, compiling a language sample (block **1906**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The language sample can include items of health-information-related-text derived from a health-related conversation between the artificial intelligence-based conversation cognitive agent and the user. In some embodiments compiling a language sample (block **1906**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0344] The method **1900** involves extracting internal concepts and entities from the language sample (block **1908**). In some embodiments, extracting internal concepts and entities from the language sample (block **1908**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The internal concepts can include descriptions of attributes of the entities. In some embodiments, extracting internal concepts and entities from the language sample (block **1908**) is accomplished through one or more of Steps 2-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0345] The method **1900** involves inferring an intent of the user from the internal concepts and the entities (block **1910**). In some embodiments, inferring an intent of the user from the internal concepts and the entities (block **1910**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, inferring an intent of the user from the internal concepts and the

entities (block **1910**) is accomplished as in Step 2 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0346] The method **1900** includes generating a logical framework **2000** for interpreting of the question (block **1912**). In some embodiments, generating a logical framework **2000** for interpreting of the question (block **1912**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, generating a logical framework **2000** for interpreting of the question (block **1912**) is accomplished as in Step 5 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0347] FIG. **20** shows an example logical framework **2000**. The logical framework **2000** includes a catalog **2002** of paths **2004** from the question **2006** to respective answers **2010**.
[0348] Each of the paths **2004** can include one or more linkages **2008** from the question **2006** to a path-specific answer **2010**.
[0349] The linkages **2008** can include the internal concepts **2012** and external concepts **2014** derived from a store of subject matter ontology data **2016**. In some embodiments, the store of subject matter ontology data **2016** is contained in a knowledge cloud. In some embodiments, the knowledge cloud is the knowledge cloud **102** of FIGS. **1** and **2**. In some embodiments, the subject matter ontology data **2016** is the subject matter ontology data **216** of FIG. **2**. In some embodiments, the subject matter ontology data **2016** includes the subject matter ontology **300** of FIG. **3**.
[0350] The method **1900** shown in FIG. **19** further includes selecting a likely path from among the paths **2004** to a likely path-dependent answer based at least in part upon the intent of the user (block **1914**). In some embodiments, selecting a likely path from among the paths **2004** to a likely path-dependent answer based at least in part upon the intent of the user (block **1914** is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The selection can also be based in part upon the sufficiency of data to complete the linkages **2008**. In some embodiments, selection can also be based in part upon the sufficiency of data to complete the linkages **2008** can be performed by a critical thinking engine that is further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The data can include user-specific data. The selection can also be based in part upon treatment sub-intents including tactical constituents related to the intent of the user by the store of subject matter ontology data **2016**. In some embodiments, selection based in part upon treatment sub-intents including tactical constituents related to the intent of the user by the store of subject matter ontology data **2016** can be performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. The selection can further occur after requesting additional data from the user. An example of requesting additional data from the user is shown in FIG. **4** on line **406** "I need some additional information in order to answer this question, was this an in-home glucose test or was it done by a lab or testing service". In some embodiments, the process of selection after requesting additional data from the user can be performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, selecting a likely path from among the paths **2004** to a likely path-dependent answer based at least in part upon the intent of the user (block **1914**) is accomplished through one or more of Steps 5-6 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".
[0351] The method **1900** involves answering the question by following the likely path to the likely path-dependent answer (block **1916**). In some embodiments, answering the question by following the likely path to the likely path-dependent answer (block **1916**) is performed by a critical thinking engine configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. In some embodiments, answering the question by following the

likely path to the likely path-dependent answer (block **1916**) is accomplished as in Step 7 as earlier discussed in the context of "Analyzing Conversational Context As Part of Conversational Analysis".

[0352] The method **1900** can further include relating inference groups of the internal concepts. In some embodiments, relating inference groups of the internal concepts is performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**. Relating inference groups of the internal concepts can be based at least in part on shared entities for which each internal concept of an inference group of internal concepts describes a respective data attribute. In some embodiments, relating inference groups of the internal concepts based at least in part on shared entities for which each internal concept of an inference group of internal concepts describes a respective data attribute can be performed by a critical thinking engine further configured for this purpose. In some embodiments, the critical thinking engine is the critical thinking engine **108** of FIG. **1**.

[0353] In some embodiments, the method **1900** of FIG. **19** is implemented as a computer program product in a computer-readable medium.

[0354] FIG. **21** shows a computer-implemented method **2100** for generated cognified data using unstructured data. In some embodiments, the method **2100** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **2100** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.

[0355] At block **2102**, the processing device may receive, at an artificial intelligence engine, a corpus of data for a patient. The corpus of data may represent unstructured data. The corpus of data may include a set of strings of characters. The corpus of data may be patient notes in an electronic medical record entered by a physician. In some embodiments, an application programming interface (API) may be used to interface with an electronic medical record system used by the physician. The API may retrieve one or more EMRs of the patient and extract the patient notes. The artificial intelligence engine may include the one or more machine learning models trained to generate cognified data based on unstructured data.

[0356] At block **2104**, the processing device may identify indicia. The indicia may be identified by processing the strings of characters. The indicia may include a phrase, a predicate, a subject, an object (e.g., direct, indirect), a keyword, a cardinal, a number, a concept, an objective, a noun, a verb, or some combination thereof.

[0357] At block **2106**, the processing device may compare the indicia to a knowledge graph representing known health related information to generate a possible health related information pertaining to the patient. In some embodiments, the indicia may be compared to numerous knowledge graphs each representing a different medical conditions. As discussed herein, the knowledge graphs may include respective nodes that include different known health related information about the medical conditions, and a logical structure that includes predicates that correlate the information in the respective knowledge graphs. The knowledge graphs and the logical structures may be generated by the one or more trained machine learning models using the known health related information. The knowledge graph may represent knowledge of a disease and the knowledge graph may include a set of concepts pertaining to the disease obtained from the known health related information and also includes relationships between the set of concepts. The known health related information associated with the nodes may be facts, concepts, complications, risks, causal effects, etc. pertaining to the medical conditions (e.g., diseases) represented by the knowledge graphs. The processing device may codify evidence-based health related guidelines pertaining to the diseases to generate the logical structures. The generated possible health related information may be a tag that is associated with the indicia in the unstructured data.

[0358] At block **2108**, the processing device may identify, using the logical structure, a structural similarity of the possible health related information and a known predicate in the logical structure. The structural similarity may be used to identify a certain pattern. The pattern may pertain to treatment, quality of care, risk adjustment, orders, referral, education and content patterns, and the like. The structural similarity and/or the pattern may be used to cognify the corpus of data.
[0359] At block **2110**, the processing device may generate, by the artificial intelligence engine, cognified data based on the structural similarity. In some embodiments, the cognified data may include a health related summary of the possible health related information. The health related summary may include conclusions, concepts, recommendations, identified gaps in the treatment plan, identified gaps in risk analysis, identified gaps in quality of care, and so forth pertaining to one or more medical conditions represented by one or more knowledge graphs that include the logic structure having the known predicate that is structurally similar to the possible health related information.
[0360] In some embodiments, generating the cognified data may include generating at least one new string of characters representing a statement pertaining to the possible health related information. Also, the artificial intelligence engine executed by the processing device may include the at least one new string of characters in the health related summary of the possible health related information. The statement may include a concept, conclusion, and/or recommendation pertaining to the possible health related information. The statement may describe an effect that results from the possible health related information.
[0361] FIG. **22** shows a method **2200** for identifying missing information in a corpus of data, in accordance with various embodiments. In some embodiments, the method **2300** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **2200** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.
[0362] At block **2202**, the processing device executing the artificial intelligence engine may identify at least one piece of information missing in the corpus of data for the patient using the cognified data. The at least one piece of information pertains to a treatment gap, a risk, gap, a quality of care gap, or some combination thereof.
[0363] At block **2204**, the processing device may cause a notification to be presented on a computing device of a healthcare personnel (e.g., physician). The notification may instruct entry of the at least one piece of information into the corpus of data (e.g., patient notes in the EMR). For example, if certain symptoms are described for a patient in the corpus of data and those symptoms are known to result from a certain medication currently prescribed to the patient, but the corpus of data does not indicate switching medications, then the at least one piece of information may identify a treatment gap and recommend switching medications to one that does not cause those symptoms.
[0364] FIG. **23** shows a method **2300** for using feedback pertaining to the accuracy of cognified data to update an artificial intelligence engine, in accordance with various embodiments. In some embodiments, the method **2300** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **2300** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.
[0365] At block **2302**, the processing device may receive feedback pertaining to whether the cognified data is accurate. For example, the physician may be presented with the cognified data on a computing device, and the physician may review the cognified data. The physician may be

presented with options to verify the accuracy of portions or all of the cognified data for the particular patient. For example, the physician may select a first graphical element (e.g., button, checkbox, etc.) next to portions of the cognified data that are accurate and may select a second graphical element next to portions of the cognified data that are inaccurate. If the second graphical element is selected, an input box may appear and a notification may be presented to provide a reason why the portion is inaccurate and to provide corrected information. The feedback may be transmitted to the cognitive intelligence platform.

[0366] At block **2304**, the processing device may update the artificial intelligence engine based on the feedback. A closed-loop feedback system may be implemented using these techniques. The feedback may enhance the accuracy of the cognified data as the artificial intelligence engine continues to learn and improve.

[0367] FIG. **24**A shows a block diagram for using the knowledge graph **500** to generate possible health related information, in accordance with various embodiments. As depicted, a physician may have entered patient notes **2400** in one or more electronic medical records (EMRs). The EMRs may be provided directly to the cognitive intelligence engine **102** and/or retrieved using an application programming interface (API) from an EMR system used by the physician. The patient notes may be extracted from the EMRs. In some embodiments, numerous patient notes from numerous consultations may be processed, synthesized, and cognified using the disclosed techniques. In some embodiments, patient notes from a single consultation may be processed, synthesized, and cognified using the disclosed techniques. The patient notes may include a set of strings of characters that arranged in sentences, phrases, and/or paragraphs. The cognitive intelligence platform **102** may process the set of strings of characters to identify indicia comprising a phrase, a predicate, a keyword, a subject, an object, a cardinal, a number, a concept, or some combination thereof.

[0368] The cognitive intelligence platform **102**, and in particular the artificial intelligence engine **109**, may compare the indicia to numerous knowledge graphs **500** each representing a respective medical condition, such as diabetes, cancer, coronary artery disease, arthritis, just to name a few examples. The artificial intelligence engine **109** may be trained to generate possible health related information by constructing logical structures based on matched indicia and known health related information (health artifacts that are established based on information from a trusted source) represented in the knowledge graphs **500**. The logical structures may be tagged to the indicia, as depicted in FIG. **24**A.

[0369] The artificial intelligence engine **109** may identify the following example indicia: “Patient X”, “sweating”, “blood glucose test”, “8 mmol/L blood sugar level”, “lost weight”, “diet the same”, “constantly tired”. The artificial intelligence engine **109** may match the indicia with known health related information in the knowledge graph **500**. For example, in the knowledge graph **500** depicted in FIG. **5**, “blood glucose test”, is a known health related artifact that is used to test for Type 2 Diabetes Mellitus. Thus, various logical structures may be constructed by the artificial intelligence engine **109** that states “blood glucose test is used to test Type 2 Diabetes Mellitus”, “Type 2 Diabetes Mellitus is diagnosed or monitored using blood glucose test” (tag **2402**), “blood glucose test measures blood sugar level”, and so forth.

[0370] The artificial intelligence engine **109** may generate other possible health related information for each of the indicia that matches known health related information in the knowledge graphs. For example, the artificial intelligence engine **109** generated example logical structure “Sweating is a symptom of medical condition Y” (tag **2404**) for the indicia “sweating”. The artificial intelligence engine **109** may generate other possible health related information for “sweating”, such as “sweating is caused by running”, “sweating is a symptom of fever”. Further, the artificial intelligence engine **109** may elaborate on the generated possible health related information by generating further possible health related information. Based on generating “sweating is a symptom of medical condition Y” (where Y is the name of the medical condition), the artificial intelligence

engine **109** may generate another logical structure “medical condition Y causes Z” (where Z is a health artifact such as another medical condition).
[0371] It should be understood that, although not shown, a logical structure may be included in the knowledge graph **500** that indicates “Type 2 Diabetes has normal blood sugar level 5-7 mmol/L”. An example possible health related information generated by the artificial intelligence engine **109** for the indicia “8 mmol/L blood sugar level” is “8 mmol/L blood sugar level is high blood sugar” (tag **2406**) based on comparing the indicia to the known health related information about acceptable blood sugar levels in the knowledge graph **500**. The artificial intelligence engine **109** may generate an additional possible health information based on tag **2406**, and the additional possible health information may state “Type 2 Diabetes Mellitus has symptom of high blood sugar” (tag **2408**).
[0372] An example possible health related information generated by the artificial intelligence engine **109** for the indicia “lost weight” may be “Weight loss is a symptom of medical condition Y” (tag **2410**) where medical condition Y is any medical condition that causes weight loss. For example, any knowledge graph that includes “weight loss”, “loss of weight”, or some variant thereof as a health artifact may be identified and one or more possible health related information may be generated indicating that weight loss is a symptom of the medical condition represented by that knowledge graph.
[0373] An example possible health related information generated by the artificial intelligence engine **109** for the indicia “constantly tired” may be “Constant fatigue is a symptom of medical condition Y” (tag **2412**) where medical condition Y is any medical condition that causes constant fatigue. For example, any knowledge graph that includes “fatigue”, “constant fatigue”, or some variant thereof as a health artifact may be identified and one or more possible health related information may be generated indicating that constant fatigue is a symptom of the medical condition represented by that knowledge graph.
[0374] The knowledge graphs that include a threshold number of matches between the indicia and the known health related matches in the knowledge graphs may be selected for further processing. The threshold may be any suitable number of matches. For example, in the depicted example, the knowledge graph **500** representing Type 2 Diabetes Mellitus may be selected because 3 tags (**2402**, **2406**, and **2408**) relate to that medical condition represented in the knowledge graph **500**.
[0375] FIG. **24**B shows a block diagram for using a logical structure to identify structural similarities with known predicates to generate cognified data, in accordance with various embodiments. The identification of structural similarities may be performed in parallel with the comparison of the indicia with the known health related information. In some embodiments, the generated possible health related information may be compared with the known predicates in the logical structures of the knowledge graphs. In some embodiments, predicates detected in the unstructured data may also be compared with the known predicates in the logical structures of the knowledge graphs. The artificial intelligence engine **500** may identify structural similarities between the possible health related information and the known predicates in the logical structures of the knowledge graphs. The artificial intelligence engine **500** may identify structural similarities between the detected predicates in the unstructured data and the known predicates in the logical structures of the knowledge graphs. In some embodiments, identifying structural similarities may refer to comparing the structure of the logical structure of the possible health related information to a known logical structure (known logical structure may refer to a logical structure established based on a trusted source), such as determining whether the subjects are the same or substantially similar, the predicates are the same or substantially similar, the objects are the same or substantially similar, and so forth.
[0376] For example, the knowledge graph **500** includes the logical structure “Type 2 Diabetes Mellitus has symptom high blood sugar”. Comparing the possible health related information represented by tag **2408** “Type 2 Diabetes Mellitus has symptom of high blood sugar” to the known logical structure in the knowledge graph **500** results in identifying a structurally similarity between

the two. Accordingly, the knowledge graph **500** may be selected for further processing.
[0377] In some embodiments, the structural similarities detected may be used to identify patterns. For example, a treatment pattern for diabetes may be detected if a blood glucose test is used, a patient is prescribed a certain medication, and the like. In some embodiments, gaps in the unstructured data may be identified based on the patterns detected. For example, if a person is determined to have a certain medical condition based on the treatment pattern identified, and it is known based on evidence-based guidelines that a certain medication should be prescribed for that treatment pattern, the artificial intelligence engine **109** may indicate there is a treatment gap if that medication has not been prescribed yet.
[0378] The knowledge graphs selected when comparing the indicia to the known health related information and the knowledge graphs selected when identifying structural similarities between the known logical structure and the possible health related information may be compared to determine whether there are overlaps. As discussed above, the knowledge graph **500** representing Type 2 Diabetes Mellitus overlaps as being selected during both operations. As a result, the knowledge graph **500** may be used for cognification. In some embodiments, any of the knowledge graphs selected during either operation may be used for cognification.
[0379] In some embodiments, the selected knowledge graphs may be used to generate cognified data **2450**. Further, the possible health related information and the matching logical structures may be used to generate the cognified data **2450**. The cognified data **2450** may include a health related summary of the possible health related information. In some embodiments, the cognified data **2450** may include conclusions, statements of facts, concepts, recommendations, identified gaps in the unstructured data that was processed, and the like.
[0380] In some embodiments, the cognified data **2450** may be used to generate a diagnosis of a medical condition for a patient. For example, if there are a threshold number of identified structural similarities between the known logical structures and the possible health related information and/or if there are a threshold number of matches between indicia and known health related information for a particular medical condition, a diagnosis may be generated for that particular medical condition. If there are numerous medical conditions identified after performing the cognification, the numerous medical conditions may be indicated as potential candidates for diagnosis. In the ongoing example, the knowledge graph **500** was selected as the overlapping knowledge graph and satisfies the threshold number of identified structural similarities and/or the threshold number of matches. Accordingly, a diagnosis that Patient X has Type 2 Diabetes Mellitus may be generated. The cognified data **2450** may include the diagnosis, as depicted.
[0381] When generating the cognified data, other health related information in the selected knowledge graph **500** that was not included in the unstructured data may be inserted. That is, sentences may be constructed using the known health related information and the predicates in the knowledge graph **50**. For example, the unstructured data did not indicate any information pertaining to complications of Type 2 Diabetes Mellitus. However, as depicted in the knowledge graph **500** of FIG. **5**, there is a logical structure that specifies “Type 2 Diabetes Mellitus has complications of stroke, coronary artery disease, diabetes foot problems, diabetic neuropathy, and/or diabetic retinopathy”. As depicted, this construction of the logical structure is included in the cognified data **2450** by the artificial intelligence engine **109**.
[0382] The cognified data **2450** may also include the tag **2406** (“8 mmol/L level of blood sugar is high blood sugar. Type 2 Diabetes Mellitus has symptom of high blood sugar”) that was generated for the unstructured data based on the known health information in the knowledge graph **500**. The artificial intelligence engine **109** may generate a recommendation based on the lost weight indicia indicated in the unstructured data. The recommendation may state “Re-measure weight at next appointment.” In addition, as discussed above, the artificial intelligence engine **109** may identify certain gaps. For example, the diagnosis that is generated indicates that the patient has Type 2 Diabetes Mellitus. The unstructured data does not indicate that medication is prescribed. However,

the knowledge graph **500** specifies that Type 2 Diabetes Mellitus is treated by "Diabetes Medicines". Accordingly, a treatment gap may be identified by the artificial intelligence engine **109** based on treatment patterns codified in the knowledge graph **500**, and a statement may be constructed and inserted in the cognified data **2450**. The statement may state "There is a treatment gap: the patient should be prescribed medication."

[0383] The cognified data **2450** may be transmitted by the cognitive intelligence platform **102** to a computing device of the service provider **112**, such as the physician who entered the unstructured data. As depicted, the cognified data **2450** may be instilled with intelligence, knowledge, and logic using the disclosed cognification techniques. The physician may quickly review the cognified data **2450** without having to review numerous patient notes from various EMRs. In some embodiments, the physician may be presented with options to verify portions or all of the cognified data **2450** is accurate. The feedback may be transmitted to the cognitive intelligence platform **102** and the artificial intelligence engine **109** may update its various machine learning models using the feedback.

[0384] FIG. **25** shows a method **2500** for providing first information pertaining to a possible medical condition of a patient to a computing device, in accordance with various embodiments. In some embodiments, the method **2500** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **2500** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.

[0385] At block **2502**, the processing device of a server may receive an electronic medical record (EMR) including notes pertaining to a patient. The EMR may be transmitted directly to the server from a computing device of the physician that entered the notes, and/or the EMR may be obtained using an application programming interface (API) interfacing with an EMR system used by the physician that entered the notes. In some embodiments, the server may receive text input by the patient. For example, the text input by the user may include symptoms the patient is experiencing and ask a question pertaining to what medical condition the patient may have. The operations of method **2500** may be used to similarly provide information to the patient based on identifying the possible medical condition using the cognification techniques.

[0386] At block **2504**, the processing device may process the notes to obtain indicia including a subject, an object, a word, a cardinal, a phrase, a concept, a sentence, a predicate, or some combination thereof. Textual analysis may be performed to extract the indicia. Processing the patient notes to obtain the indicia may further include inputting the notes into an artificial intelligence engine **109** trained to identify the indicia in text based on commonly used indicia pertaining to the possible medical condition. The artificial intelligence engine **109** may determine commonly used indicia for various medical conditions based on evidence-based guidelines, clinical trial results, physician research, or the like that are input to one or more machine learning models.

[0387] At block **2506**, the processing device may identify a possible medical condition of the patient by identifying a similarity between the indicia and a knowledge graph representing knowledge pertaining to the possible medical condition. The knowledge graph may include a set of nodes representing the set of information pertaining to the possible medical condition. The set of nodes may also include relationships (e.g., predicates) between the set of information pertaining to the possible medication condition.

[0388] In some embodiments, identifying the possible medical condition may include using a cognified data structure generated from the notes of the patient. The cognified data structure may include a conclusion based on a logic structure representing evidence-based guidelines pertaining to the possible medical condition.

[0389] In some embodiments, the similarity may pertain to a match between the indicia and a

health artifact (known health related information) included in the knowledge graph **500**. For example, “high blood pressure” may be extracted as indicia from the sentence “Patient X has high blood pressure”, and “high blood pressure” is a health artifact at a node in the knowledge graph **500** representing Type 2 Diabetes Mellitus.

[0390] In some embodiments, the similarity may pertain to a structural similarity between the logical structure (e.g., “Type 2 Diabetes has symptoms of High Blood Pressure) and the indicia (e.g., “Patient X has symptoms of High Blood Pressure”) that is included in the unstructured data. If the subject, predicates, and/or objects of the logical structure and the indicia match or substantially match (e.g., “has symptoms of High Blood Pressure” match between the logical structure and the indicia, also “Type 2 Diabetes has symptoms of High Blood Pressure” and “Patient X has symptoms of High Blood Pressure” substantially match), then the knowledge graph **500** including the logical structure is a candidate for a possible medical condition. In some embodiments, a combination of similarities identified between the match between the indicia and the health artifact and between the logical structure and the indicia may be used to identify a possible medical condition and/or cognify the unstructured data.

[0391] An artificial intelligence engine **109** may be used to identify the possible medical condition by identifying the similarity between the indicia and the knowledge graph. The artificial intelligence engine **109** may be trained using feedback from medical personnel. The feedback may pertain to whether output regarding the possible medical conditions from the artificial intelligence engine **109** are accurate for input including notes of patients.

[0392] At block **2508**, the processing device may provide, at a first time, first information of the set of information to a computing device of the patient for presentation of the computing device, the first information being associated with a root node of the set of nodes. In some embodiments, the first information may pertain to a name of the possible medical condition. As depicted in the knowledge graph **500** of FIG. **5**, the root node is associated with the name of the medical condition “Type 2 Diabetes Mellitus”. In some embodiments, the first information may pertain to a definition of the possible medical condition, instead of or in addition to the name of the possible medical condition.

[0393] FIG. **26** shows a method **2600** for providing second and third information pertaining to a possible medical condition of a patient to a computing device, in accordance with various embodiments. In some embodiments, the method **2600** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **2600** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.

[0394] At block **2602**, the processing device may provide, at a second time, second information of the set of information to the computing device of the patient for presentation on the computing device. The second information may be associated with a second node of the set of nodes, and the second time may be after the first time. The second information may be different than the first information. The second information may pertain to how the possible medical condition affects people, signs and symptoms of the possible medical condition, a way to treat the possible medical condition, a progression of the possible medical condition, complications of the possible medical condition, or some combination thereof. The second time may be selected based on when the second information is relevant to a stage of the possible medical condition. The second time may be preconfigured based on an amount of time elapsed since the first time.

[0395] At block **2604**, the processing device may provide, at a third time, third information of the set of information to the computing device of the patient for presentation on the computing device of the patient. The third information may be associated with a third node of the set of nodes, and the third time may be after the second time. The third information may be different than the first

information and the second information. The third information may pertain to how the possible medical condition affects people, signs and symptoms of the possible medical condition, a way to treat the possible medical condition, a progression of the possible medical condition, complications of the possible medical condition, or some combination thereof. The third time may be selected based on when the third information is relevant to a stage of the possible medical condition. The third time may be preconfigured based on an amount of time elapsed since the second time.

[0396] This process may continue until each node of the knowledge graph **500** are traversed to provide relevant information to the patient at relevant times until all information associated with the set of nodes has been delivered to the computing device of the patient. In this way, the patient may not be overwhelmed with a massive amount of information at once. Further, memory resources of the computing device of the patient may be saved by regulating the amount of information that is provided.

[0397] FIG. **27** shows a method **2700** for providing second information pertaining to a second possible medical condition of the patient, in accordance with various embodiments. In some embodiments, the method **2700** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **2700** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.

[0398] At block **2702**, the processing device may identify a second possible medical condition of the patient by identifying a second similarity between the indicia and a second knowledge graph representing second knowledge pertaining to the second possible medical condition. In some embodiments, the second similarity may pertain to a match between the indicia and a health artifact (known health related information) included in the second knowledge graph. For example, "vomiting" may be extracted as indicia from the sentence "patient has symptom of vomiting", and "vomiting" is a health artifact at a node in the second knowledge graph representing the flu. In some embodiments, the second similarity may pertain to a second structural similarity between a second logical structure (e.g., "Flu has symptom of vomiting) and the possible health information (e.g., "has symptom of vomiting") that is included in the unstructured data. In some embodiments a combination of the similarities between the indicia and the health artifact and between the logical structure and the possible health information may be used to identify the second possible medical condition and/or cognify the unstructured data.

[0399] At block **2704**, the processing device may provide, at the first time, second information of the second set of information to the computing device of the patient for presentation on the computing device, the second information being associated with a second root node of the second set of nodes. The second information may be provided with the first information at the first time. In some embodiments, a user interface on the computing device of the patient may present the first information and the second information concurrently on the same screen. For example, the user interface may present that the possible medical conditions include "Type 2 Diabetes Mellitus" and the "flu". It should be understood that any suitable number of possible medical conditions may be identified using the cognification techniques and the information related to those medical conditions may be provided to the computing device of the patient on a regulated basis.

[0400] In some embodiments, the patient may be presented with options to indicate whether the information provided at the various times was helpful. The feedback may be provided to the artificial intelligence engine **109** to update one or more machine learning models to improve the information that is provided to the patients.

[0401] FIG. **28** shows an example of providing first information of a knowledge graph **500** representing a possible medical condition, in accordance with various embodiments. In the depicted example, just a portion of the knowledge graph **500** representing Type 2 Diabetes Mellitus is

depicted. Based on the patient notes entered by the physician and/or the text input by the patient, the artificial intelligence engine **109** may extract indicia. Using the indicia, the artificial intelligence engine **109** may identify a possible medical condition of the patient by identifying at least one similarity between the indicia and the knowledge graph **500**. It should be understood that the artificial intelligence engine **109** identified Type 2 Diabetes Mellitus as the possible medical condition based on the similarity between the indicia and the knowledge graph **500** using the cognification techniques described herein. Accordingly, at a first time, the cognitive intelligence platform **102** may provide first information associated with the root node of the knowledge graph **500**. The root node may be associated with the name "Type 2 Diabetes Mellitus" of the medical condition. A user interface **2800** of the computing device of the patient may present the first information "Possible medical condition: Type 2 Diabetes Mellitus" at the first time.

[0402] FIG. **29** shows an example of providing second information of the knowledge graph **500** representing the possible medical condition, in accordance with various embodiments. The second information may be provided at a second time subsequent to the first time the first information was provided. The second information may be associated with at least a second node representing a health artifact of the knowledge graph **500**. The second information may be different than the first information. The second information may combine a predicate of a node that connects the second node representing the health artifact to the root node. For example, the second information may include "Type 2 Diabetes Mellitus has possible complication of prediabetes, or obesity and overweight." The second information may be presented on the user interface **2800** with the first information, as depicted. In some embodiments, just the second information may be presented on the user interface **2800** and the first information may be deleted from the user interface **2800**.

[0403] FIG. **30** shows an example of providing third information of the knowledge graph representing the possible medical condition, in accordance with various embodiments. The third information may be provided at a third time subsequent to the second time the second information was provided. The third information may be associated with at least a third node representing a health artifact of the knowledge graph **500**. The third information may be different than the first information and the second information. The third information may combine a predicate of a node that connects the third node representing the health artifact to the root node. For example, the third information may include "Type 2 Diabetes Mellitus has complication of stroke, coronary artery disease, diabetes foot problems, diabetic neuropathy, and/or diabetic retinopathy." The third information may be presented on the user interface **2800** with the first information and/or the second information, as depicted. In some embodiments, just the third information may be presented on the user interface **2800**, and the first information and the second information may be deleted from the user interface **2800**. In some embodiments, any combination of the first, second, and third information may be presented on the user interface **2800**.

[0404] In some embodiments, the various health artifacts represented by each node in the knowledge graph **500** may be provided to the computing device of the patient until all of the information in the knowledge graph **500** is provided. Additionally, if the knowledge graph **500** contains a link to another knowledge graph representing a related medical condition, the information included in that other knowledge graph may be provided to the patient. At any time, the patient may request to stop receiving information about the possible medical condition and no additional information will be provided. If the patient desires additional information faster, the patient may be presented with an option to obtain the next set of information at any time.

[0405] FIG. **31** shows a method **3100** for using cognified data to diagnose a patient, in accordance with various embodiments. In some embodiments, the method **3100** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **3100** may include operations that are implemented in computer instructions stored in a memory and executed

by a processor of a computing device.
[0406] At block **3102**, the processing device of a server may receive an electronic medical record including notes pertaining to a patient. The notes may include strings of characters arranged in sentences and/or paragraphs. The processing device may process the strings of characters and identify, in the notes, indicia including a phrase, a predicate, a subject, an object, a cardinal, a number, a concept, or some combination thereof. In some embodiments, the notes may be processed to obtain the indicia by inputting the notes into the artificial intelligence engine **109** trained to identify the indicia in text based on commonly used indicia pertaining to the medical condition.
[0407] At block **3104**, the processing device may generate cognified data using the notes. The cognified data may include a health summary of a medical condition. Generating the cognified data may further include detecting the medical condition by identifying a similarity between the indicia and a knowledge graph. For example, in some embodiments, the similarity may pertain to a match between the indicia and a health artifact (known health related information) included in the knowledge graph **500**. For example, “high blood pressure” may be extracted as indicia from the sentence “Patient X has high blood pressure”, and “high blood pressure” is a health artifact at a node in the knowledge graph **500** representing Type 2 Diabetes Mellitus. In some embodiments, the similarity may pertain to a structural similarity between the logical structure (e.g., “Type 2 Diabetes has symptoms of High Blood Pressure) and possible health related information generated using the identified indicia or subjects, predicates, and/or objects (e.g., “Patient X has symptoms of High Blood Pressure”) that is included in the unstructured data. In some embodiments, a combination of similarities between the indicia and the health artifact, and between the logical structure and the indicia/possible health related information may be used to detect the medical condition.
[0408] At block **3106**, the processing device may generate, based on the cognified data, a diagnosis of the medical condition of the patient. The diagnosis may at least identify a type of the medical condition that is detected using the cognified data. The diagnosis may be generated if a threshold number of matches between the indicia and health artifacts in the knowledge graph are identified, and/or if a threshold number of structural similarities are identified between logical structures of the knowledge graph and indicia/possible health information generated for the unstructured data. For example, the threshold numbers may be configurable and set based on a confidence level that the health artifacts that match the indicia and/or the logical structures that are similar to the indicia/possible health related information are correlated with the particular medical condition. The threshold numbers may be based on information from trusted sources, such as physicians having medical licenses.
[0409] In some embodiments, the processing device may use an artificial intelligence engine **109** that is trained using feedback from medical personnel. The feedback may pertain to whether output regarding diagnoses from the artificial intelligence engine **109** are accurate for input including notes of patients. The cognified data may include a conclusion that is identified based on a logical structure in the knowledge graph **500**, where the logical structure represents codified evidence-based guidelines pertaining to the medical condition.
[0410] At block **3108**, the processing device may provide the diagnosis to a computing device of a patient and/or a physician for presentation on the computing device. The diagnosis may be included in the cognified data. The physician may review the diagnosis and may provide feedback via graphical element(s) whether the diagnosis is accurate. The feedback may be received by the artificial intelligence engine **109** and used to update the one or more machine learning models used by the artificial intelligence engine **109** to cognify data and generate diagnoses.
[0411] FIG. **32** shows a method **3200** for determining a severity of a medical condition based on a stage and a type of the medical condition, in accordance with various embodiments. In some embodiments, the method **3200** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as

shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **3200** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device.

[0412] At block **3202**, the processing device may determine a stage of the medical condition diagnosed based on the cognified data. The stage of the medical condition may be determined based on information included in the cognified data. For example, the information in the cognified data may be indicative of the particular stage of the medical condition. Such stages may include numerical values (e.g., 1, 2, 3, 4, etc.), descriptive terms (e.g., chronic, acute, etc.), or any suitable representation capable of indicating different progressions in a range (e.g., from low to high, or from mild to severe, etc.).

[0413] The artificial intelligence engine **109** may be trained to identify the stage based on the information in the cognified data. For example, if certain symptoms are present, certain blood levels are present, certain vital signs are present, or the like for a particular medical condition, the artificial intelligence engine **109** may determine that the medical condition has reached a certain stage. The artificial intelligence engine **109** may be trained on evidence-based guidelines that correlate the various information with the particular stages. For example, it may be known that a particular stage of cancer involves symptoms such as weight loss, lack of appetite, bone pain, dry cough or shortness of breath, or some combination thereof. If those symptoms are identified for the medical condition diagnosed (cancer) for the patient, then that particular stage may be determined.

[0414] At block **3204**, the processing device may include the stage of the medical condition in the diagnosis. For example, the processing device may indicate the diagnosis is the "Patient X has stage **4** breast cancer". At block **3206**, the processing device may determine a severity of the medical condition based on the stage and the type of the medical condition. If the stage is relatively low and the medical condition is easily treatable, then the severity may be low. If the stage is relatively high (chronic) and the medical condition is difficult to treat (cancer), then the severity may be high.

[0415] At block **3208**, in response to the severity satisfying a threshold condition, the processing device may provide a recommendation to seek immediate medical attention to a computing device of the patient. The threshold condition may be configurable. In some embodiments, the threshold condition may be set based on information from a trusted source (e.g., evidence-based guidelines, clinical trial results, physician research, and the like).

[0416] FIG. **33** shows an example of providing a home user interface **3300** for an autonomous multipurpose application, in accordance with various embodiments. It should be noted that the user interfaces of the autonomous multipurpose application presented on the user device **104** of a patient may be referred to as a patient viewer herein. The home user interface **3300** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user can manage their healthcare using the home user interface **3300**. There are various options for "Health Record", "Medical Resources", "Messages", "Appointments", and "Billing and Insurance". The health record section may include information pertaining to the health of the user, such as conditions the user has, vital signs, weight, height, medications, and so forth. The medical resources section may include curated content that is tailored based on the conditions the user has and allows the user to search for any desired content using natural language processing. The messages section may enable a user to send messages to anyone on their care team, such as doctors, nurses, clinician, administrators, and so forth. The appointments section may enable a user to schedule an appointment with a person having a specialty, among other things.

[0417] A summary of the health record is presented and includes "Appointments this year", "Current medications", "Chronic conditions", and "Acute issues". Further, the home user interface **3300** includes a "Care Team" section that presents the care providers from whom the user receives

services. As depicted, “James Johnson, MD-Family Practice” is on the care team for user John Doe.
[0418] FIG. **34** shows an example of providing a user interface **3400** for selecting which person to schedule an appointment for, in accordance with various embodiments. The user interface **3400** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user interface **3400** may be presented when the user selects the “Appointments” button on the home user interface **3300**. Such a user interface **3400** may also be presented on a computing device of the service provider **112** and/or the facility **114**. For example, an administrator of a doctor's office may use the user interface **3400** on a computing device.
[0419] The user interface **3400** presents an option to select which individual for which to schedule an appointment. The options include, for example, “Yourself”, “Your Spouse”, “Your Child”, “Your Parent”, and “A Senior”. Accordingly, using the user interface **3400**, the user may schedule an appointment for multiple-family members. In some embodiments, the user interface **3400** may include an option to select a radius to search for appointments. The user entered “5 miles from my house address”. The house address of the user may be stored in a profile maintained by the cognitive intelligence platform **102**. In some embodiments, the user may enter an address and a radius to search around that address. Further, as depicted, the user interface **3400** may include an option to provide notes for appointments. The user entered “I am afraid of shots”. These notes may be presented to the care provider and/or an administrator at the office of the care provider prior to or during the appointment. Further, the notes may be maintained and presented during subsequent appointments, as well.
[0420] FIG. **35** shows an example of providing a user interface **3500** for selecting a specialty for an appointment, in accordance with various embodiments. The user interface **3500** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user interface **3500** presents numerous specialties from which the user may select. For example, the specialties include “Medical”, “Dental”, “Vision”, “Behavioral”, “Hearing”, “Vaccination”, “Lab Work”, “Health Classes”, “Health Questions”, “MedicalCare”, and “Physical Therapy”. Any suitable specialty may be included in the user interface **3500**, such that the user interface **3500** is not limited to a particular type of specialty.
[0421] FIG. **36** shows an example of providing a user interface **3600** for displaying locations of people and recommended appointment times with the people, in accordance with various embodiments. The user interface **3600** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user interface **3600** may be presented based on the selection of the specialty or specialties.
[0422] The cognitive intelligence platform **102** may be communicatively coupled with systems (e.g., clinical **3602**, patient management system, EMR system, scheduling system, etc.) of the service provider **112** having the specialties. In some embodiments, the schedule of the user may be considered when searching for available appointments. The schedules of care providers within the radius specified and matching the specialty or specialties selected may be retrieved from the systems by the cognitive intelligence platform **102**. For example, different service providers **112** having available appointments and different specialties may be presented.
[0423] As depicted, three appointments are found and recommended. Also, a map **3604** may present the locations **3606**, **3608**, and **3610** of the offices at which the service providers **112** work. The user interface **3600** presents “Schedule appointment with Dr. Johnson at 1:00 PM on Nov. 11, 2020 (0.5 miles away)”, “Schedule appointment with Dr. Jones at 2:00 PM on Dec. 11, 2020 (0.7 miles away)”, and “Schedule appointment with Dr. Thomas at 1:00 PM on Jan. 11, 2021 (1.0 miles away)”. Thus, multiple service providers **112** at different locations may be recommended for scheduling an appointment. The order of appointments may be configured to depend on distance

away from the user device **104** or address, the date and time the appointments are available, a service cost based on the insurance of the user, and so forth. In some embodiments, the specialties of the service providers **112** with recommended appointments may vary based on which specialties the user selected. For example, Dr. Johnson may be a medical doctor, and Dr. Jones may be a dentist.

[0424] FIG. **37** shows an example of providing a user interface **3700** for presenting a profile of a person, in accordance with various embodiments. The user interface **3700** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user interface **3700** may be presented when the user selects to view more details of one of the people associated with the recommended appointments.

[0425] For example, the information in the profile of "Jame Johnson, MD" includes the type of practice "Family Practice" and a brief description of Dr. Johnson. The profile also includes his education, services he performs, and languages he speaks. The profile may include other information, as well, and the presented information is for illustration purposes and is not to limit the disclosure. In some embodiments, the profile may include the types of insurance accepted by Dr. Johnson and/or the clinic/hospital at which he works.

[0426] FIG. **38** shows an example of providing a user interface **3800** that shows various payment options for the selected appointment, in accordance with various embodiments. The user interface **3800** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user interface **3800** may be presented when the user selects one of the recommended appointments presented in the user interface **3600** of FIG. **36**.

[0427] The user interface **3800** may present information indicating that "You selected the appointment with Dr. Johnson at 1:00 PM on Nov. 11, 2020 (0.5 miles away)". The cognitive intelligence platform **102** may retrieve the insurance plan for the user of the user device **104** that selected the appointment. The cognitive intelligence platform **102** may determine the deductible and/or co-pay for the insurance plan, and determine an expected payment that the user will be expected to pay based on the deductible and/or co-pay. The autonomous multipurpose application may perform one or more function calls to an application programming interface of a system associated with the insurance provider to determine what the user is expected to pay, an amount the insurance provider may cover, a deductible amount, a co-pay, and the like. For example, if the deductible for the insurance plan is $6,000, the user has paid $3,000 toward the deductible, and the service to be performed by Dr. Johnson costs $210, then the user may be expected to pay the $210 out of pocket that will apply towards the deductible because the deductible has not been met yet. In some instances, the entity (e.g., clinic, hospital, office, etc.) at which the service provider performs the service may offer a self-pay cost for particular services. In the depicted example, a self-pay costs of $40 is presented for Dr. Johnson to perform the service.

[0428] In the depicted example, electronic scheduling is not enabled, and thus, the user was allowed to select which appointment they wanted to schedule, and the user interface **3800** is presented that allows the user to select how to pay for the service to be provided at the scheduled appointment. Accordingly, the autonomous multipurpose application provides cost transparency and the ability to choose different options for paying for the service via the user interface **3800**.

[0429] FIG. **39** shows an example of providing a user interface **3900** that shows messages pertaining to appointments for a user, in accordance with various embodiments. The user interface **3900** is presented on a display of the user device **104**. The user device **104** is communicatively coupled with the cognitive intelligence platform **102** that may execute the autonomous multipurpose application. The user interface **3900** may be presented when the user selects the Messages tab on the home user interface **3300** of FIG. **33**.

[0430] As depicted, an inbox of the user presents 4 messages. A first message **3902** indicates that

the appointment was confirmed with Dr. Johnson on Nov. 11, 2020 at 1:00:00 PM. This confirmation message **3902** may be received in response to the user selecting the particular appointment and the user device transmitting a message to the cognitive intelligence platform **102**. The cognitive intelligence platform **102** may communicate via APIs with a system (e.g., EMR) associated with Dr. Johnson to send the appointment request to the system. If the appointment is still available, the system may book the appointment as a booked appointment and transmit the message **3902** back to the cognitive intelligence platform **102** and/or the user device **104**.

[0431] The messages may use cryptography and be presented by the user interface **3900** after decryption. In some embodiments, public key-private key encryption may be used to encrypt and decrypt the messages. In some embodiments, the messages may be transmitted via text messaging, emails, and/or voicemail. Thus, omni-channel messaging may be implemented by the cognitive intelligence platform **102**.

[0432] FIG. **40**A shows an example of a cognitive intelligence platform **102** receiving an image **4000** of an insurance card **4002**, in accordance with various embodiments. The image **4000** may be captured by a camera of the user device **104**. The image **4000** may be a file that is emailed to an email account of the user and accessed on the user device **104**. The image **4000** may be obtained in any suitable manner. The image **4000** may be transmitted to the cognitive intelligence platform **102**.

[0433] The cognitive intelligence platform **102** may perform imaging extraction techniques, such as optical character recognition and/or use a machine learning model trained to identify and extract certain information. The cognitive intelligence platform **102** may use the critical thinking engine **108** that executes artificial intelligence techniques pertaining to natural language processing. For example, optical character recognition may refer to electronic conversion of an image of printed text (e.g., a driver's license, an insurance plan, a certification, etc.) into machine-encoded text. OCR may be used to digitize information include on various cards, documents, and the like. In some embodiments, pattern recognition and/or computer vision may be used to extract information form the cards, documents, and the like. Computer vision may involve image understanding by processing symbolic information from image data using models constructed with the aid of geometry, physics, statistics, and/or learning theory. Pattern recognition may refer to electronic discovery of regularities in data through the use of computer algorithms and with the use of these regularities to take actions such as classifying the data into different categories and/or determining what the symbols represent in the image (e.g., words, sentences, names, numbers, identifiers, etc.).

[0434] Further, natural language understanding (NLU) may be performed on the image of the cards, documents, or the like. The NLU techniques may process unstructured data using text analytics to extract entities, relationships, keywords, semantic roles, and so forth. The NLU may extract the text from the images received by the cognitive intelligence platform **102**.

[0435] For example, FIG. **40**B shows an example of the cognitive intelligence platform **102** extracting insurance plan information and causing it to be presented on a user device **104**, in accordance with various embodiments. The insurance plan information presented on the user device **104** includes “Your insurance plan is: Bluecross Blueshield (BCBS)®”, “Your dependents are: Spouse, Child”, “Your insurance expires on: Jan. 1, 2021”, “Your deductible is: $6000”, and “You have paid $3000 of the $6000 deductible.”

[0436] FIG. **40**C shows an example of the cognitive intelligence platform **102** extracting driver's license information and causing it to be presented on the user device **104**, in accordance with various embodiments. User interface **4010** is presented on the user device **104**. As depicted, the information extracted from an image **4012** of the driver's license includes First Name (“Regina b”), Last Name (“ranoa”), Sex (“Female”), Date of Birth (“Jun. 21, 1961”), Address (“655 12 S **224**, Oakland CA 94607”), Issue Date (“Sep. 30, 2011”), Expiration Date (“Oct. 31, 2016”), and ID number (“B82364178”). Also, an image **4014** of a face of the person on the image **4012** of the driver's license may be extracted and used for a profile picture of the user. Other information that

may be extracted may include the Eye Color, Height, Weight, and so forth. The information extracted from the image **4012** may be associated with the user and stored in the cognitive intelligence platform **102**.

[0437] FIG. **40**D shows another example of the cognitive intelligence platform **102** extracting insurance plan information and causing it to be presented on the user device **104**, in accordance with various embodiments. User interface **4020** is presented on the user device **104**. As depicted, the information extract from an image **4022** of the insurance card may include various columns for "Accuracy", "Name", "Type", and "Value". The Accuracy column refers to whether the information extracted is accurate. For example, a service (application programming interface) associated with the insurance provider (HMSA) may be called and provided with the information extracted from the image **4022**.

[0438] The service may determine whether the information is accurate for the insurance plan of the user and return a response indicated "Y" or "N". The Name column refers to the name of the data. The Type column refers to the data type of the information. The Value column refers to the value of the data extracted from the image **4022**.

[0439] In the depicted example, the following information may be extracted and presented in the user interface **4020**: Company Name ("HMSA"), Subscriber Name ("KIMO M ALOHA"), Subscriber ID ("LLA000012334456"), PLAN ("80840"), RXBIN ("004336"), RXPCN ("MEDDADV"), RXGRP ("RX3982"), RXID ("A000012334456"), MEDICAL ("706"), PART D ("737"), Group ("M12421"), Primary ("DR MOKI HANA"). The cognitive intelligence platform **102** validated that each value of data is accurate and presents "Y" in the Accuracy column for each row of data. The information extracted from the image **4022** may be associated with the user and stored in the cognitive intelligence platform **102**.

[0440] FIG. **41** shows an example of providing a user interface **4100** that shows an appointment has been electronically scheduled, in accordance with various embodiments. The user device **104** presents the user interface **4100** of the autonomous multipurpose application. The user may have elected to enable electronic scheduling via an option presented on the user device **104**. The autonomous multipurpose application may be capable of allowing the user to enable or disable the electronic scheduling at any time.

[0441] In the depicted example, the user elected to enable electronic scheduling. Accordingly, when the user requests to schedule an appointment for a selected user (e.g., their self, a dependent, etc.) and a specialty of a person to perform a service at the appointment, the cognitive intelligence platform **102** may obtain the schedules of people having the specialty within a geolocation radius of the user. For example, the cognitive intelligence platform **102** may retrieve the schedules from systems (e.g., EMRs) of the service provider **112** and/or a clinical system **3602**. The cognitive intelligence platform **102** (e.g., autonomous multipurpose application) may analyze multiple factors when selecting which appointment to schedule. The multiple factors may include availability of the people having the specialty, availability of the user, ratings of the people having the specialty, proximity to the user of the people having the specialty, insurance considerations, and the like. For example, the cognitive intelligence platform **102** may determine an expected payment amount the selected user will be expected to pay for the service to be performed based on a deductible and/or co-pay specified in the insurance plan of the selected user. The cognitive intelligence platform **102** may also determine a self-pay cost that the selected user will be expected to pay without using insurance.

[0442] The cognitive intelligence platform **102** may select the appointment with Dr. Johnson based on the factors described above. Accordingly, the user interface **4100** presents "An appointment has been electronically scheduled and confirmed with Dr. Johnson at 1:00 PM on Nov. 11, 2020 (0.5 miles away). Further, the cognitive intelligence platform **102** may select the option for the self-pay cost for the appointment without using insurance because the self-pay cost is cheaper than the expected payment amount using insurance. Accordingly, the user interface **4100** presents "The

appointment will include self-pay cost of $40 because the deductible has not been met and using insurance would cost $210." Further, the user interface **4100** may present options to allow the user to "Change payment method", "Change appointment", "Change insurance", "View profile of Dr. Johnson", and "Provide notes for appointment". Other options may include "Schedule another appointment".

[0443] FIG. **42** shows an example of providing a user interface **4200** that shows a user needs financial aid for a particular service, in accordance with various embodiments. The user interface **4200** may be presented on a device of the service provider **112**. The service provider **112** may be the physician, administrator, or the like. The cognitive intelligence platform **102** may determine, based on the insurance plan of the user, that the user may need financial aid to pay for the service. For example, if the insurance is a high deductible and the service cost is expensive, then the cognitive intelligence platform **102** may determine the user may want financial aid. The user interface **4200** presents "User X needs financial aid to pay for the service. Their deductible has $3000 left and the service will cost $210 using insurance." In such a scenario, the service provider **112** may discuss financial aid with the user prior to the user coming in for the appointment, during the appointment, and/or after the appointment.

[0444] FIGS. **43-45** show methods **4300**, **4400**, and **4500** for scheduling an appointment between a person having a specialty and a user, FIGS. **52-54** show methods **5200**, **5300**, and **5400** for checking-in a user for a scheduled appointment. In some embodiments, various of the operations in the methods **4300**, **4400**, **4500**, **5200**, **5300**, and/or **5400** may be performed in combination.

[0445] FIG. **43** shows a method for scheduling an appointment based on whether a user has elected to enable electronic scheduling, in accordance with various embodiments. In some embodiments, the method **4300** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **4300** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **4300** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.

[0446] At block **4302**, the processing device may obtain a set of schedules for people having a specialty. The processing device may obtain the set of schedules for the set of people having the specialty from at least an electronic medical record system, a patient management system, a scheduling management system, or the like. In some embodiments, the set of schedules may be obtained for people within a geographic radius of a location of the user (e.g., home address of the user) or computing device of the user. The specialty may be selected by the user of the user device **104**. For example, the user may desire to go to a dentist for a teeth cleaning or problem they are experiencing with a tooth, the user may desire to go to a medical doctor for certain symptoms they are experiencing, and so forth. To that end, a set of specialties to be selected from may include at least two of a dentist, a medical doctor, an optometrist, a behavioral psychologist, a chiropractor, a physician's assistant, and a masseuse.

[0447] At block **4304**, the processing device may determine whether a user has elected to enable electronic scheduling. A user interface of the autonomous multipurpose application may be presented on the user device **104** and may present an option to enable or disable electronic scheduling of appointments.

[0448] At block **4306**, responsive to determining the user has elected to enable electronic scheduling, the processing device may determine (block **4308**) which person of the set of people has an available appointment based on the set of schedules, transmit (block **4310**) a request to book the available appointment for the person to provide a service to the user, receive (block **4312**) a response indicating the available appointment is booked as a booked appointment between the

person and the user, and provide (block **4314**) a notification pertaining to the booked appointment.
[0449] At block **4316**, responsive to determining the user has not elected to enable electronic scheduling, the processing device may determine (block **4318**) which person of the set of people has an available appointment based on the set of schedules, and provide (block **4320**) a notification pertaining to the person having the available appointment to a computing device of the user, where the notification includes a recommended date and time for the available appointment. For example, multiple recommended available appointments may be provided for presentation on a user interface on the user device **104**. The recommended available appointments and the locations of the service providers **112** associated with the recommended available appointments may be presented in text form (e.g., a list) on the user interface and/or in a map. The recommended available appointments may each provide a date and time of the appointment, an identity of the service provider **112** to perform the service, a distance from the user or the user device **104**, or some combination thereof. The distance from the user device **104** may be determined using global positioning system (GPS) coordinates of the user device **104** and the location of the service provider **112**.
[0450] In some embodiments, determining which person of the set of people has the available appointment may be based on the available appointment having a future date and time that is closest to a current date and time the request was received. Further, the determination of which person of the set of people has the available appointment may be based on a schedule of the user, insurance considerations (e.g., whether a deductible has been met, and/or a co-pay cost) for the service, and the like.
[0451] In some embodiments, the notification pertaining to the booked appointment may be provided to the user device **104**, a computing device of the service provider **112**, a computing device of an administrator of the service provider **112**, and/or a computing device of a facility **114**. The notification may be a secure message displayed by a user interface of the autonomous multipurpose application, a secure text message, a secure email, and/or a secure voicemail/telephone call.
[0452] FIG. **44** shows a method **4400** for selecting a payment option between a co-pay cost and a self-pay cost, in accordance with various embodiments. In some embodiments, the method **4400** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **4400** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **4400** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.
[0453] At block **4402**, the processing device may obtain an image of an insurance card of the user. The image may be captured using a camera of the user device **104** and may be transmitted to the processing device of the cognitive intelligence engine **102** from the user device **104**.
[0454] At block **4404**, the processing device may process the image to extract information pertaining to an insurance plan of the user. The processing device may use various artificial intelligence techniques to extract the information, such as optical character recognition, pattern recognition, or the like. One or more machine learning models may be trained to identify the text included at portions of the insurance card based on training data that uses labels. For example, supervised training using training data including numerous images of insurance cards with labels identifying pertinent text and identifiers. The trained machine learning models may identify the pertinent text and extract the text from the image by processing pixels and/or using object character recognition.
[0455] At block **4406**, the processing device may determine, based on the insurance plan, an expected payment that the user will pay for the service in view of a deductible and/or co-pay

specified in the insurance plan. The processing device may be communicatively coupled with a system of the insurance provider. The processing device may verify the information extracted from the insurance card with the system of the insurance provider. Further, the processing device may obtain the amount of the deductible, an amount already paid towards the deductible, a co-pay, and the like. In one example, if the user has paid $3000 towards a $6000 deductible, and a service costs $210, then the user may be responsible for the $210 since the deductible is not satisfied. However, in some instances, the deductible may be satisfied and the user may be expected to pay a lower amount (e.g., co-pay of $20).

[0456] At block **4408**, the processing device may determine, without considering the insurance plan, a self-pay cost the user is expected to pay for the service. Some entities may provide flat fees for certain services performed by the service providers **112** without considering insurance. For example, a service may include a routine physical and may be a flat fee of $40.

[0457] At block **4410**, the processing device may select to pay using the insurance plan of the user when the expected payment is less than the self-pay cost. At block **4412**, the processing device may select to pay without using the insurance plan of the user when the self-pay cost is less than the expected payment. If payment information for the user is stored in a profile of the user, the selected payment option may be paid prior to the appointment, during the appointment, or after completion of the appointment via electronic communication with a system of the service provider **112** or a financial institution associated with the service provider **112**. For example, when the user checks-in for the scheduled appointment, the selected payment option may be electronically paid by the autonomous multipurpose application. In some embodiments, the user may pay when they check-in for the appointment at the location of the scheduled appointment.

[0458] FIG. **45** shows providing various costs associated with a service to a computing device of a user, in accordance with various embodiments. In some embodiments, the method **4500** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **4500** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **4500** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.

[0459] Method **4500** may be performed when the user has elected to disable electronic scheduling.

[0460] At block **4502**, the processing device may receive an appointment request for a person to provide a service to a user. The appointment request may include a specialty of the person to provide the service to the user. The appointment request may specify an address and a radius around the address from which to search for available appointments. In some embodiments, the appointment request may specify using a geolocation of the user device **104** and a radius around the geolocation from which to search for available appointments. In some embodiments, the appointment request may specify an identity of the person to provide the service to the user. The cognitive intelligence platform **102** may provide recommended available appointments with the person having the specialty.

[0461] At block **4504**, the processing device may determine, based on the insurance plan, an expected payment that the user will pay for the service in view of a deductible specified and/or co-pay in the insurance plan.

[0462] At block **4506**, the processing device may determine, without considering the insurance plan, a self-pay cost the user is expected to pay for the service. The self-pay cost may be obtained from a system associated with the facility **114**, clinic, or entity at which the service provider **112** is providing the service for the appointment. For example, an entity (e.g., company) may be a store that includes a clinic and there may be fixed self-pay costs for various services, such as vaccines,

physicals, consultations, etc.
[0463] At block **4508**, the processing device may cause the expected payment, the self-pay cost, or some combination thereof to be presented on a computing device of the user (user device **104**), a computing device of an administrator, a computing device of a person having the specialty (e.g., service provider **112**), or some combination thereof. The user may select the payment option that is preferred and a request to book the selected appointment with the selected payment option may be transmitted to a system (e.g., EMR, scheduling management system, patient management system, etc.) associated with the person having the specialty and/or the facility **114** at which the person having the specialty will perform the service for the selected appointment. If the selected appointment is confirmed, a response may be transmitted to the cognitive intelligence platform **102** and a message may be sent to the user device **104** confirming the appointment.
[0464] FIG. **46** shows an example of providing a user interface **4600** for checking-in a user for a service, in accordance with various embodiments. The user device **104** presents the user interface **4100** of the autonomous multipurpose application. As depicted, an option **4602** (e.g., input box) may be presented for the user to enter their name, and another option **4604** (e.g., button) be presented to allow the user to begin the check-in process. When the user selects the option **4604**, a check-in request may be transmitted to the cognitive intelligence platform **102**. The check-in request may include the name the user provided, or any suitable identifier for the user. The cognitive intelligence platform **102** may retrieve any check-in documents (e.g., consents, medical history, any suitable check-in document, etc.) associated with the name or identifier of the user. The cognitive intelligence platform **102** may store any check-in documents the user has completed at any service provider **112** that has a system (e.g., EMR) communicatively coupled with the cognitive intelligence platform **102**. That is, the cognitive intelligence platform **102** may function as a centralized repository for any check-in documents such that the user does not to refill the same check-in documents if they go to a different service provider **112**.
[0465] Instead, if the check-in documents required for a new service provider **112** are complete, the cognitive intelligence platform **102** may transmit those check-in documents to the system (e.g., EMR) associated with the new service provider **112**, and the user will be checked-in without having to refill out the check-in documents. If the check-in documents are not complete, the cognitive intelligence platform **102** may cause the user device **104** to present the incomplete check-in documents for the user to complete.
[0466] For example, FIG. **47** shows an example of providing a user interface **4700** that shows additional required information is needed for a check-in document, in accordance with various embodiments. The user interface **4700** may be presented on the user device **104** and/or a computing device of an administrator. In the depicted example, the user is checking-in for an appointment scheduled with service provider **112.2** (e.g., a dentist). Service provider **112.2** requires completion of check-in document “Form A.2”. The user previously went to an appointment with service provider **112.1** (e.g., medical provider), where the user completed check-in document “Form A.1”. The cognitive intelligence platform **102** received the completed check-in document “Form A.1”, associated it with the identity of the user, and stored it in a database. As depicted, the cognitive intelligence platform **102** is storing Form A.1, Insurance plan, Consent forms, and Licenses for the user.
[0467] Form A.2 includes most of the same information as Form A.1, but Form A.2 includes a new field of information that was not included in Form A.1. Accordingly, the user interface **4700** indicates “It looks like we need to get some more information from you for your medical history pertaining to our Form A.2. We were able to obtain most of your medical history information from another form you completed in the past (e.g., Form A. 1) for your medical provider.”
[0468] Accordingly, as depicted, the fields in Form A.2 for “Previous surgeries” (“Appendectomy”) and “Date of previous surgeries” (“Sep. 9, 2010”) is prefilled with the information obtained from Form A.1. The new field “Have you had a root canal?” is specific to the service provider **112.2** and

is incomplete. The user may enter yes or no in the field and submit the Form A.2 to the cognitive intelligence platform **102** to maintain for future reference.

[0469] For example, FIG. **48**A shows an example of providing a user interface **4800** that shows check-in is complete, an estimated wait time, and curated content tailored for a condition of the user, in accordance with various embodiments. The Form A.2 is now stored in the cognitive intelligence platform **102**, as depicted. The user interface **4800** of the autonomous multipurpose application may be presented on the user device **104** and/or a computing device of an administrator of the service provider **112**.

[0470] The user interface **4800** indicates "Patient X has been successfully checked-in! All forms and documents are complete. Thank you." Further, the cognitive intelligence platform may estimate the wait time using one or more machine learning models and/or artificial intelligence techniques. The estimation at the patient level may be based on the time of check-in and how many patients are waiting in various specialty queues. The estimation may also account for multiple physicians having the same specialty that are working the day of the appointment. In some instances, patients may check-in randomly, may have multiple appointments, and/or arrive late. These scenarios may be accounted for to provide the estimated wait time. In some embodiments, the wait time may be estimated based on the average wait time for a given specialty at a particular facility **114**. In some embodiments, the wait time may be estimated based on historical information for the service provider **112** with which the patient has the appointment. The historical information may include an average amount of time it takes the service provider **112** to perform the particular services for patients that are in the wait queue in front of the patient waiting. As depicted, the user interface **4800** presents "Your estimated wait time for a diabetes follow-up with Dr. Johnson is 20 minutes."

[0471] In addition, the cognitive intelligence platform **102** may use the knowledge cloud **106** to retrieve curated content associated with a condition for which the patient is seeking treatment at the appointment. For example, the user may have scheduled the appointment for the condition Diabetes. As depicted, the user interface **4800** presents content recommended for the user, such as "Diabetes: what are the lab values?", "Diabetes: treatments", "Symptoms of Diabetes", and "Causes of Diabetes". The content may be links that the user may select to read and/or view the content. The content may include articles, videos, documents, pictures, etc. that are reviewed, curated, and/or approved by licensed medical professionals. In some embodiments, the cognitive intelligence platform **102** may also retrieve curated content for any condition of the patient that the cognitive intelligence platform **102** is aware of. For example, if the patient has asthma, content pertaining to asthma may be provided. As such, the amount of information presented to a user may not overwhelm the user and may provide an enhanced experience because the content is tailored to their conditions. Further, computing resources (processing, memory) and network bandwidth may be reduced because the user may not perform searches for information pertaining to their conditions since content pertaining to their conditions is presented on the user interface **4800**. This may enable educating the user about their conditions while the user waits.

[0472] Further, in some embodiments, if the user desires to search for additional content, the user may select an option **4802** and enter a natural language search query into an input box. Natural language processing may be used as described herein to obtain content pertaining to the search query.

[0473] FIG. **48**B shows an example of providing a user interface **4810** that shows an estimated wait time for a scheduled appointment, in accordance with various embodiments. The user interface **4810** of the autonomous multipurpose application may be presented on the user device **104** and/or a computing device of an administrator of the service provider **112**. As depicted, the user may have scheduled two appointments for May 30. The first appointment is for a first person "Adrian Smith" and the second appointment is for a second person "Zahra Smith". The user interface **4810** indicates the wait time for a first appointment is 20 minutes. The user interface **4810** also presents a self-pay estimate of $45 for each medical appointment with the same medical doctor. Further, an

estimated total ($90.00) for the scheduled appointments is presented. Options **4812** and **4812** may also be presented. Option **4812** may allow the user to add another appointment for their self or any dependent. Option **4814** may allow the user to check-in for the appointments for each user. Further, the user may cancel and/or reschedule any appointments presented on user interface **4810**.

[0474] Accordingly, the user interface **4810** enables a user to manage multiple appointments for multiple different users in a single user interface **4810**. Thus, the user does not have to log into different systems or user interfaces to view their scheduled appointments for different users. As a result, computing resources may be saved using the disclosed techniques, and the user experience may be enhanced using the user interface **4810**.

[0475] FIG. **49** shows an example of providing a user interface **4900** that allows searching for content and provides recommended content based on a condition of the user, in accordance with various embodiments. The user interface **4900** of the autonomous multipurpose application may be presented on the user device **104**. The user interface **4900** may be accessed by the user selecting the "Medical Resources" tab on the home user interface **3300** in FIG. **33**. The cognitive intelligence platform **102** may store information pertaining to the user that indicates the user has a certain condition (e.g., "Ischemic Stroke"). Accordingly, the cognitive intelligence platform **102** may cause curated content ("Learning About an Ischemic Stroke" and "Transient Ischemic Attach: Care Instructions") to be presented on the user interface **4900** using artificial intelligence. Also, input box **4902** may enable a user to search for conditions, medications, symptoms, and so forth. The cognitive intelligence platform **102** may process the natural language as described herein to provide the content associated with the entered search query.

[0476] In addition, graphical elements (e.g., buttons) may be presented for the user to browse medical information. The medical information to be browsed may include conditions, symptoms, medications, procedures, labs, and so forth. When a graphical element is selected, content associated with the medical information may be retrieved from the knowledge cloud **106** and presented on the user interface **4900**.

[0477] FIG. **50** shows an example of providing a user interface **5000** to check symptoms, in accordance with various embodiments. The user interface **5000** of the autonomous multipurpose application may be presented on the user device **104**. The user interface **5000** may include a graphical representation **5002** of a human body (e.g., male and/or female). The graphical representation **5002** may include different portions that are selectable by clicking on the portions (using a mouse and/or a finger on a touchscreen) or mousing-over the portions to highlight the portions. As depicted, the user selected a portion corresponding to eyes. A pop-up menu **5004** may appear that includes a list of symptoms to select from. As depicted, the symptoms in the pop-up menu **5004** include "Burns to the Eye", "Eye Injuries", "Eye Problems, Noninjury", "Fishhook Injuries", "Objects in the Eye", "Pinkeye". The user may select "Burns to the Eye".

[0478] Accordingly, FIG. **51** shows an example of providing a user interface **5100** that provides details about symptoms that have been authored and reviewed by medical doctors, in accordance with various embodiments. The user interface **5100** of the autonomous multipurpose application may be presented on the user device **104**. The user interface **5100** may present content retrieved from the knowledge cloud **106** pertaining to the symptoms "Burns to the Eye". As depicted, the user interface **5100** includes a section **5102** that presents information pertaining to the content, such as the content is "Current as of Sep. 23, 2018", "Author: Healthpoint Staff", "Medical Review: William H. Blahd Jr. MD, FACEP-Emergency Medicine, Kathleen Romito MD-Family Medicine, Adam Husney MD-Family Medicine". Accordingly, the user may verify that the content presented is current and has been reviewed by people having medical licenses. Such content may provide comfort to the user that the user can trust the content they are presented.

[0479] FIG. **52** shows a method **5200** of maintaining and transmitting check-in documents for a user to numerous different computing devices associated with people performing different specialties, in accordance with various embodiments. In some embodiments, the method **5200** is

implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **5200** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **5200** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.

[0480] At block **5202**, the processing device may maintain a set of check-in documents for a user. For example, the cognitive intelligence platform **102** may retrieve the check-in documents that are required to be filled out for each service provider **112** for appointments with the service providers. The check-in documents may be consent forms for distributing health information, consent forms for procedures, consent forms for minors, medical history documents, and so forth. There may be overlap between information that is requested amongst the set of check-in documents. For example, the medical history document for a first specialty of a service provider **112.1** (medical doctor) may require the user to enter their previous surgeries and the medical history document for a second specialty of a second service provider **112.2** (dentist) may also require the user to enter their previous surgeries. This information may be stored the first time the user enters the information in the medical history document at a first appointment and prefilled if the user needs to add other information to the medical history document for a subsequent appointment. Accordingly, the cognitive intelligence platform **102** may function as a central repository of check-in documents for multiple specialties and for multiple users.

[0481] At block **5204**, the processing device may receive, from the user device **104**, a set of requests to check-in the user for a set of scheduled appointments where a set of people each having a different respective specialty of a set of specialties are to provide a different respective service to the user. The set of specialties may include medical doctors, dentists, optometrists, ophthalmologists, chiropractors, masseuses, orthodontists, behavioral specialists, therapists, physical therapists, clinicians, or some combination thereof. In some embodiments, the set of requests may be received over a period of time and each of the set of scheduled appointments may be scheduled at different dates, times, or both.

[0482] At block **5206**, the processing device may determine respective subsets of the set of check-in documents that are required to be complete for each of the different respective specialty of each of the set of people. In some instances, the respective subsets of the set of check-in documents may include the same check-in documents (e.g., medical history form, consent form). In some instances, the respective subsets of the set of check-in documents may include one or more different check-in documents and/or one or more different information to be provided by the user.

[0483] In some embodiments, for each of the set of scheduled appointments, the processing device may determine whether check-in requirements are satisfied. The check-in requirements may be satisfied when required information in each of the respective subsets of the set of check-in documents has already been provided. In some embodiments, responsive to determining the check-in requirements for one of the set of scheduled appointments is satisfied, the processing device may check-in the user for the one of the scheduled appointments.

[0484] In some embodiments, responsive to determining the check-in requirements for one of the set of scheduled appointments is not satisfied because one of the respective subsets of the set of check-in documents is lacking a portion of the required information, the processing device may cause the computing device to present a notification that the portion of the required information is lacking. The processing device may receive the portion of the required information and update the one of the respective subsets of the set of check-in documents with the portion of the required information. Further, the processing device may check-in the user for the one of the set of schedule appointments once the update is complete.

[0485] At block **5208**, the processing device may transmit each of the respective subsets of the set of check-in documents to a set of computing devices each associated with each of the different respective specialty. The respective subsets of the check-in documents may be cryptographically signed. For example, public key and private key encryption may be used to cryptographically sign the respective subsets of the check-in documents.
[0486] In some embodiments, the processing device may update the set of check-in documents based on input from the user, input from the set of people having the specialties, output from a machine learning model trained to determine when certain information needs to be updated, information obtained from a third-party source (e.g., information about a child dependent entered by a parent), or some combination thereof. In some embodiments, the machine learning model may be trained to determine when the insurance plan is about to expire and cause a notification to be presented on the user device **104** indicating that the insurance plan information should be updated.
[0487] The disclosed techniques may eliminate manual or paper check in. The disclosed techniques may Maintain and satisfy all check-in requirements from a multi-specialty perspective and electronically transmitting up-to-date and sending cryptographically signed check-in documents to the doctor's office/practice management software/electronic health record software instead of paper.
[0488] FIG. **53** shows a method of determining whether the user has completed certain check-in documents required for a booked appointment, in accordance with various embodiments. In some embodiments, the method **5300** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **5300** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **5300** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.
[0489] At block **5302**, the processing device may determine which documents the user has to complete for a booked appointment or scheduled appointment. This determination may be made when the user requests to check-in for the booked appointment.
[0490] At block **5304**, the processing device may determine whether the user has completed the documents.
[0491] At block **5306**, responsive to determining the user has not completed the documents, the processing device may electronically fill in (block **5308**) fields with any information the user has already provided for the documents, and cause (block **5310**) the documents with the electronically filled in fields to be presented on a computing device of the user (user device **104**) for further completion. Responsive to determining the documents are complete, the processing device may check-in the user and provide an estimated wait time for presentation on the user device **104**. Further, the processing device may cause curated content tailored for one or more conditions of the user to be presented on the user device **104**.
[0492] FIG. **54** shows a method **5400** of providing an estimated wait time to a computing device of the user, in accordance with various embodiments. In some embodiments, the method **5400** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **5400** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **5400** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.
[0493] At block **5402**, the processing device may check-in a user for a scheduled appointment with

a person having a specialty to perform a service. The checking-in may be completed when the user has provided the information in the check-in documents for the specialty of the person to perform a service at the scheduled appointment.

[0494] At block **5404**, the processing device may determine, using a machine learning model, an estimated wait time based on an average amount of time it takes people having the specialty to perform the service for the users. In some embodiments, the estimation at the patient level may be based on the time of check-in and how many patients are waiting in various specialty queues. The estimation may also account for multiple physicians having the same specialty that are working the day of the scheduled appointment. In some instances, patients may check-in randomly, may have multiple appointments, and/or arrive late. These scenarios may be accounted for to provide the estimated wait time. In some embodiments, the wait time may be estimated based on historical information for the service provider **112** with which the patient has the appointment. The historical information may include an average amount of time it takes the person having the specialty to perform the particular services for patients that are in the wait queue in front of the patient waiting.

[0495] At block **5406**, the processing device may provide the estimated wait time to a computing device of the user for presentation on a user interface of the computing device of the user (user device **104**).

[0496] At block **5408**, the processing device may provide curated content tailored for the user based on the service, the specialty, a condition pertaining to the service, other conditions associated with the user, or some combination thereof. Accordingly, the disclosed techniques educate the user with pertinent information while the user waits in a lobby or waiting room to be called back to an office for the scheduled appointment.

[0497] At block **5410**, the processing device may maintain documents for the user and a dependent of the user and provide the documents to any requesting client device. The documents may be check-in documents described above. The cognitive intelligence platform **102** may maintain the check-in documents for each person of a family. A request client device may include a system (e.g., EMR) of a new service provider **112** that the user has not been to yet and/or a system (e.g., EMR) of a previous service provider **112** that requests updated information.

[0498] FIG. **55** shows an example of providing a user interface **5500** that includes options to select a condition, a number of areas of the condition to manage, and which areas of the condition to manage, in accordance with various embodiments. The options are depicted in section **5502**, **5504**, and **5506**, respectively. The user may have logged into, using the user device **104**, the autonomous multipurpose application with credentials associated with a patient/user role. As such, the user interface **5500** of the patient viewer may be provided by the autonomous multipurpose application and presented on the user device **104**.

[0499] As depicted, section **5502** presents text “Please select one of your conditions that you would like to manage”. The conditions that are presented in section **5502** may be conditions diagnosed for the user logged into the patient viewer (e.g., via 2-factor authentication) having the user interface **5500**. For example, the cognitive intelligence platform **102** may maintain a data structure for each patient that stores each condition diagnosed for the patient. In section **5502**, the conditions associated with the logged-in user are “Type 2 Diabetes Mellitus”, “Arthritis”, “Multiple Sclerosis”. The user selected “Type 2 Diabetes Mellitus”, which may cause a knowledge graph representing Type 2 Diabetes Mellitus to be accessed in the knowledge cloud **106**. Further, a patient graph for Type 2 Diabetes Mellitus of the user may be accessed in the knowledge cloud **106** as a result of the selection. It should be noted that more than one condition may be selected by the user to manage, and the patient viewer may present a care plan for each respective condition selected. If the user does not select one or more conditions, a default selection may be made, such as selecting all of the conditions of the user.

[0500] Different respective data structures (e.g., patient graphs) pertaining to each condition of the user may be maintained by the cognitive intelligence platform **102**. In some embodiments, the

patient graphs may include elements (e.g., health artifacts) represented by nodes that are linked based on relationships. The elements included in the patient graph may represent content consumed by, actions performed by, and/or interactions performed by the user.

[0501] A root node of a patient graph for a condition may include a type of the condition with which the user is diagnosed. If the user is recently diagnosed, the patient graph for the condition of the user may just include the root node, since the user has not performed any actions and/or interactions, or consumed content. As described further below, the disclosed techniques may compare the patient graph for a condition with a knowledge graph for that condition and generate a care plan. The care plan may include various action instructions for a patient, a medical personnel, and/or an administrator.

[0502] In section **5504**, the user interface **5500** presents an option to "Please select how many areas of the selected condition that you would like to manage". The user entered "3" into the input text box on the user interface **5500**. It should be understood that the user may choose any suitable number of areas to manage. In some embodiments, if the user does not input a number, a default number may be used.

[0503] In section **5506**, the user interface **5500** presents the various areas of the selected condition. The areas for Type 2 Diabetes Mellitus may include "Medications", "Symptoms", "Tests", "Self-care", "Complication information", etc. These areas may correspond to elements in the knowledge graph for the condition Type 2 Diabetes Mellitus. In the depicted example, the user selected "Medications", "Symptoms", and "Tests". If the user does not make a selection of the areas, then a default selection may be made, such as all of the areas of the condition. The selections of the condition(s), the number of areas of the condition, and/or the areas of the condition may be transmitted to the cognitive intelligence platform **102**.

[0504] FIG. **56** shows an example of a knowledge graph **5600**, a patient graph **5602**, and a care plan **5604**, in accordance with various embodiments. The knowledge graph **5600** may pertain to any suitable medical condition and include numerous elements (e.g., health artifacts) represented by nodes and relationships between the nodes represented by edges. For example, the knowledge graph **5600** includes a root node **5612**; a first layer of nodes **5620**, **5622**, **5624**, **5626**, and **5628**; and a second layer of nodes **5630**, and **5632**. The root node **5612** may include information pertaining to a type of the medical condition, such as "Multiple Sclerosis". The edges connecting the root node **5612** to the first layer of nodes **5620**, **5622**, **5624**,**5626**, and **5628** may represent a relationship between the root node **5612** and the first layer of nodes **5620**, **5622**, **5624**,**5626**, and **5628**. For example, the edge connecting the root node **5612** and **5620** may represent a relationship "has symptoms of" and the node **5620** may represent a health artifact "tingling and numbness". The knowledge graph **5600** may include a superset of curated medical knowledge of the medical condition represented by the nodes and relationships pertaining to the medical condition.

[0505] The patient graph **5602** may be tailored for a particular user and may correspond to the condition represented by the knowledge graph **5600**. For example, the patient graph **5602** may correspond to the medical condition "Multiple Sclerosis". In some embodiments, the nodes in the patient graph **5602** may represent the health artifacts (e.g., actions, interactions, content, concepts, facts, protocols, evidence-based guidelines, etc.) which the user has performed, interacted, experienced, reported, consumed, been treated for, been diagnosed, and/or been prescribed. For example, the node **5628** may represent a particular test for Multiple Sclerosis. The user may have performed the particular test for Multiple Sclerosis. As such, the node **5628** is included in the patient graph **5602**. The node **5628** may include a type of the particular test, a timestamp of the particular test, a result of the particular test, and the like.

[0506] Nodes **5626** and **5632** may correspond to other health artifacts which the user has performed, interacted, consumed, been treated for, been diagnosed, and/or been prescribed. As such, the nodes **5626** and **5632** are included in the patient graph **5602**.

[0507] In the depicted example, the user may not have interacted with and/or performed the health

artifacts associated with the nodes **5620**, **5622**, **5624**, and **5630** in the knowledge graph for Multiple Sclerosis. Accordingly, the nodes **5620**, **5622**, **5624**, and **5630** are not included in the patient graph **5602** for Multiple Sclerosis for the user. For example, the user may not have performed the action of performing a disease-modify therapy technique for treating Multiple Sclerosis. The health artifact for the disease-modifying therapy technique may be represented by node **5622**, and thus, node **5622** is not included in the patient graph **5602**.

[0508] The cognitive intelligence platform **102** may compare the patient graph **5602** to the knowledge graph **5600** to determine which areas of the condition Multiple Sclerosis to manage to generate the care plan **5604**. Further, the cognitive intelligence platform **102** may consider the areas the user selected to manage when generating the care plan **5604**. The patient graph **5602** may be projected onto the knowledge graph **5600**. Overlapping nodes that are included in both the patient graph **5602** and the knowledge graph **5600** may be identified (e.g., highlighted in a first color). Further, nodes that are included in the knowledge graph **5600** and not included in the patient graph **5602** may also be identified (e.g., highlighted in a second color).

[0509] In some embodiments, the nodes that are present in the knowledge graph **5600** and not present in the patient graph **5602** may be selected to include in the care plan **5604**. As depicted, nodes **5620**, **5622**, **5624**, and **5632** are present in the knowledge graph **5600** and not in the patient graph **5602**. Accordingly, the care plan **5604** may be generated to include the root node **5612** and the nodes **5620**, **5622**, **5624**, and **5632**. One or more action instructions may be generated and associated with each of the nodes **5620**, **5622**, **5624**, and **5632**.

[0510] For example, node **5620** may represent medications to take for the condition, and an action instruction may be generated to recommend the user discuss being prescribed a different medication for the condition. Other action instructions pertaining to various health artifacts may include scheduling a follow-up appointment, performing a certain test for the condition, reading certain recommended curated medical content pertaining to the condition, performing certain self-care treatments, and the like. In some embodiments, nodes may be selected to include in the care plan **5604** based on the areas of the condition the user selected to manage as well as the number of the areas of the condition the user selected to manage.

[0511] The care plan **5604** may be converted into natural language for each particular role. For example, the natural language representing the care plan **5604** may be tailored for providing action instructions to a user, the natural language representing the care plan **5604** may be tailored for providing action instructions to a medical personnel, and the natural language representing the care plan **5604** may be tailored for providing action instructions to an administrator. For example, the natural language conversion of the care plan **5604** may include an action instruction for the patient that specifies “Discuss changing medications with your physician”. In another example, the natural language conversion of the care plan **5604** may include an action instruction for the medical personnel that specifies “Discuss changing medications with the patient”. Each respective natural language conversion representing the care plan **5604** may be presented on the respective patient viewer, clinic viewer, and administrator viewer. The natural language conversion may be in text format and presented on the various viewers and/or may be in audio format and may be output by a speaker of a computing device.

[0512] FIGS. **57**A-**57**C**57**A-**57**D show examples for generating a care plan **5750** using a knowledge graph **500** and a patient graph **5700**, in accordance with various embodiments. In particular, FIG. **57**A depicts the knowledge graph **500** (first data structure) for the medical condition “Type 2 Diabetes Mellitus”. For purposes of explanation, it should be understood that the knowledge graph **500** includes a superset of health artifacts (e.g., elements represented by nodes) pertaining to Type 2 Diabetes Mellitus. The ontological medical data included in the knowledge graph **500** may be maintained by the knowledge cloud **106** and updated based on any changes and/or discoveries regarding medical knowledge of Type 2 Diabetes Mellitus.

[0513] FIG. **57**B depicts the patient graph **5700** (second data structure) for a particular user having

the condition Type 2 Diabetes Mellitus. The patient graph **5700** may also include an engagement profile as metadata that stores interactions of the patient with the various health artifacts presented in a care plan for the user. The interactions may be used to track a level of compliance with the care plan for the user. In some embodiments, the health artifacts represented by the nodes may be added to the patient graph as the patient interacts with the health artifacts. In some embodiments, the health artifacts may be added to the patient graph **5700** if the patient interacts with the health artifact to a threshold level.

[0514] As depicted, the patient graph **5700** includes a subset of the superset of health artifacts included in the knowledge graph **500**. For example, the patient graph **5700** includes a node representing a "Blood Glucose Test" health artifact that the patient performed. Various information (e.g., result, timestamp, etc.) pertaining to the blood glucose test may be associated with the node. However, the patient graph **5700** does not include a node representing the "A1c" health artifact that is included in the knowledge graph **500** because the patient has not interacted with that health artifact yet. In other words the patient has not performed the A1c test yet.

[0515] Other nodes representing health artifacts that are included in the knowledge graph **500** and not in the patient graph **5700** (e.g., due to the patient not interacting with those health artifacts yet) are a node representing "Endocrine, Nutritional and Metabolic Conditions", a node representing "possible complication of" connected to nodes representing "Prediabetes" and "Obesity and Overweight", and a node representing "prevented by" connected to a node representing "Metformin".

[0516] To generate the care plan **5750** depicted in FIG. **57**C, the cognitive intelligence platform **102** (e.g., the autonomous multipurpose application, the critical thinking engine **108**, and/or the knowledge cloud **106**) may compare the patient graph **5700** to the knowledge graph **500**. Comparing the patient graph **5700** to the knowledge graph **500** may include projecting the patient graph **5700** onto the knowledge graph **500**. In some embodiments, projecting the patient graph **5700** onto the knowledge graph **500** may include overlaying the patient graph **500** on the knowledge graph **500**, and/or plotting the patient graph **5700** in a same space as the knowledge graph **500**. Based on the comparing, the cognitive intelligence platform **102** may select a subset of the superset of health artifacts in the knowledge graph **500**. The selecting may be based on identifying nodes representing health artifacts that are included in the knowledge graph **500** and not the patient graph **5700**, and/or on areas of the condition the patient selected to manage in FIG. **55**. Continuing the example in FIG. **55**, the patient selected to manage the areas of "Medications", "Symptoms", and "Tests".

[0517] As depicted in FIG. **57**C, the care plan **5750** represents the patient graph **5700** projected onto the knowledge graph **500**. The nodes that are filled in (black circles) represent health artifacts that are included in the care plan based on the selecting described above. The nodes that are not filled in (empty circles) represent health artifacts that are not included in the care plan **5750**. The cognitive intelligence platform **102** selected the node representing "A1c" test to include in the care plan **5750** because the patient graph **5700** included a node representing the blood glucose test and did not include a node representing the A1c test that is included in the knowledge graph **500**. Further, the patient selected to manage "Tests", so including the health artifact A1c test fits that area.

[0518] The patient also selected to manage the areas of "Medications" and "Symptoms". Accordingly, the cognitive intelligence platform **102** included nodes representing health artifacts pertaining to those areas. In particular, the nodes included for the "Symptoms" area are "has symptom" connected to "High Blood Sugar" and the nodes included for the "Medicines" area are "treated by" connected to "Diabetes Medicines".

[0519] Although some nodes are included in the knowledge graph **500** and not in the patient graph **5700**, such as the "possible complication of" connected to "Prediabetes" and "Obesity and Overweight" health artifacts, they may not be included in the care plan **5750** because those nodes

are associated with areas the patient did not select to manage.
[0520] The care plan **5750** may be converted into natural language text by the cognitive intelligence platform **102** using the natural language database **122** according to the techniques disclosed herein. The cognitive intelligence platform **102** may generate action instructions pertaining to the health artifacts included in the care plan **5750**. FIG. **57**D depicts the care plan **5750** in the natural language text presented in a user interface **5700** of the patient viewer on the user device **104**. Although the depicted natural language text is tailored for the patient, in some embodiments, the natural language text may be tailored for the medical personnel or the administrator when presented in the clinic viewer or the administrator viewer respectively.
[0521] It should be noted that the natural language text of the care plan **5750** depicted is an example and is for explanatory purposes. Any suitable variation of the natural language text is envisioned in this disclosure. The natural language text in the user interface **5700** presents "Please find information and/or action instructions pertaining to the **3** areas you selected relating to Type 2 Diabetes Mellitus below:".
[0522] For the "Medications" area, the natural language text presents information about types of medications for the condition: "The types of medication available to treat Type 2 Diabetes Mellitus include: medication A, medication B, and medication C." Further, the natural language text presents an action instruction for the patient: "You are currently prescribed medication A. If it is not working as desired, discuss medication change with your physician".
[0523] Further, the cognitive intelligence platform **102** may compare the patient graphs of each condition of the patient to determine if there are conflicts, redundancy, and the like. For example, natural language text presents another action instruction based on artificial-intelligence analysis performed by the cognitive intelligence platform **102**: "We see that you are also prescribed medication D for condition Y. Medication B and medication D are not compatible and may cause issues. Be sure to discuss this with your physician."
[0524] For the "Symptoms" area, the natural language text presents information about types of symptoms for the condition: "Type 2 Diabetes Mellitus has the following symptoms: High Blood Sugar." Further, the natural language text presents an action instruction for the patient: "If you have high blood sugar, contact your physician".
[0525] For the "Tests" area, the natural language text presents information about types of tests for the condition: "The types of tests for Type 2 Diabetes Mellitus include: A1c Test and Blood Glucose Test." Further, the natural language text presents an action instruction for the patient: "You have already had an A1c Test. You can take an A1c test to get additional results, or you can retake the Blood Glucose Test".
[0526] FIG. **58** shows a method **5800** for generating a care plan using a knowledge graph and a patient graph, in accordance with various embodiments. In some embodiments, the method **5800** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **5800** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **5800** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.
[0527] At block **5802**, the processing device may select a first data structure corresponding to a first condition of a patient. The first data structure may be a knowledge graph of medical ontological data of the condition. The first data structure may include a set of health artifacts pertaining to the first condition and the set of health artifacts may be connected via relationships between the health artifacts.
[0528] At block **5804**, the processing device may compare a second data structure with the first

data structure. The second data structure may be a patient graph of the patient. The second data structure corresponds to the patient and the first condition of the patient, and the second data structure may include a subset of the set of health artifacts. If the second data structure includes the set of health artifacts of the first data structure, then a determination may be made by the processing device that the patient is managing the condition as desired.

[0529] At block **5806**, the processing device may select, based on the comparing, another subset of the set of health artifacts in the first data structure. The processing device may receive input from the computing device (user device **104**), and the input may specify an area of the condition the patient selects to manage. The area may include a type (e.g., Medications, Symptoms, Tests, etc.) of health artifacts in the set of the health artifacts. The processing device may select, based on the comparing, the another subset of the set of health artifacts in the first data structure by selecting the another subset based on the number and the type of health artifacts specified by the patient. In some embodiments, the processing device may select the another subset of the set of health artifacts based on which health artifacts are included in the first data structure and that are not included in the second data structure. The subset of the set of health artifacts may correspond with interactions already performed by the patient, and the another subset of the set of health artifacts may correspond with interactions that have not yet been performed by the patient.

[0530] At block **5808**, the processing device may generate a care plan including a third data structure that includes at least the another subset of the set of health artifacts. The third data structure may be a graph structure and include nodes representing the another subset of the set of health artifacts and relationships between the nodes.

[0531] At block **5810**, the processing device may cause the care plan to be presented on a computing device. The processing device may include, in the care plan, action an instruction pertaining to the another subset of the set of health artifacts. In some embodiments, the care plan is tailored based on the role of the user logged into the autonomous multipurpose application. For example, a care plan may be tailored for a patient/user role, for a care provider (e.g., medical personnel) role, for an administrator role, and the like. The action instruction may be directed toward the role of the person to receive the care plan. Each respective tailored plan may be presented on a respective computing device of the person having the respective role.

[0532] In some embodiments, the processing device may generate natural language representing the another subset of the set of health artifacts included in the third data structure. The processing device may cause the natural language to be presented on the computing device.

[0533] In some embodiments, the processing device may determine a value of patient compliance with the care plan based on tracked interactions of the patient and the another subset of the set of health artifacts. The tracked interactions may include activity of the patient using the computing device. The activity may include a selection using an input peripheral of the computing device, an amount of time the patient actively uses an application, an amount of time the patient spends viewing a particular user interface, a search query entered by the patient, or some combination thereof. The tracked interactions may include an indication from an external system that the patient has interacted with the health artifact of the another subset of the set of health artifacts. For example, the indication may be an EMR record from an EMR system of a care provider of the patient. The EMR record may indicate the user had a test performed by the care provider. The test (e.g., A1c) may be for a condition (e.g., Diabetes) and the health artifact in the patient graph of the user may be updated.

[0534] In some embodiments, the processing device may select a fourth data structure (e.g., a knowledge graph) corresponding to a second condition of the patient. The fourth data structure may include a second set of health artifacts pertaining to the second condition, and the first (e.g., Type 2 Diabetes Mellitus) and second condition (e.g., Multiple Sclerosis) are different. The processing device may compare a fifth data structure (e.g., a patient graph) with the fourth data structure. The fifth data structure pertains to the patient and the second condition of the patient, and the fifth data

structure may include a second subset of the second set of health artifacts. The processing device may select, based on the comparing, a third subset of the set of health artifacts in the fourth data structure. The processing device may generate the care plan including the third data structure that includes at least the another subset of the set of health artifacts and the third subset of the set of health artifacts. In this way, the care plan may include health artifacts pertaining to two different conditions of the patient. It should be understood that the care plan may be generated to include the health artifacts of any suitable number of conditions of the patient. The care plan may include action instructions pertaining to each condition represented in the care plan for the patient.

[0535] FIG. **59** shows a method **5900** for updating a patient graph based on an interaction with a health artifact by the patient, in accordance with various embodiments.

[0536] In some embodiments, the method **5900** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **5900** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **5900** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**. The operations of the method **5900** in FIG. **59** may be performed in some combination with the operations of the method **5800** in FIG. **58**.

[0537] At block **5902**, the processing device may receive information corresponding to a health artifact of the set of health artifacts in the first data structure. The information may pertain to an interaction with a user interface of the patient viewer, to an appointment for a condition, to an interaction with a browser, to any interaction on the user device **104**, to a medical test being performed, to exercise performed by the user, to familial medical history of the user, to a diet of the user, to scheduling an appointment, to consuming recommended curated content, and so forth. In some embodiments the information may be received from a source including an electronic medical records system, an application programming interface, a claims system, an electronic health virtual assistant, an application executing on the user device **104**, a data store, or some combination thereof.

[0538] At block **5904**, the processing device may determine, based on the information, that the patient has interacted with the health artifact.

[0539] At block **5906**, the processing device may generate an engagement profile for the patient using the health artifact with the information. In some embodiments, if an engagement profile is already generated, the processing device may update the engagement profile for the patient in the patient graph.

[0540] At block **5908**, the processing device may update the second data structure with the engagement profile for the patient. Updating the second data structure with the engagement profile for the patient may refer to storing metadata including the engagement profile with the second data structure and/or correlating the metadata and the second data structure.

[0541] At block **5910**, the processing device may update the second data structure (the patient graph) to include the health artifact with the information.

[0542] At block **5912**, the processing device may cause an indication to be presented on the computing device. The indication may include an updated care plan that indicates the interaction with the health artifact. For example, if the interaction with the health artifact is the patient performing a test pertaining to the condition, the updated care plan may present an indication that the test results are normal, abnormal, etc. and may include an action instruction pertaining to the test (e.g., “discuss the test results with your physician”).

[0543] FIG. **60**A-E show examples of modifying a care plan based on a detected emotion of the patient, a detected tone of the patient, a different medical outcome entered by a physician, or some

combination thereof, in accordance with various embodiments. FIG. **60**A depicts a user **6000** (e.g., patient) using the user device **104**. The cognitive intelligence platform **102** provided a care plan **6002** that was originally generated for the patient for a medical condition of the patient. The care plan **6002** may include an action instruction pertaining to the medical condition of the user **6000**, such as an instruction to read certain recommended content for the medical condition, schedule an appointment with a physician, perform a certain test for the medical condition, etc.

[0544] When the care plan **6002** is presented to the user via display of the user device **104**, the user device **104** may receive various input data from the user **6000**. For example, the user may enter text **6010** using any suitable input peripheral (e.g., mouse, keyboard, touchscreen) of the user device **104**, the user may speak words **6012** that a microphone of the user device **104** receives, and/or the user device **104** may capture an image **6014** (e.g., still-image, series of images, video) of the user's face and/or body using a camera of the user device **104**. The input data **6010**, **6012**, and/or **6014** may be transmitted by the user device **104** to the cognitive intelligence platform **102**.

[0545] The cognitive intelligence platform **102** may process the input data to detect a tone of the user **6000** and/or an emotion of the user **6000**. For example, a machine learning model may be trained on training data that identifies patterns between images **6014** of certain facial expressions/body language and certain emotions (e.g., happy, angry, sad, etc.). In that regard, facial recognition techniques may be used, such as detecting the face and/or body, scanning the face and/or body, creating targets, matching the targets, and verifying. The machine learning model may receive the image **6014** of the user **6000** as input and output the emotion of the user **6000**. Further, spoken words **6012** and/or the text **6010** may be processed by a machine learning model that is trained on training data that identifies patterns between the spoken words and/or text and certain emotions and/or tones (e.g., attitude of the user **6000** towards the subject presented on the user device **104**). The tones may include cheerful, pessimistic, optimistic, sarcastic, hostile, and the like. The machine learning model may use certain natural language processing techniques disclosed herein.

[0546] In some embodiments, the input data **6010**, **6012**, and/or **6014** may be received by the cognitive intelligence platform **102** when the care plan **6002** is presented on the user device **104**. In some embodiments, the input data **6010**, **6012**, and/or **6014** may be received by the cognitive intelligence platform **102** at any time the user is using the user device (e.g., even if the user is not logged into or using the autonomous multipurpose application of the cognitive intelligence platform **102**.

[0547] If the cognitive intelligence platform **102** receives the input data **6010**, **6012**, and/or **6014** when the care plan **6002** is presented to the user **6000** on the user device **104**, and the cognitive intelligence platform **102** detects a negative emotion (e.g., angry) and/or tone (e.g., hostile), the cognitive intelligence platform **102** may modify the care plan **6002** to generate an updated care plan **6020**. The updated care plan **6020** may include a different subset of health artifacts than the care plan **6002**. The different subset of health artifacts may be selected based on various criteria. For example, the different subset of health artifacts may be selected from a knowledge graph as long as the different subset of health artifacts includes a randomly selected health artifact that was not included in the care plan **6002**.

[0548] In some embodiments, the different set of health artifacts in the updated care plan **6020** may be selected based on the detected tone and/or emotion. For example, a machine learning model may be trained to generate updated care plans based on training data that includes care plans that have historically improved a users' tone and/or emotion. That is, the machine learning model may be trained to receive a care plan, detected emotion, and/or detected tone, and to generate an updated care plan using the care plan, detected emotion, and/or detected tone based on certain health artifacts of the medical condition that are not included in the care plan and that have historically improved the current emotion and/or tone of the user **6000**. Accordingly, the cognitive intelligence platform **102** may track the detected conditions and/or tones of the users in reaction to care plans

that are presented on the user device **104**.
[0549] In some embodiments, if the detected emotion (e.g., happy) and/or tone (e.g., cheerful) is positive, the cognitive intelligence platform **102** may modify the care plan to generate an updated care plan **6020**. The updated care plan **6020** may include a different subset of health artifacts than the care plan **6002**. The different subset of health artifacts may be selected based on various criteria. For example, the different subset of health artifacts may be selected from a knowledge graph as long as the different subset of health artifacts includes a randomly selected health artifact that was not included in the care plan **6002**.
[0550] In some embodiments, the different set of health artifacts in the updated care plan **6020** may be selected based on the detected tone and/or emotion. For example, if the detected tone and/or emotion is positive, a machine learning model may be trained to generate updated care plans that include health artifacts with which the user **6000** is likely to interact due to the positive tone and/or emotion. A machine learning model may be trained to receive a care plan, detected emotion, and/or detected tone, and to generate an updated care plan using the care plan, detected emotion, and/or detected tone based on certain health artifacts of the medical condition that are not included in the care plan and that have historically shown a likelihood of being interacted with by the user **6000** when the user **6000** exhibits the positive emotion and/or tone.
[0551] Further, the cognitive intelligence platform **102** may receive the input data **6010**, **6012**, and/or **6014** at any time the user is using the user device **104**. The cognitive intelligence platform **102** may use a machine learning model trained to output certain updated care plans **6020** based on the detected emotion and/or tone of the user **6000** based on the received input data **6010**, **6012**, and/or **6014**. For example, if the cognitive intelligence platform **102** detects the user has an angry emotional state, the cognitive intelligence platform **102** may use a machine learning model trained to include certain health artifacts in an updated care plan **6020** that historically improve the emotional state of the user **6000**.
[0552] FIG. **60**B depicts an example updated care plan **6020**.**1**. For purposes of explanation, the original care plan **6002** was the care plan **5750** depicted in FIG. **57**C. The care plan **5750** may have been presented in the patient viewer on the user device **104** and included the information pertaining to “Symptoms” area. Input data, such as the image **6014** (e.g., face image, body image), may be received by the cognitive intelligence platform **102** and processed. The cognitive intelligence platform **102** may input the image **6014** into the machine learning model trained to detect an emotion and/or tone of the user **6000** based on a facial expression and/or body language of the user **6000** in the image **6014**.
[0553] The cognitive intelligence platform **102** may determine the user **6000** experienced a negative emotion (e.g., angry) when viewing the “Symptoms” area of the care plan **5750**. Accordingly, the cognitive intelligence platform **102** may modify the care plan **5750** to generate updated care plan **6020**.**1** based on the negative emotion. For example, the cognitive intelligence platform **102** may include at least one different health artifact in the updated care plan **6020**.**1** than was included in the care plan **5750**. In some embodiments, a machine learning model may be trained to select health artifacts that historically improve a user's emotion when angry. Further, the cognitive intelligence platform **102** may remove the health artifacts determined to be associated with causing the negative emotion.
[0554] As depicted, the updated care plan **6020**.**1** includes new health artifacts represented by node “has complication” connected to nodes “Coronary Artery Disease”, “Diabetes Foot Problems”, “Diabetic Neuropathy”, and “Diabetic Retinopathy”. Further, the updated care plan **6020**.**1** removed the health artifacts represented by node “has symptom” connected to node “High Blood Sugar”. Providing the updated care plan **6020**.**1** may improve the experience of the user using the computing device **104** and may increase the likelihood the user continues to use the computing device **104**.
[0555] FIG. **60***c* depicts an example updated care plan **6020**.**2**. For purposes of explanation, the

original care plan **6002** was the care plan **5750** depicted in FIG. **57**C. The care plan **5750** may have been presented in the patient viewer on the user device **104**. A physician may desire a certain medical outcome for the condition Type 2 Diabetes Mellitus. For example, the physician may desire to enhance the treatment of the medical condition. Accordingly, the physician may select various health artifacts to include in the updated care plan **6020**.**2**. In the depicted example, the physician selected to include nodes represented as health artifacts "has self-care" connected to "Weight Management", "Diabetic Diet", "Healthy Eating", "Diabetes Foot Care", and "Being Active". Information and/or action instructions may be generated and include in a natural language conversion of the updated care plan **6020**.**2** in the patient viewer, clinic viewer, and/or administrator viewer.

[0556] The updated care plan **6020**.**1** may be converted into natural language text by the cognitive intelligence platform **102** using the natural language database **122** according to the techniques disclosed herein. The cognitive intelligence platform **102** may generate action instructions pertaining to the health artifacts included in the care plan **6020**.**1**. FIG. **60**D depicts the care plan **6020**.**1** in the natural language text presented in a user interface **6060** of the patient viewer on the user device **104**. Although the depicted natural language text is tailored for the patient, in some embodiments, the natural language text may be tailored for the medical personnel or the administrator when presented in the clinic viewer or the administrator viewer respectively.

[0557] It should be noted that the natural language text of the care plan **6020**.**1** depicted is an example and is for explanatory purposes. Any suitable variation of the natural language text is envisioned in this disclosure. The natural language text in the user interface **6060** presents "Please find information and/or action instructions relating to Type 2 Diabetes Mellitus below to Type 2 Diabetes Mellitus below:".

[0558] For the "Medications" area and the "Tests" area, the natural language text is the same as described with reference to FIG. **57**D.

[0559] As depicted, the "Symptoms" natural language text has been removed from the updated care plan **6020**.**1** and natural language text is added for health artifacts pertaining to the "Complications" and presented in the user interface **6060**. The user interface **6060** presents information about types of complications for the condition: "Type 2 Diabetes Mellitus has complications of stroke, coronary artery disease, diabetes foot problems, diabetic neuropathy, diabetic retinopathy." Further, the natural language text presents an action instruction for the patient: "Here is recommended medical content relating to those complications. Please read them.". The action instruction may include links to the various recommended medical content. Further, the natural language text presents another action instruction: "Speak to your physician about the complications".

[0560] The updated care plan **6020**.**2** may be converted into natural language text by the cognitive intelligence platform **102** using the natural language database **122** according to the techniques disclosed herein. The cognitive intelligence platform **102** may generate action instructions pertaining to the health artifacts included in the care plan **6020**.**2**. FIG. **60**E depicts the care plan **6020**.**2** in the natural language text presented in a user interface **6070** of the patient viewer on the user device **104**. Although the depicted natural language text is tailored for the patient, in some embodiments, the natural language text may be tailored for the medical personnel or the administrator when presented in the clinic viewer or the administrator viewer respectively.

[0561] It should be noted that the natural language text of the care plan **6020**.**2** depicted is an example and is for explanatory purposes. Any suitable variation of the natural language text is envisioned in this disclosure. The natural language text in the user interface **6070** presents "Please find information and/or action instructions relating to Type 2 Diabetes Mellitus below to Type 2 Diabetes Mellitus below:".

[0562] For the "Medications" area, the "Symptoms" area, and the "Tests" area, the natural language text is the same as described with reference to FIG. **57**D.

[0563] As depicted, natural language text is added for health artifacts pertaining to the "Self-Care" and presented in the user interface **6070**. As previously discussed, the health artifacts pertaining to "has self-care" were selected to be added based on the physician desiring a particular medical outcome. The user interface **6070** presents an action instruction for the patient: "Try self-care treatments for Type 2 Diabetes Mellitus including: weight management, diabetic diet, healthy eating, diabetes foot care, and being active.".

[0564] FIG. **61** shows a method **6100** for modifying a care plan based on a detected emotion of the patient, a detected tone of the patient, a different medical outcome entered by a physician, or some combination thereof, in accordance with various embodiments. In some embodiments, the method **6100** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **6100** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **6100** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**.

[0565] At block **6102**, the processing device may compare a first data structure with a second data structure. The first data structure (e.g., knowledge graph) includes a set of health artifacts pertaining to a first condition of the patient. The second data structure (e.g., patient graph) pertains to the patient and the first condition of the patient, and the second data structure includes a subset of the set of the health artifacts.

[0566] At block **6104**, responsive to the comparing, the processing device may generate the care plan including another subset of the set of health artifacts. The subset of the health artifacts may correspond with actions already performed by the patient, and the another subset of the set of the health artifacts may correspond with actions that have not yet been performed by the patient. The comparing may include projecting the second data structure onto the first data structure. The processing device may include, in the care plan, action instructions pertaining to the another subset of the set of the health artifacts. The action instructions may be directed toward a medical personnel, the patient, and/or an administrator depending on the role to which the care plan is tailored.

[0567] At block **6106**, the processing device may modify the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof. In some embodiments, the processing may modify the another subset of the set of the health artifacts in real-time or near real-time. Real-time or near real-time may refer to performing an action in 2 seconds or less.

[0568] In some embodiments, the processing device may detect the tone of the patient based on spoken words by the patient, text entered by the patient, or some combination thereof. In some embodiments, the processing device may detect the emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.

[0569] In some embodiments, the processing device may cause the care plan including the modifications to the another subset of the set of the health artifacts to be presented on a computing device. The care plan may be converted into natural language and may be tailored based on role of the person logged into the autonomous multipurpose application at the computing device. For example, the natural language may be tailored for the patient/user role, the care provider (e.g., medical personnel) role, and/or the administrator role.

[0570] In some embodiments, the processing device may modify the another set of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health artifact that corresponds to

the medical outcome and to include the health artifact in the another subset of the set of the health artifacts. For example, the physician may select to include in the care plan health artifacts pertaining to self-care treatment for Type 2 Diabetes Mellitus when the care plan originally generated is lacking those health artifacts. The physician may be attempting to reduce the effects of the condition faster as the desired medical outcome of the inclusion of the health artifacts by the physician.

[0571] In some embodiments, the processing device may receive input from a computing device (user device **104**). The input may specify a number and an area of the first condition the patient desires to manage. The area may include a type of health artifacts in the set of the health artifacts the patient selects to manage for the first condition. The processing device may select, based on the comparing, the another subset of the set of the health artifacts in the first data structure by selecting the another subset based on the number and the type of health artifacts specified by the patient.

[0572] FIG. **62** shows a method **6200** for using a net promoter score to update a machine learning model to output different health artifacts, in accordance with various embodiments. In some embodiments, the method **6200** is implemented on a cognitive intelligence platform. In some embodiments, the cognitive intelligence platform is the cognitive intelligence platform **102** as shown in FIG. **1**. In some embodiments, the cognitive intelligence platform is implemented on the computing device **1400** shown in FIG. **14**. The method **6200** may include operations that are implemented in computer instructions stored in a memory and executed by a processor of a computing device. In some embodiments, the method **6200** includes operations performed by the cognitive agent **110** (autonomous multipurpose application), the knowledge cloud **106**, and/or the critical thinking engine **108** of the cognitive intelligence platform **102** as shown in FIG. **1**. The operations of the method **6200** in FIG. **62** may be performed in some combination with the operations of the method **6100** in FIG. **61**.

[0573] At block **6202**, the processing device may generate a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan. A net promoter score may be used to gauge the loyalty of a customer and an entity providing the care plan. The net promoter score may be generated based on feedback received from patients, medical personnel, and/or administrators that use the care plan. The feedback may specify how likely the patients, medical personnel, and/or administrators are to recommend the cognitive intelligence platform **102**, the features (e.g., generation of useful care plans and modifying the care plans based on tone, emotion, and/or medical outcome) of the cognitive intelligence platform **102**, and the like. The net promoter score may be generated by subtracting the percentage of customers who respond between a first range (e.g., scores from 0 and 6) from the percentage of customers who respond with a score between a second range (e.g., scores from 9 to 10).

[0574] At block **6204**, the processing device may update a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition. For example, training data may be generated by collecting the care plans for medical conditions that received scores in the second range (high scores, positive feedback) and the care plans for medical conditions that received scores in the first range (low scores, positive feedback), and determining the differences in the care plans that resulted in the scores in the first range and the second range. The training data may include input data of the condition and output data of the care plans based on the differences.

[0575] The various aspects, embodiments, implementations or features of the described embodiments can be used separately or in any combination. Various aspects of the described embodiments can be implemented by software, hardware or a combination of hardware and software. The described embodiments can also be embodied as computer readable code on a computer readable medium. The computer readable medium is any data storage device that can store data which can thereafter be read by a computer system. Examples of the computer readable

medium include read-only memory, random-access memory, CD-ROMs, DVDs, magnetic tape, hard disk drives, solid-state drives, and optical data storage devices. The computer readable medium can also be distributed over network-coupled computer systems so that the computer readable code is stored and executed in a distributed fashion.

[0576] Consistent with the above disclosure, the examples of systems and method enumerated in the following clauses are specifically contemplated and are intended as a non-limiting set of examples.

[0577] Clause 1. A cognitive intelligence platform, comprising: [0578] a first system configured to execute a knowledge cloud, the first system comprising: [0579] a first processor; and [0580] a first memory coupled to the first processor, the first memory storing instructions that cause the knowledge cloud to: [0581] receive inputs from medical facilities; and [0582] receive inputs from service providers; [0583] a second system configured to implement a critical thinking engine, the critical thinking engine communicably coupled to the knowledge cloud, the second system comprising: [0584] a second processor; and [0585] a second memory coupled to the second processor, the second memory storing instructions that cause the critical thinking engine to receive inputs from the knowledge cloud; and [0586] a third system configured to implement a cognitive agent, the cognitive agent communicably coupled to the critical thinking engine and the knowledge cloud, the third system comprising: [0587] a third processor; and [0588] a third memory coupled to the third processor, the third memory storing instructions that cause the cognitive agent to: [0589] receive an originating question from a user related to a subject matter; [0590] execute, using the critical thinking engine, a first round of analysis to generate an answer; and [0591] provide the answer to the user including a recommendation associated with the subject matter.

[0592] Clause 2. The cognitive intelligence platform of any preceding clause, wherein the second memory stores instructions that further cause the critical thinking engine to: [0593] receive a first information; [0594] receive a second information that contradicts the first information; and [0595] process the first information and second information.

[0596] Clause 3. The cognitive intelligence platform of any preceding clause, wherein the second memory stores instructions that further cause the critical thinking engine to: [0597] parse the originating question; [0598] retrieve data from the knowledge cloud; and [0599] perform a causal analysis of the data in view of the originating question, wherein the causal analysis, in part, informs the answer.

[0600] Clause 4. The cognitive intelligence platform of any preceding clause, wherein the second memory stores instructions that further cause the critical thinking engine to: [0601] receive the originating question from the cognitive agent; [0602] assess a first chain of logic associated with the originating question; [0603] assess a second chain of logic associated with the originating question; and provide the answer to the cognitive agent, wherein the answer is associated with the first chain of logic.

[0604] Clause 5. The cognitive intelligence platform of any preceding clause, wherein the third memory stores instructions that further cause the cognitive agent to communicate a logical argument that leads to a conclusion, wherein the conclusion, in part, informs the recommendation associated with the subject matter.

[0605] Clause 6. The cognitive intelligence platform of any preceding clause, wherein the third memory stores instructions that further cause the cognitive agent to: [0606] render for display, to the user, a chain of logic that leads to the conclusion; [0607] receive, from the user, an adjustment to the chain of logic; and [0608] affect change in the critical thinking engine.

[0609] Clause 7. The cognitive intelligence platform of any preceding clause, wherein the third memory stores instructions that further cause the cognitive agent to: [0610] render for display a micro survey; [0611] receive data associated with the micro survey, wherein the data, in part, informs the recommendation associated with the subject matter.

[0612] Clause 8. The cognitive intelligence platform of any preceding clause, wherein when the

cognitive agent provides the answer to the user, the third memory causes the cognitive agent to integrate data from at least three selected from the group consisting of: a micro survey, a physician's office, common sense knowledge, domain knowledge, an evidence-based medicine guideline, a clinical ontology, and curated medical advice.

[0613] Clause 9. A system comprising: [0614] a knowledge cloud; [0615] a critical thinking engine, the critical thinking engine communicably coupled to the knowledge cloud; and [0616] a cognitive agent, the cognitive agent communicably coupled to the critical thinking engine and the knowledge cloud, wherein the cognitive agent is configured to interact with a user using natural language.

[0617] Clause 10. The system of any preceding clause, wherein the cognitive agent interacts with the user using at least one selected from the group consisting of: touch-based input, audio input, and typed input.

[0618] Clause 11. The system of claim any preceding clause, wherein the critical thinking engine is configured to: [0619] receive a first information; [0620] receive a second information that contradicts the first information; and [0621] process the first information and the second information.

[0622] Clause 12. The system of any preceding clause, wherein the cognitive agent is configured to: [0623] receive an originating question from the user related to a subject matter; [0624] execute, using the critical thinking engine, a logical reasoning to generate an answer; and [0625] provide the answer to the user including a recommendation associated with the subject matter.

[0626] Clause 13. The system of any preceding clause, wherein the critical thinking engine is configured to: [0627] parse the originating question; [0628] retrieve data from the knowledge cloud; and [0629] perform a causal analysis of the data in view of the originating question, wherein the causal analysis, in part informs the answer.

[0630] Clause 14. The system of any preceding clause, wherein the critical thinking engine is configured to: [0631] receive the originating question from the cognitive agent; [0632] assess a first chain of logic associated with the originating question; [0633] assess a second chain of logic associated with the originating question; and [0634] provide the answer to the cognitive agent, wherein the answer is associated with the first chain of logic.

[0635] Clause 15. The system of any preceding clause, wherein the cognitive agent is further configured to render for display a chain of logic that leads to a conclusion, wherein the conclusion, in part, informs the answer.

[0636] Clause 16. A computer readable media storing instructions that are executable by a processor to cause a computer to execute operations comprising: [0637] executing a cognitive intelligence platform that further comprises: [0638] a knowledge cloud; [0639] a critical thinking engine communicably coupled to the knowledge cloud; and [0640] a cognitive agent communicably coupled to the critical thinking engine and the knowledge cloud, wherein the cognitive agent is configured to: [0641] receive an originating question from a user related to a subject matter; [0642] execute, using the critical thinking engine, a logical reasoning to generate an answer; and [0643] provide the answer to the user including a recommendation associated with the subject matter.

[0644] Clause 17. The computer-readable media of any preceding clause, wherein the cognitive agent executing within the cognitive intelligence platform is further configured to: [0645] render for display a micro survey; [0646] receive data associated with the micro survey, wherein the data, in part, informs the recommendation associated with the subject matter.

[0647] Clause 18. The computer-readable media of any preceding clause, wherein the critical thinking engine executing within the cognitive intelligence platform is further configured to: [0648] receive the originating question from the cognitive agent; [0649] assess a first chain of logic associated with the originating question to create a first answer; [0650] assess a second chain of logic associated with the originating question to create a second answer, wherein the first answer

contradicts the second answer; and [0651] provide the first answer to the cognitive agent, wherein the first answer is the answer provided to the user.
[0652] Clause 19. The computer-readable media of any preceding clause, wherein the cognitive agent executing within the cognitive intelligence platform is further configured to render for display the first chain of logic to the user.
[0653] Clause 20. The computer-readable media of any preceding clause, wherein the cognitive agent executing within the cognitive intelligence platform is further configured to integrate data from at least three selected from the group consisting of: a micro survey, a physician's office, common sense knowledge, domain knowledge, an evidence-based medicine guideline, a clinical ontology, and curated medical advice.
[0654] Clause 21. A computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template, the method comprising: [0655] receiving a user-generated natural language medical information query at an artificial intelligence-based diagnostic conversation agent from a user interface on a mobile device; [0656] responsive to content of the user-generated natural language medical information query, selecting a diagnostic fact variable set relevant to generating a medical advice query answer for the user-generated natural language medical information query by classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications associated with respective diagnostic fact variable sets; [0657] compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set, wherein the compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set further comprises: [0658] extracting a first set of user-specific medical fact variable values from a local user medical information profile associated with the user-generated natural language medical information query, and [0659] requesting a second set of user-specific medical fact variable values through natural-language questions sent to the user interface on the mobile device; and [0660] responsive to the user-specific medical fact variable values, generating a medical advice query answer in response to the user-generated natural language medical information query.
[0661] Clause 22. The computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template of any preceding clause, wherein the compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set further comprises: [0662] extracting a third set of user-specific medical fact variable values comprising lab result values from the local user medical information profile associated with the user-generated natural language medical information query.
[0663] Clause 23. The computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template of any preceding clause, wherein the compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set further comprises: [0664] extracting a fourth set of user-specific medical fact variable values from a remote medical data service profile associated with the local user medical information profile.
[0665] Clause 24. The computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template of any preceding clause, wherein the compiling user-specific medical fact variable values for one or more respective medical fact variables of the diagnostic fact variable set further comprises: [0666] extracting a fifth set of user-specific medical fact variable values derived from demographic characterizations provided by a remote data service analysis of the local user medical information profile.
[0667] Clause 25. The computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template of any

preceding clause, wherein the generating the medical advice query answer in response to the user-generated natural language medical information query further comprises providing, in addition to text responsive to a medical question presented in the user-generated natural language medical information query, a treatment action-item recommendation responsive to user-specific medical fact variable values and non-responsive to the medical question presented in the user-generated natural language medical information query.

[0668] Clause 26. The computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template of any preceding clause, wherein the generating the medical advice query answer in response to the user-generated natural language medical information query further comprises providing, in addition to text responsive to a medical question presented in the user-generated natural language medical information query, a medical education media resource responsive to the user-specific medical fact variable values and non-responsive to the medical question presented in the user-generated natural language medical information query.

[0669] Clause 27. The computer-implemented method for answering a user-generated natural language medical information query based on a diagnostic conversational template of any preceding clause, wherein selecting a diagnostic fact variable set relevant to generating a medical advice query answer for the user-generated natural language medical information query by classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications associated with respective diagnostic fact variable set further comprises classifying the user-generated natural language medical information query into one of a set of domain-directed medical query classifications based on relevance to the local user medical information profile associated with the user-generated natural language medical information query.

[0670] Clause 28. A computer program product in a computer-readable medium for answering a user-generated natural language query, the computer program product in a computer-readable medium comprising program instructions which, when executed, cause a processor of a computer to perform: [0671] receiving a user-generated natural language query at an artificial intelligence-based conversation agent from a user interface; [0672] responsive to content of the user-generated natural language query, selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets; [0673] compiling user-specific fact variable values for one or more respective fact variables of the fact variable set; and [0674] responsive to the fact variable values, generating the query answer in response to the user-generated natural language query.

[0675] Clause 29. The computer program product in a computer-readable medium for answering a user-generated natural language query of any preceding clause, wherein the program instructions which, when executed, cause the processor of the computer to perform compiling user-specific fact variable values for one or more respective fact variables of the fact variable set further comprise program instructions which, when executed, cause the computer program product to perform: [0676] extracting a first set of user-specific fact variable values from a local user profile associated with the user-generated natural language query; and [0677] requesting a second set of user-specific fact variable values through a conversational template comprising natural-language questions sent to the user interface on a mobile device.

[0678] Clause 30. The computer program product in a computer-readable medium for answering a user-generated natural language query of any preceding clause, wherein the program instructions which, when executed, cause the processor of the computer to perform compiling user-specific fact variable values for one or more respective fact variables of the fact variable set further comprise program instructions which, when executed, cause the computer program product to perform: [0679] extracting a third set of user-specific fact variable values from a remote data service profile

associated with the local user profile.
[0680] Clause 31. The computer program product in a computer-readable medium for answering a user-generated natural language query of any preceding clause, wherein the program instructions which, when executed, cause the processor of the computer to perform compiling user-specific fact variable values for one or more respective fact variables of the fact variable set further comprise program instructions which, when executed, cause the computer program product to perform:
[0681] extracting a fourth set of user-specific fact variable values derived from demographic characterizations provided by a remote data service analysis of the local user profile.
[0682] Clause 32. The computer program product in a computer-readable medium for answering a user-generated natural language query of any preceding clause, wherein program instructions which, when executed, cause the processor of the computer to perform the generating the query answer in response to the user-generated natural language query further comprise program instructions which, when executed, cause the processor of the computer to perform providing, in addition to text responsive to a question presented in the user-generated natural language query, an action-item recommendation responsive to the fact variable values and non-responsive to the question presented in the user-generated natural language query.
[0683] Clause 33. The computer program product in a computer-readable medium for answering a user-generated natural language query of any preceding clause, wherein the program instructions which, when executed, cause the processor of the computer to perform generating the query answer in response to the user-generated natural language query further comprise program instructions which, when executed, cause the processor of the computer to perform providing, in addition to text responsive to a question presented in the user-generated natural language query, an education media resource responsive to the fact variable values and non-responsive to the question presented in the user-generated natural language query.
[0684] Clause 34. The computer program product in a computer-readable medium for answering a user-generated natural language query of any preceding clause, wherein the program instructions which, when executed, cause the processor of the computer to perform selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets further comprise program instructions which, when executed, cause the processor of the computer to perform classifying the user-generated natural language query into one of a set of domain-directed query classifications based on relevance to a local user profile associated with the user-generated natural language query.
[0685] Clause 35. A cognitive intelligence platform for answering a user-generated natural language query, the cognitive intelligence platform comprising: [0686] a cognitive agent configured for receiving a user-generated natural language query at an artificial intelligence-based conversation agent from a user interface; [0687] a critical thinking engine configured for, responsive to content of the user-generated natural language query, selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets; and a knowledge cloud compiling user-specific fact variable values for one or more respective fact variables of the fact variable set; and wherein, responsive to the fact variable values, the cognitive agent is further configured for generating the query answer in response to the user-generated natural language query.
[0688] Clause 36. The cognitive intelligence platform of any preceding clause, wherein the knowledge cloud is further configured for: [0689] extracting a first set of user-specific fact variable values from a local user profile associated with the user-generated natural language query; and
[0690] requesting a second set of user-specific fact variable values through a conversational template comprising natural-language questions sent to the user interface on a mobile device.
[0691] Clause 37. The cognitive intelligence platform of any preceding clause, wherein the

knowledge cloud is further configured for: [0692] extracting a third set of user-specific fact variable values from a remote data service profile associated with the local user profile.
[0693] Clause 38. The cognitive intelligence platform of any preceding clause, wherein the knowledge cloud is further configured for: [0694] extracting a fourth set of user-specific fact variable values derived from demographic characterizations provided by a remote data service analysis of the local user profile.
[0695] Clause 39. The cognitive intelligence platform of any preceding clause, wherein cognitive agent is further configured for providing, in addition to text responsive to a question presented in the user-generated natural language query, an action-item recommendation responsive to the fact variable values and non-responsive to the question presented in the user-generated natural language query.
[0696] Clause 40. The cognitive intelligence platform of any preceding clause, wherein the critical thinking engine is further configured for providing, in addition to text responsive to a question presented in the user-generated natural language query, an education media resource responsive to the fact variable values and non-responsive to the question presented in the user-generated natural language query.
[0697] Clause 41. A computer-implemented method for answering a user-generated natural language query, the method comprising: [0698] receiving a user-generated natural language query at an artificial intelligence-based conversation agent from a user interface; [0699] responsive to content of the user-generated natural language query, selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets; [0700] compiling user-specific fact variable values for one or more respective fact variables of the fact variable set; and [0701] responsive to the fact variable values, generating the query answer in response to the user-generated natural language query.
[0702] Clause 42. The method of any preceding clause, wherein the compiling user-specific fact variable values for one or more respective fact variables of the fact variable set further comprises: [0703] extracting a first set of user-specific fact variable values from a local user profile associated with the user-generated natural language query; and [0704] requesting a second set of user-specific fact variable values through a conversational template comprising natural-language questions sent to the user interface on a mobile device.
[0705] Clause 43. The method of any preceding clause, wherein the compiling user-specific fact variable values for one or more respective fact variables of the fact variable set further comprises: [0706] extracting a third set of user-specific fact variable values from a remote data service profile associated with the local user profile.
[0707] Clause 44. The method of any preceding clause, wherein the compiling user-specific fact variable values for one or more respective fact variables of the fact variable set further comprises: [0708] extracting a fourth set of user-specific fact variable values derived from demographic characterizations provided by a remote data service analysis of the local user profile.
[0709] Clause 45. The method of any preceding clause, wherein the generating the query answer in response to the user-generated natural language query further comprises providing, in addition to text responsive to a question presented in the user-generated natural language query, an action-item recommendation responsive to the fact variable values and non-responsive to the question presented in the user-generated natural language query.
[0710] Clause 46. The method of any preceding clause, wherein the generating the query answer in response to the user-generated natural language query further comprises providing, in addition to text responsive to a question presented in the user-generated natural language query, an education media resource responsive to the fact variable values and non-responsive to the question presented in the user-generated natural language query.

[0711] Clause 47. The method of any preceding clause, wherein selecting a fact variable set relevant to generating a query answer for the user-generated natural language query by classifying the user-generated natural language query into one of a set of domain-directed query classifications associated with respective fact variable sets further comprises classifying the user-generated natural language query into one of a set of domain-directed query classifications based on relevance to a local user profile associated with the user-generated natural language query.

[0712] Clause 48. A computer-implemented method for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system, the method comprising: [0713] receiving from a medical conversational user interface a user-generated natural language medical information query at an artificial intelligence-based medical conversation cognitive agent; [0714] extracting from the user-generated natural language medical information query a medical question from a user of the medical conversational user interface; [0715] compiling a medical conversation language sample, wherein the medical conversation language sample comprises items of health-information-related-text derived from a health-related conversation between the artificial intelligence-based medical conversation cognitive agent and the user; [0716] extracting from the medical conversation language sample internal medical concepts and medical data entities present within the medical conversation language sample, wherein the internal medical concepts comprise descriptions of medical attributes of the medical data entities; [0717] inferring a therapeutic intent of the user from the internal medical concepts and the medical data entities; [0718] generating a therapeutic paradigm logical framework for interpreting of the medical question, wherein [0719] the therapeutic paradigm logical framework comprises a catalog of medical logical progression paths from the medical question to respective therapeutic answers, [0720] each of the medical logical progression paths comprises one or more medical logical linkages from the medical question to a therapeutic path-specific answer, and [0721] the medical logical linkages comprise the internal medical concepts and external therapeutic paradigm concepts derived from a store of medical subject matter ontology data; [0722] selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based upon the therapeutic intent of the user; and [0723] answering the medical question by following the likely medical information path to the likely path-dependent medical information answer.

[0724] Clause 49. The computer-implemented method for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system of any of any of the preceding clauses, further comprising relating medical inference groups of the internal medical concepts.

[0725] Clause 50. The computer-implemented method for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system of any of any of the preceding clauses, wherein the relating medical inference groups of the internal medical concepts further comprises relating groups of the internal medical concepts based at least in part on shared medical data entities for which each internal medical concept of a medical inference group of internal medical concepts describes a respective medical data attribute.

[0726] Clause 51. The computer-implemented method for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system of any of the preceding clauses, wherein selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based upon the intent further comprises selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based in part upon the therapeutic intent of the user and in part upon sufficiency of medical diagnostic data to complete the medical logical linkages.

[0727] Clause 52. The computer-implemented method for answering natural language medical

information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system of any of the preceding clauses, wherein selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based upon the intent further comprises selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer after requesting additional medical diagnostic data from the user.

[0728] Clause 53. The computer-implemented method for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system of any of the preceding clauses, wherein selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based upon the intent further comprises selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based in part upon treatment sub-intents comprising tactical constituents related to the therapeutic intent of the user by the store of medical subject matter ontology data.

[0729] Clause 54. The computer-implemented method for answering natural language medical information questions posed by a user of a medical conversational interface of a cognitive artificial intelligence system of any of the preceding clauses, wherein selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based upon the intent further comprises selecting a likely medical information path from among the medical logical progression paths to a likely path-dependent medical information answer based in part upon the therapeutic intent of the user and in part upon sufficiency of medical diagnostic data to complete the medical logical linkages, wherein the medical diagnostic data to complete the medical logical linkages includes user-specific medical diagnostic data.

[0730] Clause 55. A cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system, the cognitive intelligence platform comprising: [0731] a cognitive agent configured for receiving from a user interface a user-generated natural language query, wherein the cognitive agent is an artificial intelligence-based conversation agent; [0732] a knowledge cloud containing a store of subject matter ontology data; [0733] a critical thinking engine configured for: [0734] extracting from the user-generated natural language query a question from a user of the user interface, [0735] compiling a language sample, wherein the language sample comprises items of text derived from a conversation between the artificial intelligence-based conversation agent and the user, [0736] extracting from the language sample internal concepts and entities present within the language sample, wherein the internal concepts comprise descriptions of attributes of the entities, [0737] inferring an intent of the user from the internal concepts and the entities, [0738] generating a logical framework for interpreting of the question, wherein [0739] the logical framework comprises a catalog of paths from the question to respective answers, [0740] each of the paths comprises one or more linkages from the question to a path-specific answer, and [0741] the linkages comprise the internal concepts and external concepts derived from the store of subject matter ontology data, [0742] selecting a likely path from among the paths to a likely path-dependent answer based upon the intent, and [0743] answering the question by following the likely path to the likely path-dependent answer.

[0744] Clause 56. The cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the critical thinking engine is further configured for relating groups of the internal concepts.

[0745] Clause 57. The cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the critical thinking engine is further configured for relating groups of the internal concepts by relating groups of the internal concepts based at least in part on shared

entities for which each internal concept of a group of internal concepts describes a respective attribute.
[0746] Clause 58. The cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the critical thinking engine is further configured for selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer based in part upon the intent and in part upon sufficiency of data to complete the linkages.
[0747] Clause 59. The cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the critical thinking engine is further configured for selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer after requesting additional data from the user.
[0748] Clause 60. The cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of 8, wherein the critical thinking engine is further configured for selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer based in part upon sub-intents comprising tactical constituents related to the intent by the store of subject matter ontology data.
[0749] Clause 61. The cognitive intelligence platform for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the critical thinking engine is further configured for selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer based in part upon the intent and in part upon sufficiency of data to complete the linkages, wherein the data to complete the linkages includes user-specific data.
[0750] Clause 62. A computer program product in a computer-readable medium for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system, the computer program product in a computer-readable medium comprising instructions, which, when executed, cause a processor of a computer to perform: [0751] receiving from a user interface a user-generated natural language query at an artificial intelligence-based conversation agent; [0752] extracting from the user-generated natural language query a question from a user of the user interface; [0753] compiling a language sample, wherein the language sample comprises items of text derived from a conversation between the artificial intelligence-based conversation agent and the user; [0754] extracting from the language sample internal concepts and entities present within the language sample, wherein the internal concepts comprise descriptions of attributes of the entities; [0755] inferring an intent of the user from the internal concepts and the entities; [0756] generating a logical framework for interpreting of the question, wherein [0757] the logical framework comprises a catalog of paths from the question to respective answers, [0758] each of the paths comprises one or more linkages from the question to a path-specific answer, and [0759] the linkages comprise the internal concepts and external concepts derived from a store of subject matter ontology data; [0760] selecting a likely path from among the paths to a likely path-dependent answer based upon the intent; and [0761] answering the question by following the likely path to the likely path-dependent answer.
[0762] Clause 63. The computer program product in a computer-readable medium for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, further comprising instructions, which, when executed, cause the processor of the computer to perform relating groups of the internal concepts.
[0763] Clause 64. The computer program product in a computer-readable medium for answering

natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the instructions, which, when executed, cause the processor of the computer to perform relating groups of the internal concepts further comprise instructions, which, when executed, cause the processor of the computer to perform relating groups of the internal concepts based at least in part on shared entities for which each internal concept of a group of internal concepts describes a respective attribute.

[0764] Clause 65. The computer program product in a computer-readable medium for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the instructions, which, when executed, cause the processor of the computer to perform selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprise instructions, which, when executed, cause the processor of the computer to perform selecting a likely path from among the paths to a likely path-dependent answer based in part upon the intent and in part upon sufficiency of data to complete the linkages.

[0765] Clause 66. The computer program product in a computer-readable medium for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein instructions, which, when executed, cause the processor of the computer to perform selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprise instructions, which, when executed, cause the processor of the computer to perform selecting a likely path from among the paths to a likely path-dependent answer after requesting additional data from the user.

[0766] Clause 67. The computer program product in a computer-readable medium for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein the instructions, which, when executed, cause the processor of the computer to perform selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprise instructions, which, when executed, cause the processor of the computer to perform selecting a likely path from among the paths to a likely path-dependent answer based in part upon sub-intents comprising tactical constituents related to the intent by the store of subject matter ontology data.

[0767] Clause 68. A method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system, the method comprising: [0768] receiving from a user interface a user-generated natural language query at an artificial intelligence-based conversation agent; [0769] extracting from the user-generated natural language query a question from a user of the user interface; [0770] compiling a language sample, wherein the language sample comprises items of text derived from a conversation between the artificial intelligence-based conversation agent and the user; [0771] extracting from the language sample internal concepts and entities present within the language sample, wherein the internal concepts comprise descriptions of attributes of the entities; [0772] inferring an intent of the user from the internal concepts and the entities; [0773] generating a logical framework for interpreting of the question, wherein [0774] the logical framework comprises a catalog of paths from the question to respective answers, [0775] each of the paths comprises one or more linkages from the question to a path-specific answer, and [0776] the linkages comprise the internal concepts and external concepts derived from a store of subject matter ontology data; [0777] selecting a likely path from among the paths to a likely path-dependent answer based upon the intent; and [0778] answering the question by following the likely path to the likely path-dependent answer.

[0779] Clause 69. The method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, further comprising relating groups of the internal concepts.

[0780] Clause 70. The method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein

the relating groups of the internal concepts further comprises relating groups of the internal concepts based at least in part on shared entities for which each internal concept of a group of internal concepts describes a respective attribute.
[0781] Clause 71. The method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer based in part upon the intent and in part upon sufficiency of data to complete the linkages.
[0782] Clause 72. The method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer after requesting additional data from the user.
[0783] Clause 73. The method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer based in part upon sub-intents comprising tactical constituents related to the intent by the store of subject matter ontology data.
[0784] Clause 74. The method for answering natural language questions posed by a user of a conversational interface of an artificial intelligence system of any of the preceding clauses, wherein selecting a likely path from among the paths to a likely path-dependent answer based upon the intent further comprises selecting a likely path from among the paths to a likely path-dependent answer based in part upon the intent and in part upon sufficiency of data to complete the linkages, wherein the data to complete the linkages includes user-specific data.
[0785] Clause 75. A computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream, the computer-implemented method comprising: [0786] receiving segments of a medical information natural language conversation stream at an artificial intelligence-based health information conversation agent from a medical information conversation user interface; [0787] responsive to medical information content of a user medical information profile associated with the medical information natural language conversation stream, defining a desired clinical management outcome objective relevant to health management criteria and related health management data attributes of the user medical information profile; [0788] identifying a set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective; [0789] selecting from among the set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective a medical intervention likely to advance the clinical management outcome objective; [0790] presenting in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the medical intervention likely to advance the clinical management outcome objective; and [0791] presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining a correlation between the medical intervention likely to advance the clinical management outcome objective and achievement of the clinical management outcome objective.
[0792] Clause 76. The computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream of any preceding clause, wherein the selecting from among the set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective a medical intervention likely to advance the clinical management outcome objective further comprises: [0793] selecting from among the set of potential therapeutic interventions correlated to advancement of the clinical

management outcome objective the medical intervention likely to advance the clinical management outcome objective based on a set of factors comprising likelihood of patient compliance with the a recommendation for the a medical intervention likely to advance the clinical management outcome objective and a statistical likelihood that the action will materially advance the clinical management outcome objective.

[0794] Clause 77. The computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining a cost-benefit analysis comparing likely results of performance of the action likely to advance the clinical management outcome objective and likely results of non-performance of the action likely to advance the clinical management outcome objective.

[0795] Clause 78. The computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream of any preceding clause, wherein the selecting from among the set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective a medical intervention likely to advance the clinical management outcome objective further comprises: [0796] selecting from among the set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective the medical intervention likely to advance the clinical management outcome objective based on a set of factors comprising likelihood total expected cost expectation associated with the recommendation for the a medical intervention likely to advance the clinical management outcome objective.

[0797] Clause 79. The computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream of any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a conversation stream reinforcing the recommendation after expiration of a delay period.

[0798] Clause 80. The computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream of any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining reasons for selection of the clinical management outcome objective.

[0799] Clause 81. The computer-implemented method for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream of any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises notifying third party service providers of the clinical management outcome objective and the recommendation.

[0800] Clause 82. A computer program product in a non-transitory computer-readable medium for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream, the computer program product in a non-transitory computer-

readable medium comprising instructions which, when executed cause a processor of a computer to perform: [0801] receiving segments of a medical information natural language conversation stream at an artificial intelligence-based health information conversation agent from a medical information conversation user interface; [0802] responsive to medical information content of a user medical information profile associated with the medical information natural language conversation stream, defining a clinical management outcome objective relevant to health management criteria and related health management data attributes of the profile; [0803] selecting a medical intervention likely to advance the clinical management outcome objective; and [0804] presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective.

[0805] 83. The computer program product in a non-transitory computer-readable medium of any preceding clause, wherein the instructions which, when executed cause the processor of the computer to perform selecting a medical intervention likely to advance the clinical management outcome objective further comprise instructions which, when executed cause the processor of the computer to perform: [0806] identifying a set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective; and [0807] selecting the action likely to advance the user outcome objective based on a set of factors comprising likelihood of performance of the action likely to advance the user outcome objective and likelihood that the action will materially advance the user outcome objective.

[0808] Clause 84. The computer program product in a non-transitory computer-readable medium of any preceding clause, wherein the instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprise instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining a correlation between the action likely to advance the clinical management outcome objective and achievement of the clinical management outcome objective.

[0809] Clause 85. The computer program product in a non-transitory computer-readable medium of any preceding clause, wherein the instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprise instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining a plan of subsequent actions likely to advance the clinical management outcome objective.

[0810] Clause 86. The computer program product in a non-transitory computer-readable medium of any preceding clause, wherein the instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprise instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a conversation stream reinforcing the recommendation after expiration of a delay period.

[0811] Clause 87. The computer program product in a non-transitory computer-readable medium of any preceding clause, wherein the instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language

conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprise instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining reasons for selection of the clinical management outcome objective.

[0812] Clause 88. The computer program product in a non-transitory computer-readable medium of any preceding clause, wherein the instructions which, when executed cause the processor of the computer to perform presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprise instructions which, when executed cause the processor of the computer to perform notifying third party service providers of the clinical management outcome objective and the recommendation.

[0813] Clause 89. A system for providing therapeutic medical action recommendations in response to a medical information natural language conversation stream, the system comprising: [0814] a knowledge cloud configured for receiving segments of a medical information natural language conversation stream at an artificial intelligence-based health information from a medical information conversation user interface of a cognitive agent; [0815] a critical thinking engine configured for: [0816] responsive to medical information content of a user medical information profile associated with the medical information natural language conversation stream in the knowledge cloud, defining a clinical management outcome objective relevant to health management criteria and related health management data attributes of the profile, and [0817] selecting a medical intervention likely to advance the clinical management outcome objective; and [0818] the cognitive agent, wherein the cognitive agent is configure for presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective.

[0819] Clause 90. The system of any preceding clause, wherein the selecting a medical intervention likely to advance the clinical management outcome objective further comprises: [0820] identifying a set of potential therapeutic interventions correlated to advancement of the clinical management outcome objective; and [0821] selecting the action likely to advance the user outcome objective based on a set of factors comprising likelihood of performance of the action likely to advance the user outcome objective and likelihood that the action will materially advance the user outcome objective.

[0822] Clause 91. The system of claim any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining a correlation between the action likely to advance the clinical management outcome objective and achievement of the clinical management outcome objective.

[0823] Clause 92. The system of any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream segment explaining a plan of subsequent actions likely to advance the clinical management outcome objective.

[0824] Clause 93. The system of any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a therapeutic advice conversation stream

segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a conversation stream reinforcing the recommendation after expiration of a delay period.
[0825] Clause 94. The system of any preceding clause, wherein the presenting to the user in the medical information natural language conversation stream a conversation stream segment designed to stimulate execution of the action likely to advance the clinical management outcome objective further comprises presenting to the user in the medical information natural language conversation stream a conversation stream segment explaining reasons for selection of the clinical management outcome objective.
[0826] Clause 95. A computer-implemented method for providing action recommendations in response to a user-generated natural language conversation stream, the method comprising: [0827] receiving segments of a user-generated natural language conversation stream at an artificial intelligence-based conversation agent from a user interface; [0828] responsive to content of a user profile associated with the user-generated natural language conversation stream, defining a user action outcome objective relevant to attributes of the profile; [0829] selecting an action likely to advance the user action outcome objective; and [0830] presenting to the user in the user-generated natural language conversation stream a conversation stream segment designed to motivate performance of the action likely to advance the user action outcome objective.
[0831] Clause 96. The method of any preceding clause, wherein the selecting an action likely to advance the user action outcome objective further comprises: [0832] identifying a set of actions correlated to advancement of the user action outcome objective; and [0833] selecting the action likely to advance the user outcome objective based on a set of factors comprising likelihood of performance of the action likely to advance the user outcome objective and likelihood that the action will materially advance the user outcome objective.
[0834] Clause 97. The method of any preceding clause, wherein the presenting to the user in the user-generated natural language conversation stream a conversation stream segment designed to motivate performance of the action likely to advance the user action outcome objective further comprises presenting to the user in the user-generated natural language conversation stream a conversation stream segment explaining a correlation between the action likely to advance the user action outcome objective and achievement of the user action outcome objective.
[0835] Clause 98. The method of any preceding clause, wherein the presenting to the user in the user-generated natural language conversation stream a conversation stream segment designed to motivate performance of the action likely to advance the user action outcome objective further comprises presenting to the user in the user-generated natural language conversation stream a conversation stream segment explaining a plan of subsequent actions likely to advance the user action outcome objective.
[0836] Clause 99. The method of any preceding clause, wherein the presenting to the user in the user-generated natural language conversation stream a conversation stream segment designed to motivate performance of the action likely to advance the user action outcome objective further comprises presenting to the user in the user-generated natural language conversation stream a conversation stream reinforcing the recommendation after expiration of a delay period.
[0837] Clause 100. The method of any preceding clause, wherein the presenting to the user in the user-generated natural language conversation stream a conversation stream segment designed to motivate performance of the action likely to advance the user action outcome objective further comprises presenting to the user in the user-generated natural language conversation stream a conversation stream segment explaining reasons for selection of the user action outcome objective.
[0838] Clause 101. The method of any preceding clause, wherein the presenting to the user in the user-generated natural language conversation stream a conversation stream segment designed to motivate performance of the action likely to advance the user action outcome objective further

comprises notifying third party service providers of the user action outcome objective and the recommendation.
[0839] Clause 102. A method comprising: [0840] receiving, at an artificial intelligence engine, a corpus of data for a patient, wherein the corpus of data includes a plurality of strings of characters; [0841] identifying, in the plurality of strings of characters, indicia comprising a phrase, a predicate, a keyword, a subject, an object, a cardinal, a number, a concept, or some combination thereof; [0842] comparing the indicia to a knowledge graph representing known health related information to generate a possible health related information pertaining to the patient; [0843] identifying, using a logical structure, a structural similarity of the possible health related information and a known predicate in the logical structure; and [0844] generating, by the artificial intelligence engine, cognified data based on the structural similarity.
[0845] Clause 103. The method of any preceding clause, further comprising generating the knowledge graph using the known health related information, wherein the knowledge graph represents knowledge of a disease and the knowledge graph comprises a plurality of concepts pertaining to the disease obtained from the known health related information, and the knowledge graph comprises relationships between the plurality of concepts.
[0846] Clause 104. The method of any preceding clause, wherein the cognified data comprises a health related summary of the possible health related information.
[0847] Clause 105. The method of any preceding clause, wherein generating, by the artificial intelligence engine, the cognified data further comprises: [0848] generating at least one new string of characters representing a statement pertaining to the possible health related information; and [0849] including the at least one new string of characters in the health related summary of the possible health related information.
[0850] Clause 106. The method of any preceding clause, wherein the statement describes an effect that results from the possible health related information.
[0851] Clause 107. The method of any preceding clause, further comprising codifying evidence based health related guidelines pertaining to a disease to generate the logical structure.
[0852] Clause 108. The method of any preceding clause, further comprising: [0853] identifying at least one piece of information missing in the corpus of data for the patient using the cognified data, wherein the at least one piece of information pertains to a treatment gap, a risk gap, a quality of care gap, or some combination thereof; and [0854] causing a notification to be presented on a computing device of a healthcare personnel, wherein the notification instructs entry of the at least one piece of information.
[0855] Clause 109. The method of any preceding clause, wherein using the logical structure to identify the structural similarity of the indicia and the known predicate in the logical structure further comprises identifying, based on the structural similarity of the indicia and the known predicate in the logical structure, a treatment pattern, a referral pattern, a quality of care pattern, a risk adjustment pattern, or some combination thereof in the corpus of data.
[0856] Clause 110. The method of any preceding clause, further comprising: [0857] receiving feedback pertaining to whether the cognified data is accurate; and [0858] updating the artificial intelligence engine based on the feedback.
[0859] Clause 111. The method of any preceding clause, a tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to execute an artificial intelligence engine to: [0860] receive a corpus of data for a patient, wherein the corpus of data includes a plurality of strings of characters; [0861] identify, in the plurality of strings of characters, indicia comprising a phrase, a predicate, a keyword, a cardinal, a number, a concept, or some combination thereof; [0862] compare the indicia to a knowledge graph representing known health related information to generate a possible health related information pertaining to the patient; [0863] identify, using a logical structure, a structural similarity of the indicia and a known predicate in the logical structure; and [0864] generate cognified data based on the similarity and the

possible health related information.
[0865] Clause 112. The computer-readable medium of any preceding clause, wherein the artificial intelligence engine is further to generate the knowledge graph using the known health related information, wherein the knowledge graph represents knowledge of a disease and the knowledge graph comprises a plurality of concepts pertaining to the disease obtained from the known health related information, and the knowledge graph comprises relationships between the plurality of concepts.
[0866] Clause 113. The computer-readable medium of any preceding clause, wherein the cognified data comprises a health related summary of the possible health related information.
[0867] Clause 114. The computer-readable medium of any preceding clause, wherein generating, based on the pattern, the cognified data further comprises: [0868] generating at least one new string of characters representing a statement pertaining to the possible health related information; and [0869] including the at least one new string of characters in the health related summary of the possible health related information.
[0870] Clause 115. The computer-readable medium of any preceding clause, wherein the statement describes an effect that results from the possible health related information
[0871] Clause 116. The computer-readable medium of any preceding clause, wherein the artificial intelligence engine is further to codify evidence based health related guidelines pertaining to a disease to generate the logical structure.
[0872] Clause 117. The computer-readable medium of any preceding clause, wherein the artificial intelligence engine is further to: [0873] identify at least one piece of information missing in the corpus of data for the patient using the cognified data, wherein the at least one piece of information pertains to a treatment gap, a risk gap, a quality of care gap, or some combination thereof; and [0874] cause a notification to be presented on a computing device of a healthcare personnel, wherein the notification instructs entry of the at least one piece of information.
[0875] Clause 118. The computer-readable medium of any preceding clause, wherein using the logical structure to identify the structural similarity of the indicia and the known predicate in the logical structure further comprises identifying, based on the structural similarity of the indicia and the known predicate in the logical structure, a treatment pattern, a referral pattern, a quality of care pattern, a risk adjustment pattern, or some combination thereof in the corpus of data.
[0876] Clause 119. The computer-readable medium of any preceding clause, wherein the artificial intelligence engine is further to: [0877] receive feedback pertaining to whether the cognified data is accurate; and [0878] update the artificial intelligence engine based on the feedback.
[0879] Clause 120. a system, comprising: [0880] a memory device storing instructions; and [0881] a processing device operatively coupled to the memory device, wherein the processing device executes the instructions to: [0882] receive, at an artificial intelligence engine, a corpus of data for a patient, wherein the corpus of data includes a plurality of strings of characters; [0883] identify, in the plurality of strings of characters, indicia comprising a phrase, a predicate, a keyword, a cardinal, a number, a concept, or some combination thereof; [0884] compare the indicia to a knowledge graph representing known health related information to generate a possible health related information pertaining to the patient; [0885] identify, using a logical structure, a structural similarity of the indicia and a known predicate in the logical structure; and [0886] generate, by the artificial intelligence engine, cognified data based on the similarity and the possible health related information.
[0887] Clause 121. The system of any preceding claim, wherein the processing device is further to: [0888] receive feedback pertaining to whether the cognified data is accurate; and [0889] update the artificial intelligence engine based on the feedback.
[0890] Clause 122. A method for controlling distribution of a plurality of information pertaining to a medical condition, the method comprising: [0891] receiving, at a server, an electronic medical record comprising notes pertaining to a patient; [0892] processing the notes to obtain indicia

comprising a word, a cardinal, a phrase, a sentence, a predicate, or some combination thereof;
[0893] identifying a possible medical condition of the patient by identifying a similarity between the indicia and a knowledge graph representing knowledge pertaining to the possible medical condition, wherein the knowledge graph comprises a plurality of nodes representing the plurality of information pertaining to the possible medical condition; and [0894] providing, at a first time, first information of the plurality of information to a computing device of the patient for presentation on the computing device, the first information being associated with a root node of the plurality of nodes.
[0895] Clause 123. The method of any preceding claim, further comprising providing, at a second time, second information of the plurality of information to the computing device of the patient for presentation on the computing device, the second information being associated with a second node of the plurality of nodes, and the second time being after the first time.
[0896] Clause 124. The method of any preceding claim, wherein the second information pertains to how the possible medical condition affects people, signs and symptoms of the possible medical condition, a way to treat the possible medical condition, a progression of the possible medical condition, or some combination thereof.
[0897] Clause 125. The method of any preceding claim, wherein the second time is selected based on when the second information is relevant to a stage of the possible medical condition.
[0898] Clause 126. The method of any preceding claim, further comprising providing, at a third time, third information of the plurality of information to the computing device of the patient for presentation on the computing device, the third information being associated with a third node of the plurality of nodes, and the third time being after the second time.
[0899] Clause 127. The method of any preceding claim, wherein identifying the possible medical condition by identifying the similarity between the indicia and the knowledge graph further comprises using an artificial intelligence engine that is trained using feedback from medical personnel, wherein the feedback pertains to whether output regarding possible medical conditions from the artificial intelligence engine is accurate for input comprising notes of patients.
[0900] Clause 128. The method of any preceding claim, wherein the first information pertains to a name of the possible medical condition, a definition of the possible medical condition, or some combination thereof.
[0901] Clause 129. The method of any preceding claim, wherein identifying the possible medical condition by identifying the similarity between the indicia and the knowledge graph further comprises using a cognified data structure generated from the notes of the patient, wherein the cognified data structure includes a conclusion based on a logical structure representing codified evidence based guidelines pertaining to the possible medical condition.
[0902] Clause 130. The method of any preceding claim, wherein processing the patient notes to obtain the indicia further comprises inputting the notes into an artificial intelligence engine trained to identify the indicia in text based on commonly used indicia pertaining to the possible medical condition.
[0903] Clause 131. The method of any preceding claim, further comprising: [0904] identifying a second possible medical condition of the patient by identifying a second similarity between the indicia and a second knowledge graph representing second knowledge pertaining to the second possible medical condition, wherein the second knowledge graph comprises a second plurality of nodes representing a second plurality of information pertaining to the second possible medical condition; and [0905] providing, at the first time, second information of the second plurality of information to the computing device of the patient for presentation on the computing device, the second information being associated with a second root node of the second plurality of nodes.
[0906] Clause 132. A tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to: [0907] receive an electronic medical record comprising notes pertaining to a patient; [0908] process the notes to obtain indicia comprising a

word, a cardinal, a phrase, a sentence, a predicate, or some combination thereof; [0909] identify a possible medical condition of the patient by identifying a similarity between the indicia and a knowledge graph representing knowledge pertaining to the possible medical condition, wherein the knowledge graph comprises a plurality of nodes representing the plurality of information pertaining to the possible medical condition; and [0910] provide, at a first time, first information of the plurality of information to a computing device of the patient for presentation on the computing device, the first information being associated with a root node of the plurality of nodes.

[0911] Clause 133. The computer-readable medium of any preceding clause, wherein the processing device is further to provide, at a second time, second information of the plurality of information to the computing device of the patient for presentation on the computing device, the second information being associated with a second node of the plurality of nodes, and the second time being after the first time.

[0912] Clause 134. The computer-readable medium of any preceding clause, wherein the second information pertains to how the possible medical condition affects people, signs and symptoms of the possible medical condition, a way to treat the possible medical condition, a progression of the possible medical condition, or some combination thereof.

[0913] Clause 135. The computer-readable medium of any preceding clause, wherein the second time is selected based on when the second information is relevant to a stage of the possible medical condition.

[0914] Clause 136. The computer-readable medium of any preceding clause, further comprising providing, at a third time, third information of the plurality of information to the computing device of the patient for presentation on the computing device, the third information being associated with a third node of the plurality of nodes, and the third time being after the second time.

[0915] Clause 137. The computer-readable medium of any preceding clause, wherein detecting the possible medical condition by identifying the similarity between the indicia and the knowledge graph further comprises using an artificial intelligence engine that is trained using feedback from medical personnel, wherein the feedback pertains to whether output regarding possible medical conditions from the artificial intelligence engine is accurate.

[0916] Clause 138. The computer-readable medium of any preceding clause, wherein the first information pertains to a name of the possible medical condition, a definition of the possible medical condition, or some combination thereof.

[0917] Clause 139. The computer-readable medium of any preceding clause, wherein detecting the possible medical condition by identifying the similarity between the indicia and the knowledge graph further comprises using a cognified data structure generated from the notes of the patient, wherein the cognified data structure includes a conclusion about the predicate that is identified in a logic structure representing codified evidence based guidelines pertaining to the possible medical condition.

[0918] Clause 140. The computer-readable medium of any preceding clause, wherein processing the patient notes to obtain the indicia further comprises inputting the notes into an artificial intelligence engine trained to identify the indicia in text based on commonly used indicia pertaining to the possible medical condition.

[0919] Clause 141. a system, comprising: [0920] a memory device storing instructions; [0921] a processing device communicatively coupled to the memory device, the processing device executes the instructions to: [0922] receive, at a server, an electronic medical record comprising notes pertaining to a patient; [0923] process the notes to obtain indicia comprising a word, a cardinal, a phrase, a sentence, a predicate, or some combination thereof; [0924] identify a possible medical condition of the patient by identifying a similarity between the indicia and a knowledge graph representing knowledge pertaining to the possible medical condition, wherein the knowledge graph comprises a plurality of nodes representing the plurality of information pertaining to the possible medical condition; and [0925] provide, at a first time, first information of the plurality of

information to a computing device of the patient for presentation on the computing device, the first information being associated with a root node of the plurality of nodes.
[0926] Clause 142. A method for diagnosing a medical condition through cognification of unstructured data, the method comprising: [0927] receiving, at a server, an electronic medical record comprising notes pertaining to a patient; [0928] generating cognified data using the notes, wherein the cognified data comprises a health summary of the medical condition; [0929] generating, based on the cognified data, a diagnosis of the medical condition of the patient, wherein the diagnosis at least identifies a type of the medical condition; and [0930] providing the diagnosis to a computing device for presentation on the computing device.
[0931] Clause 143. The method of any preceding clause, further comprising identifying, in the notes, indicia comprising a phrase, a predicate, a keyword, a cardinal, a number, a concept, or some combination thereof;
[0932] Clause 144. The method of any preceding clause, wherein generating the cognified data further comprises detecting the medical condition by identifying a similarity between the indicia and a knowledge graph.
[0933] Clause 145. The method of any preceding clause, further comprising using an artificial intelligence engine that is trained using feedback from medical personnel, wherein the feedback pertains to whether output regarding diagnoses from the artificial intelligence engine are accurate for input comprising notes of patients.
[0934] Clause 146. The method of any preceding clause, wherein the cognified data includes a conclusion that is identified based on a logic structure representing codified evidence based guidelines pertaining to the medical condition.
[0935] Clause 147. The method of any preceding clause, further comprising processing the notes to obtain indicia by inputting the notes into an artificial intelligence engine trained to identify the indicia in text based on commonly used indicia pertaining to the medical condition.
[0936] Clause 148. The method of any preceding clause, wherein generating the diagnosis further comprises: [0937] determining a stage of the medical condition based on the cognified data; and [0938] including the stage of the medical condition in the diagnosis.
[0939] Clause 149. The method of any preceding clause, further comprising: [0940] determining a severity of the medical condition based on the stage and the type of the medical condition; [0941] in response to the severity satisfying a threshold condition, providing a recommendation to seek immediate medical attention to a computing device of the patient.
[0942] Clause 150. A tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to: [0943] receive, at a server, an electronic medical record comprising notes pertaining to a patient; [0944] generate cognified data using the notes, wherein the cognified data comprises a health summary of the medical condition; [0945] generate, based on the cognified data, a diagnosis of the medical condition of the patient, wherein the diagnosis at least identifies a type of the medical condition; and [0946] provide the diagnosis to a computing device for presentation on the computing device.
[0947] Clause 151. The computer-readable medium of any preceding clause, wherein the processing device is further to identify, in the notes, indicia comprising a phrase, a predicate, a keyword, a cardinal, a number, a concept, or some combination thereof; [0948] Clause 152. The computer-readable medium of any preceding clause, wherein generating the cognified data further comprises detecting the medical condition by identifying a similarity between the indicia and a knowledge graph.
[0949] Clause 153. The computer-readable medium of any preceding clause, wherein the processing device is further to use an artificial intelligence engine that is trained using feedback from medical personnel, wherein the feedback pertains to whether output regarding diagnoses from the artificial intelligence engine are accurate for input comprising notes of patients.
[0950] Clause 154. The computer-readable medium of any preceding clause, wherein the cognified

data includes a conclusion about a predicate in the notes that is identified in a logic structure representing codified evidence based guidelines pertaining to the medical condition.
[0951] Clause 155. The computer-readable medium of any preceding clause, wherein the processing device is further to process the patient notes to obtain indicia by inputting the notes into an artificial intelligence engine trained to identify the indicia in text based on commonly used indicia pertaining to the medical condition.
[0952] Clause 156. The computer-readable medium of any preceding clause, wherein generating the diagnosis further comprises: [0953] determining a stage of the medical condition based on the cognified data; and [0954] including the stage of the medical condition in the diagnosis.
[0955] Clause 157. The computer-readable medium of any preceding clause, wherein the processing device is further to: [0956] determine a severity of the medical condition based on the stage and the type of the medical condition; [0957] in response to the severity satisfying a threshold condition, provide a recommendation to seek immediate medical attention to a computing device of the patient.
[0958] Clause 158. A system, comprising: [0959] a memory device storing instructions; and [0960] a processing device communicatively coupled to the memory device, the processing device executes the instructions to: [0961] receive, at a server, an electronic medical record comprising notes pertaining to a patient; [0962] generate cognified data using the notes, wherein the cognified data comprises a health summary of the medical condition; [0963] generate, based on the cognified data, a diagnosis of the medical condition of the patient, wherein the diagnosis at least identifies a type of the medical condition; and [0964] provide the diagnosis to a computing device for presentation on the computing device.
[0965] Clause 159. The system of any preceding clause, wherein the processing device is further to identify, in the notes, indicia comprising a phrase, a predicate, a keyword, a cardinal, a number, a concept, or some combination thereof;
[0966] Clause 160. The system of any preceding clause, wherein generating the cognified data further comprises detecting the medical condition by identifying a similarity between the indicia and a knowledge graph.
[0967] Clause 161. The system of any preceding clause, wherein the processing device is further to use an artificial intelligence engine that is trained using feedback from medical personnel, wherein the feedback pertains to whether output regarding diagnoses from the artificial intelligence engine are accurate for input comprising notes of patients.
[0968] Clause 162. A method for a processing device executing an autonomous multipurpose application, comprising: [0969] obtaining a plurality of schedules for people having a specialty; [0970] determining whether a user has elected to enable electronic scheduling; and [0971] responsive to determining the user has elected to enable electronic scheduling: [0972] determining which person of the plurality of people has an available appointment based on the plurality of schedules; [0973] transmitting a request to book the available appointment for the person to provide a service to the user; [0974] receiving a response indicating the available appointment is booked as a booked appointment between the person and the user; and [0975] providing a notification pertaining to the booked appointment.
[0976] Clause 163. The method of any preceding clause, further comprising: [0977] obtaining an image of an insurance card of the user; [0978] processing the image to extract information pertaining to an insurance plan of the user; [0979] determining, based on the insurance plan, an expected payment that the user will pay for the service in view of a deductible specified in the insurance plan.
[0980] Clause 164. The method of any preceding clause, further comprising: [0981] determining, without considering the insurance plan, a self-pay cost the user is expected to pay for the service; [0982] selecting to pay using the insurance plan of the user when the expected payment is less than the self-pay cost; and [0983] selecting to pay without using the insurance plan of the user when the

self-pay cost is less than the expected payment.
[0984] Clause 165. The method of any preceding clause, wherein determining which person of the plurality of people has the available appointment is further based on the available appointment having a future date and time that is closest to a current date and time the request was received.
[0985] Clause 166. The method of any preceding clause, further comprising: [0986] determining an expected payment that the user will pay for the service in view of a deductible specified in an insurance plan of the user; [0987] determining, without considering the insurance plan, a self-pay cost the user is expected to pay for the treatment without using the insurance plan; [0988] causing the expected payment, the co-pay cost, or some combination thereof to be presented on a computing device of the user, a computing device of an administrator, a computing device of the person, or some combination thereof.
[0989] Clause 167. The method of any preceding clause, wherein the processing device executing the autonomous multipurpose application obtains the plurality of schedules for the plurality of people having the specialty from at least an electronic medical record system.
[0990] Clause 168. The method of any preceding clause, wherein the obtaining the plurality of schedules for the plurality of people having the specialty further comprises obtaining the plurality of schedules for the plurality of people within a geographic radius of a location of the user.
[0991] Clause 169. The method of any preceding clause, further comprising: [0992] providing the notification pertaining to the booked appointment to a computing device of the user, a computing device of an administrator, a computing device of the person, or some combination thereof.
[0993] Clause 170 The method of any preceding clause, further comprising: [0994] receiving a selection of the specialty from a plurality of specialties comprising at least two of a dentist, a medical doctor, an optometrist, a behavioral psychologist, a chiropractor, and a physician assistant.
[0995] Clause 171. The method of any preceding clause, further comprising: [0996] determining, using a machine learning model, an estimated wait time based on an average amount of time it takes people having the specialty to perform the service for users; and [0997] providing the estimated wait time to the computing device of the user for presentation on a user interface of the computing device of the user.
[0998] Clause 172. The method of any preceding clause, further comprising: [0999] responsive to determining the user has not elected to enable electronic scheduling: determining which person of the plurality of people has an available appointment based on the plurality of schedules; and [1000] providing a notification pertaining to the person having the available appointment to a computing device of the user, wherein the notification comprises a recommended date and time for the available appointment.
[1001] Clause 173. The method of any preceding clause, further comprising: [1002] determining which documents the user has to complete for the booked appointment; [1003] determining whether the user has completed the documents; [1004] responsive to determining the user has completed the documents, providing one or more references pertaining to a condition associated with the service; and [1005] responsive to determining the user has not completed the documents: [1006] electronically filling in fields with any information the user has already provided for the documents; and [1007] causing the documents with the electronically filled in fields to be presented on a computing device of the user for further completion.
[1008] Clause 174. The method of any preceding clause, further comprising: [1009] providing, to a computing device of the user, curated content tailored to a condition associated with the service the person is going to provide to the user during the booked appointment.
[1010] Clause 175. The method of any preceding clause, further comprising: [1011] responsive to determining the user has not elected to enable electronic scheduling: [1012] determining which people of the plurality of people have available appointments based on the plurality of schedules; and recommending the available appointments to a computing device of the user, wherein at least two of the people associated with the available appointments provide the service at different

locations.
[1013] Clause 176. A tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to execute an autonomous multipurpose application to: [1014] receive an appointment request for a person to provide a service to a user; [1015] determine an expected payment that the user will pay for the service in view of a deductible specified in an insurance plan associated with the user; [1016] determine, without considering the insurance plan, a self-pay cost the user is expected to pay for the service; and [1017] cause the expected payment, the self-pay cost, or some combination thereof to be presented on a computing device of the user, a computing device of an administrator, a computing device of the person, or some combination thereof.
[1018] Clause 177. The computer-readable medium of any preceding clause, wherein the processing device is further to receive a selection to schedule the appointment based on the co-pay cost or the self-pay cost.
[1019] Clause 178. The computer-readable medium of any preceding clause, wherein the user is a dependent of a primary holder of the insurance plan.
[1020] Clause 179. The computer-readable medium of any preceding clause, wherein the processing device is further to: [1021] obtain a schedule for the person; [1022] determine whether the user has elected to enable electronic scheduling; and [1023] responsive to determining the user has elected to enable electronic scheduling: determine an available appointment based on the schedule; [1024] transmit a request to book the available appointment for the person to provide the service to the user; [1025] receive a response indicating the available appointment is booked as a booked appointment between the person and the user; and [1026] provide a notification pertaining to the booked appointment.
[1027] Clause 180. The computer-readable medium of any preceding clause, wherein the processing device is further to: [1028] check-in a user for the appointment; [1029] determine, using a machine learning model, an estimated wait time based on an average amount of time it takes people to perform the service for users; and [1030] provide the estimated wait time to the computing device of the user for presentation on a user interface of the computing device of the user.
[1031] Clause 181. A system, comprising: [1032] a memory storing instructions that implement an autonomous multipurpose application; and [1033] a processing device communicatively coupled to the memory, the processing device capable of executing the autonomous multipurpose application to: [1034] check-in a user for a scheduled appointment with a person having a specialty to perform a service; [1035] determine, using a machine learning model, an estimated wait time based on an average amount of time it takes people having the specialty to perform the service for users; and [1036] provide the estimated wait time to a computing device of the user for presentation on a user interface of the computing device of the user.
[1037] Clause 182. The system of any preceding clause, wherein the processing device is further to: [1038] determine which documents the user is required to complete for the scheduled appointment; [1039] determine whether the user has completed the documents; [1040] responsive to determining the user has completed the documents, provide curated content pertaining to a condition associated with the service; and [1041] responsive to determining the user has not completed the documents: [1042] electronically fill fields with any information the user has already provided for the documents; and [1043] cause the documents with the electronically filled in fields to be presented on a computing device of the user for further completion of fields missing information.
[1044] Clause 183. The system of any preceding clause, wherein the processing device is further to provide curated content tailored for the user based on the service, the specialty, a condition pertaining to the service, other conditions associated with the user, or some combination thereof.
[1045] Clause 184. The system of any preceding clause, wherein the processing device is further to

maintain documents for the user and a dependent of the user and provide the documents to any requesting client device, wherein the documents comprise medical history, consent forms, medical records, or some combination thereof, and the requesting client device comprises an electronic medical record system.
[1046] Clause 185. The system of any preceding clause, wherein, prior to the scheduled appointment being scheduled, the processing device is further to: [1047] obtain a schedule for the person; [1048] determine whether the user has elected to enable electronic scheduling; and [1049] responsive to determining the user has elected to enable electronic scheduling: determine an available appointment based on the schedule; [1050] transmit a request to schedule the available appointment for the person to provide the service to the user; [1051] receive a response indicating the available appointment is booked as the scheduled appointment between the person and the user; and [1052] provide a notification pertaining to the scheduled appointment.
[1053] Clause 186. The system of any preceding clause, wherein, prior to the scheduled appointment being scheduled, the processing device is further to: [1054] receive an appointment request for the person to provide a service to the user; [1055] determine an expected payment that the user will pay for the service in view of a deductible specified in an insurance plan associated with the user; [1056] determine, without considering insurance plan, a self-pay cost the user is expected to pay for the service; [1057] causing the expected payment, the self-pay pocket cost, or some combination thereof to be presented on a computing device of the user, a computing device of an administrator, a computing device of the person, or some combination thereof.
[1058] Clause 187. A method, comprising: [1059] maintaining a set of check-in documents for a user; [1060] receiving, from a computing device, a plurality of requests to check-in the user for a plurality of scheduled appointments where a plurality of people each having a different respective specialty of a plurality of specialties are to provide a different respective service to the user; [1061] determining respective subsets of the set of check-in documents that are required to be complete for each of the different respective specialty of each of the plurality of people; and [1062] transmitting each of the respective subsets of the set of check-in documents to a plurality of computing devices each associated with each of the different respective specialty, wherein the respective subsets are cryptographically signed.
[1063] Clause 188. The method of any preceding clause, further comprising, for each of the plurality of scheduled appointments, determining whether check-in requirements are satisfied, wherein the check-in requirements are satisfied when required information in each of the respective subsets of the set of check-in documents has already been provided.
[1064] Clause 189. The method of any preceding clause, further comprising: [1065] responsive to determining the check-in requirements for one of the plurality of scheduled appointments is not satisfied because one of the respective subsets of the set of check-in documents is lacking a portion of the required information, cause the computing device to present a notification that the portion of the required information is lacking; [1066] receiving the portion of the required information; [1067] updating the one of the respective subsets of the set of check-in documents with the portion of the required information; and [1068] checking-in the user for the one of the plurality of scheduled appointments.
[1069] Clause 190. The method of any preceding clause, further comprising: [1070] responsive to determining the check-in requirements for one of the plurality of scheduled appointments is satisfied, checking-in the user for the one of the scheduled appointments.
[1071] Clause 191. The method of any preceding clause, further comprising: [1072] updating the set of check-in documents based on input from the user, input from the plurality of people, output from a machine learning model trained to determine when certain information needs to be updated, information obtained from a third-party source, or some combination thereof.
[1073] Clause 192. The method of any preceding clause, wherein the plurality of specialties comprises a medical doctor, a dentist, an optometrist, a physician's assistant, a chiropractor, an

orthodontist, a behavioral specialist, a therapist, a masseuse, a physical therapist, or some combination thereof.
[1074] Clause 193. The method of any preceding clause, wherein the plurality of requests are received over a period of time and each of the plurality of scheduled appointments are scheduled at different dates, times, or both.
[1075] Clause 194. The method of any preceding clause, further comprising: [1076] checking-in the user for one of the plurality of scheduled appointments with a person having a specialty to provide a service to the user; [1077] determining, using a machine learning model, an estimated wait time based on an average amount of time it takes people having the specialty to perform the service for users; and [1078] providing the estimated wait time to a computing device of the user for presentation on a user interface of the computing device of the user.
[1079] Clause 195. The method of any preceding clause, further comprising, prior to scheduling one of the plurality of scheduled appointments: [1080] obtaining a schedule for a person having the different respective specialty associated with the one of the plurality of scheduled appointments; [1081] determining whether the user has elected to enable electronic scheduling; and [1082] responsive to determining the user has elected to enable electronic scheduling: [1083] determining an available appointment based on the schedule; [1084] transmitting a request to book the one of the plurality of scheduled appointments for the person to provide the different respective service to the user; [1085] receiving a response indicating the one of the plurality of scheduled appointments is booked as a booked appointment between the person and the user; and [1086] providing a notification pertaining to the booked appointment.
[1087] Clause 196. The method of any preceding clause, further comprising, prior to scheduling one of the plurality of scheduled appointments: [1088] receiving an appointment request for a person to provide a service to the user; [1089] determine an expected payment that the user will pay for the service in view of a deductible specified in an insurance plan associated with the user; [1090] determining, without considering the insurance plan, a self-pay cost the user is expected to pay for the service; and [1091] causing the expected payment, the self-pay cost, or some combination thereof to be presented on a computing device of the user, a computing device of an administrator, a computing device of the person, or some combination thereof.
[1092] Clause 197. A method for electronically generating a care plan, the method comprising: [1093] selecting a first data structure corresponding to a first condition of a patient, wherein the first data structure comprises a set of health artifacts pertaining to the first condition; [1094] comparing a second data structure with the first data structure, wherein the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of health artifacts; [1095] selecting, based on the comparing, another subset of the set of health artifacts in the first data structure; [1096] generating the care plan comprising a third data structure that includes at least the another subset of the set of health artifacts; and [1097] causing the care plan to be presented on a computing device.
[1098] Clause 198. The method of any preceding clause, further comprising generating natural language representing the another subset of the set of health artifacts included in the third data structure, and causing the natural language to be presented on the computing device.
[1099] Clause 199. The method of any preceding clause, further comprising: [1100] receiving information corresponding to a health artifact of the set of health artifacts in the first data structure; [1101] determining, based on the information, that the patient has interacted with the health artifact; [1102] updating the second data structure to include the health artifact with the information; and [1103] causing an indication to be presented on the computing device, wherein the indication indicates that the patient has interacted with the health artifact.
[1104] Clause 200. The method of any preceding clause, further comprising: [1105] receiving the information from a source comprising an electronic medical records system, an application programming interface, a claims system, an electronic health virtual assistant, an application

executing on the computing device, a data store, or some combination thereof, [1106] generating an engagement profile for the patient using the health artifact with the information; and [1107] updating the second data structure with the engagement profile for the patient.
[1108] Clause 201. The method of any preceding clause, further comprising determining a value of patient compliance with the care plan based on tracked interactions of the patient and the another subset of the set of health artifacts.
[1109] Clause 202. The method of any preceding clause, wherein the tracked interactions comprise: [1110] activity of the patient using the computing device, wherein the activity comprises: [1111] a selection using an input peripheral of the computing device, [1112] an amount of time the patient actively uses an application, [1113] an amount of time the patient spends viewing a particular user interface, [1114] a search query entered by the patient, or some combination thereof; [1115] an indication from an external system that the patient has interacted with the health artifact of the another subset of the set of health artifacts; or [1116] some combination thereof.
[1117] Clause 203. The method of any preceding clause, further comprising including, in the care plan, an action instruction pertaining to the another subset of the set of health artifacts, wherein the action instructions are directed toward a medical personnel, the patient, or both.
[1118] Clause 204. The method of any preceding clause, further comprising: [1119] selecting a fourth data structure corresponding to a second condition of the patient, wherein the fourth data structure comprises a second set of health artifacts pertaining to the second condition, and the first condition and the second condition are different; [1120] comparing a fifth data structure with the fourth data structure, wherein the fifth data structure pertains to the patient and the second condition of the patient, and the fifth data structure comprises a second subset of the second set of health artifacts; [1121] selecting, based on the comparing, a third subset of the set of health artifacts in the fourth data structure; [1122] generating the care plan comprising the third data structure that includes at least the another subset of the set of health artifacts and the third subset of the set of health artifacts.
[1123] Clause 205. The method of any preceding clause, further comprising: [1124] receiving input from the computing device, wherein the input specifies a number and an area of the condition the patient selects to manage, wherein the area comprises a type of health artifacts in the set of the health artifacts; and [1125] selecting, based on the comparing, the another subset of the set of health artifacts in the first data structure by selecting the another subset based on the number and the type of health artifacts specified by the patient.
[1126] Clause 206. The method of any preceding clause, wherein the subset of the set of health artifacts correspond with interactions already performed by the patient, and the another subset of the set of health artifacts correspond with interactions that have not yet been performed by the patient.
[1127] Clause 207. The method of any preceding clause, wherein the comparing further comprises projecting the second data structure onto the first data structure.
[1128] Clause 208. The method of any preceding clause, wherein the first data structure is a knowledge graph, and the second data structure is a patient graph.
[1129] Clause 209. A tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to: [1130] select a first data structure corresponding to a first condition of a patient, wherein the first data structure comprises a set of health artifacts pertaining to the first condition; [1131] compare a second data structure with the first data structure, wherein the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of health artifacts; [1132] select, based on the comparing, another subset of the set of health artifacts in the first data structure; [1133] generate the care plan comprising a third data structure that includes at least the another subset of the set of health artifacts; and [1134] cause the care plan to be presented on a computing device.

[1135] Clause 210. The computer-readable medium of any preceding claim, wherein the processing device is further to generate natural language representing the another subset of the set of health artifacts included in the third data structure, and causing the natural language to be presented on the computing device.
[1136] Clause 211. The computer-readable medium of any preceding claim, wherein the processing device is further to: [1137] receive information corresponding to a health artifact of the set of health artifacts in the first data structure; [1138] determine, based on the information, that the patient has interacted with the health artifact; [1139] update the second data structure to include the health artifact with the information; and [1140] cause an indication to be presented on the computing device, wherein the indication indicates that the patient has interacted with the health artifact.
[1141] Clause 212. The computer-readable medium of any preceding claim, wherein the processing device is further to: [1142] receive the information from a source comprising an electronic medical records system, an application programming interface, a claims system, an electronic health virtual assistant, an application executing on the computing device, a data store, or some combination thereof, [1143] generate an engagement profile for the patient using the health artifact with the information; and [1144] update the second data structure with the engagement profile for the patient.
[1145] Clause 213. The computer-readable medium of any preceding claim, wherein the processing device is further to determine a value of patient compliance with the care plan based on tracked interactions of the patient and the another subset of the set of health artifacts.
[1146] Clause 214. The computer-readable medium of any preceding claim, wherein the tracked interactions comprise: [1147] activity of the patient using the computing device, wherein the activity comprises: [1148] a selection using an input peripheral of the computing device, [1149] an amount of time the patient actively uses an application, [1150] an amount of time the patient spends viewing a particular user interface, [1151] a search query entered by the patient, or some combination thereof; [1152] an indication from an external system that the patient has interacted with the health artifact of the another subset of the set of health artifacts; or [1153] some combination thereof.
[1154] Clause 215. The computer-readable medium of any preceding claim, wherein: [1155] the subset of the set of health artifacts correspond with actions already performed by the patient, and the another subset of the set of health artifacts correspond with actions that have not yet been performed by the patient; [1156] the comparing further comprises projecting the second data structure onto the first data structure; and [1157] the first data structure is a knowledge graph, and the second data structure is a patient graph.
[1158] Clause 216. A system comprising: [1159] a memory storing instructions; [1160] a processing device communicatively coupled to the memory, the processing device capable of executing the instructions to: [1161] select a first data structure corresponding to a first condition of a patient, wherein the first data structure comprises a set of health artifacts pertaining to the first condition; [1162] compare a second data structure with the first data structure, wherein the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of health artifacts; [1163] select, based on the comparing, another subset of the set of health artifacts in the first data structure; [1164] generate the care plan comprising a third data structure that includes at least the another subset of the set of health artifacts; and [1165] cause the care plan to be presented on a computing device.
[1166] Clause 217. A method for electronically generating a care plan, the method comprising: [1167] comparing a first data structure with a second data structure, wherein the first data structure comprises a set of health artifacts pertaining to a first condition of the patient, and the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of the health artifacts; [1168] responsive to the comparing, generating

the care plan comprising another subset of the set of health artifacts; and [1169] modifying the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.
[1170] Clause 218. The method of any preceding clause, further comprising modifying the another subset of the set of the health artifacts in real-time or near real-time.
[1171] Clause 219. The method of any preceding clause, further comprising detecting the detected tone of the patient based on words spoken by the patient, text entered by the patient, or some combination thereof.
[1172] Clause 220. The method of any preceding clause, further comprising detecting the detected emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.
[1173] Clause 221. The method of any preceding clause, further comprising causing the care plan including modifications to the another subset of the set of the health artifacts to be presented on a computing device.
[1174] Clause 222. The method of any preceding clause, further comprising modifying the another subset of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health artifact that corresponds to the medical outcome and to include the health artifact in the another subset of the set of the health artifacts.
[1175] Clause 223. The method of any preceding clause, further comprising: [1176] generating a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan; and [1177] updating a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition.
[1178] Clause 224. The method of any preceding clause, further comprising including, in the care plan, action instructions pertaining to the another subset of the set of the health artifacts, wherein the action instructions are directed toward a medical personnel, the patient, or both.
[1179] Clause 225. The method of any preceding clause, further comprising: [1180] receiving input from a computing device, wherein the input specifies a number and a type of health artifacts in the set of the health artifacts the patient selects to manage; and [1181] selecting, based on the comparing, the another subset of the set of the health artifacts in the first data structure by selecting the another subset based on the number and the type of health artifacts specified by the patient.
[1182] Clause 226. The method of any preceding clause, wherein: [1183] the subset of the set of the health artifacts correspond with actions already performed by the patient, and the another subset of the set of the health artifacts correspond with actions that have not yet been performed by the patient; [1184] the comparing further comprises projecting the second data structure onto the first data structure; and [1185] the first data structure is a knowledge graph, and the second data structure is a patient graph.
[1186] Clause 227. A tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to: [1187] compare a first data structure with a second data structure, wherein the first data structure comprises a set of health artifacts pertaining to a first condition of the patient, and the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of the health artifacts; [1188] responsive to the comparing, generate a care plan comprising another subset of the set of health artifacts; and [1189] modify the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.
[1190] Clause 228. The computer-readable medium of any preceding clause, wherein the

processing device is further to modify the another subset of the set of the health artifacts in real-time or near real-time.
[1191] Clause 229. The computer-readable medium of any preceding clause, wherein the processing device is further to detect the detected tone of the patient based on words spoken by the patient, text entered by the patient, or some combination thereof.
[1192] Clause 230. The computer-readable medium of any preceding clause, wherein the processing device is further to detect the detected emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.
[1193] Clause 231. The computer-readable medium of any preceding clause, wherein the processing device is further to cause the care plan including modifications to the another subset of the set of the health artifacts to be presented on a computing device.
[1194] Clause 232. The computer-readable medium of any preceding clause, wherein the processing device is further to modify the another subset of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health artifact that corresponds to the medical outcome and to include the health artifact in the another subset of the set of the health artifacts.
[1195] Clause 233. The computer-readable medium of any preceding clause, wherein the processing device is further to: [1196] generate a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan; and [1197] update a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition.
[1198] Clause 234. A system comprising: [1199] a memory storing instructions; [1200] a processing device communicatively coupled to the memory, the processing device capable of executing the instructions to: [1201] compare a first data structure with a second data structure, wherein the first data structure comprises a set of health artifacts pertaining to a first condition of the patient, and the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of the health artifacts; [1202] responsive to the comparing, generate a care plan comprising another subset of the set of health artifacts; and [1203] modify the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.
[1204] Clause 235, The system of any preceding clause, wherein the processing device is further to: [1205] generate a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan; and [1206] update a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition.
[1207] Clause 236. The system of any preceding clause, wherein the processing device is further to: [1208] modify the another subset of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health artifact that corresponds to the medical outcome and to include the health artifact in the another subset of the set of the health artifacts; [1209] detect the detected tone of the patient based on words spoken by the patient, text entered by the patient, or some combination thereof; and [1210] detect the detected emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.
[1211] The foregoing description, for purposes of explanation, used specific nomenclature to provide a thorough understanding of the described embodiments. However, it should be apparent to

one skilled in the art that the specific details are not required in order to practice the described embodiments. Thus, the foregoing descriptions of specific embodiments are presented for purposes of illustration and description. They are not intended to be exhaustive or to limit the described embodiments to the precise forms disclosed. It should be apparent to one of ordinary skill in the art that many modifications and variations are possible in view of the above teachings.

[1212] The above discussion is meant to be illustrative of the principles and various embodiments of the present invention. Numerous variations and modifications will become apparent to those skilled in the art once the above disclosure is fully appreciated. It is intended that the following claims be interpreted to embrace all such variations and modifications.

Claims

1. A method for electronically generating a care plan, the method comprising: comparing a first data structure with a second data structure, wherein the first data structure comprises a set of health artifacts pertaining to a first condition of the patient, and the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of the health artifacts; responsive to the comparing, generating the care plan comprising another subset of the set of health artifacts; and modifying the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.

2. The method of claim 1, further comprising modifying the another subset of the set of the health artifacts in real-time or near real-time.

3. The method of claim 1, further comprising detecting the detected tone of the patient based on words spoken by the patient, text entered by the patient, or some combination thereof.

4. The method of claim 1, further comprising detecting the detected emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.

5. The method of claim 1, further comprising causing the care plan including modifications to the another subset of the set of the health artifacts to be presented on a computing device.

6. The method of claim 1, further comprising modifying the another subset of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health artifact that corresponds to the medical outcome and to include the health artifact in the another subset of the set of the health artifacts.

7. The method of claim 1, further comprising: generating a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan; and updating a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition.

8. The method of claim 1, further comprising including, in the care plan, action instructions pertaining to the another subset of the set of the health artifacts, wherein the action instructions are directed toward a medical personnel, the patient, or both.

9. The method of claim 1, further comprising: receiving input from a computing device, wherein the input specifies a number and a type of health artifacts in the set of the health artifacts the patient selects to manage; and selecting, based on the comparing, the another subset of the set of the health artifacts in the first data structure by selecting the another subset based on the number and the type of health artifacts specified by the patient.

10. The method of claim 1, wherein: the subset of the set of the health artifacts correspond with actions already performed by the patient, and the another subset of the set of the health artifacts correspond with actions that have not yet been performed by the patient; the comparing further

comprises projecting the second data structure onto the first data structure; and the first data structure is a knowledge graph, and the second data structure is a patient graph.

11. A tangible, non-transitory computer-readable medium storing instructions that, when executed, cause a processing device to: compare a first data structure with a second data structure, wherein the first data structure comprises a set of health artifacts pertaining to a first condition of the patient, and the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of the health artifacts; responsive to the comparing, generate a care plan comprising another subset of the set of health artifacts; and modify the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.

12. The computer-readable medium of claim 11, wherein the processing device is further to modify the another subset of the set of the health artifacts in real-time or near real-time.

13. The computer-readable medium of claim 11, wherein the processing device is further to detect the detected tone of the patient based on words spoken by the patient, text entered by the patient, or some combination thereof.

14. The computer-readable medium of claim 11, wherein the processing device is further to detect the detected emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.

15. The computer-readable medium of claim 11, wherein the processing device is further to cause the care plan including modifications to the another subset of the set of the health artifacts to be presented on a computing device.

16. The computer-readable medium of claim 11, wherein the processing device is further to modify the another subset of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health artifact that corresponds to the medical outcome and to include the health artifact in the another subset of the set of the health artifacts.

17. The computer-readable medium of claim 1, wherein the processing device is further to: generate a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan; and update a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition.

18. A system comprising: A memory storing instructions; A processing device communicatively coupled to the memory, the processing device capable of executing the instructions to: compare a first data structure with a second data structure, wherein the first data structure comprises a set of health artifacts pertaining to a first condition of the patient, and the second data structure pertains to the patient and the first condition of the patient, and the second data structure comprises a subset of the set of the health artifacts; responsive to the comparing, generate a care plan comprising another subset of the set of health artifacts; and modify the another subset of the set of health artifacts in the care plan based on a detected tone of the patient, a detected emotion of the patient, a medical outcome desired by a physician, or some combination thereof.

19. The system of claim 18, wherein the processing device is further to: generate a net promoter score based on the detected tone of the patient, the detected emotion of the patient, or both in response to the patient interacting with the care plan; and update a machine learning model based on the net promoter score being below a threshold value to obtain an updated machine learning model that outputs different health artifacts for subsequent patients having the condition.

20. The system of claim 18, wherein the processing device is further to: modify the another subset of the set of the health artifacts in the care plan based on the medical outcome desired by the physician by receiving instructions from a computing device of a physician to select a health

artifact that corresponds to the medical outcome and to include the health artifact in the another subset of the set of the health artifacts; detect the detected tone of the patient based on words spoken by the patient, text entered by the patient, or some combination thereof; and detect the detected emotion of the patient based on words spoken by the patient, text entered by the patient, a detected facial expression of the patient, or some combination thereof.