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(54) **FACILITATION OF THE DISTRIBUTION OF SCIENTIFIC DATA**

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(51) **Int. Cl.**

G06Q 30/00	(2012.01)
G06Q 30/06	(2012.01)
G06Q 20/06	(2012.01)
G16B 50/30	(2019.01)
G06Q 50/26	(2012.01)
G06Q 20/36	(2012.01)

(52) **U.S. Cl.**

CPC **G06Q 30/0619** (2013.01); **G06Q 20/0658** (2013.01); **G06Q 20/3672** (2013.01); **G06Q 50/26** (2013.01); **G16B 50/30** (2019.02)

(58) **Field of Classification Search**

None
See application file for complete search history.

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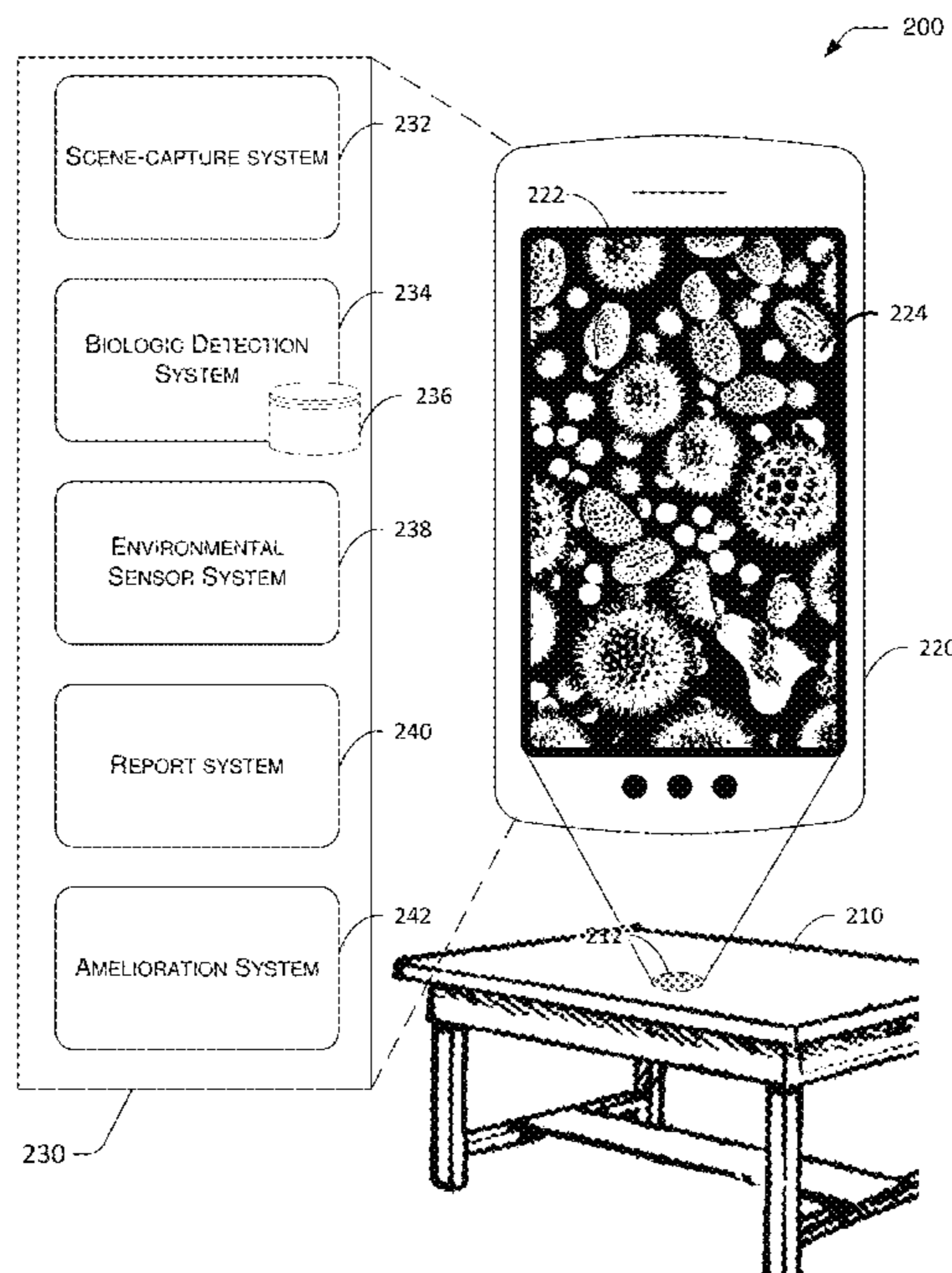
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(57) **ABSTRACT**

A technology that facilitates distribution of scientific data are disclosed. Exemplary implementations may: obtain scientific data from sci-data gathering devices; analyze gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data; offer the analyzed sci-data via a marketplace in exchange for a monetary value; and deliver the analyzed sci-data via the marketplace in exchange for an acceptable monetary value.

17 Claims, 10 Drawing Sheets



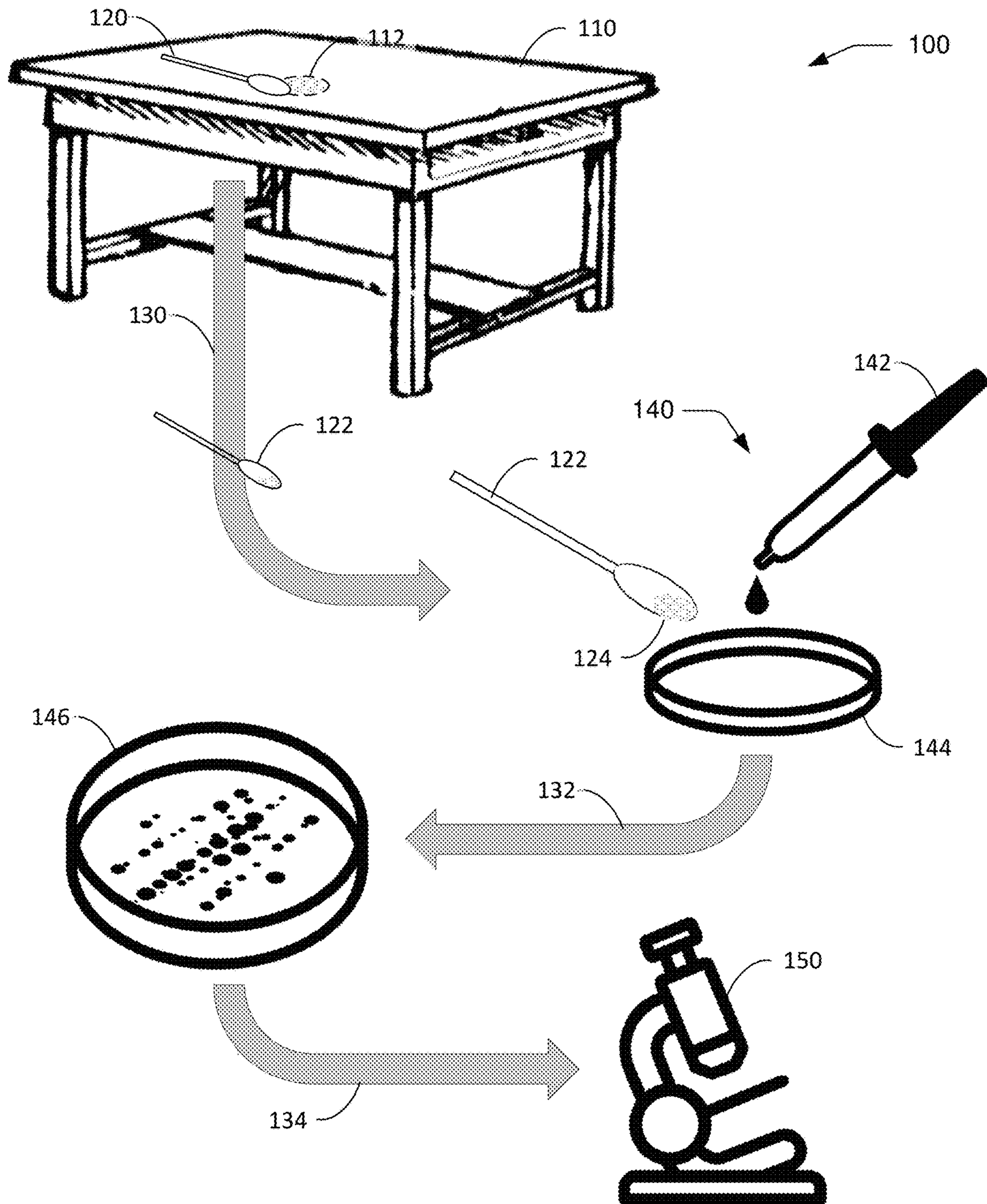


Fig. 1 Background

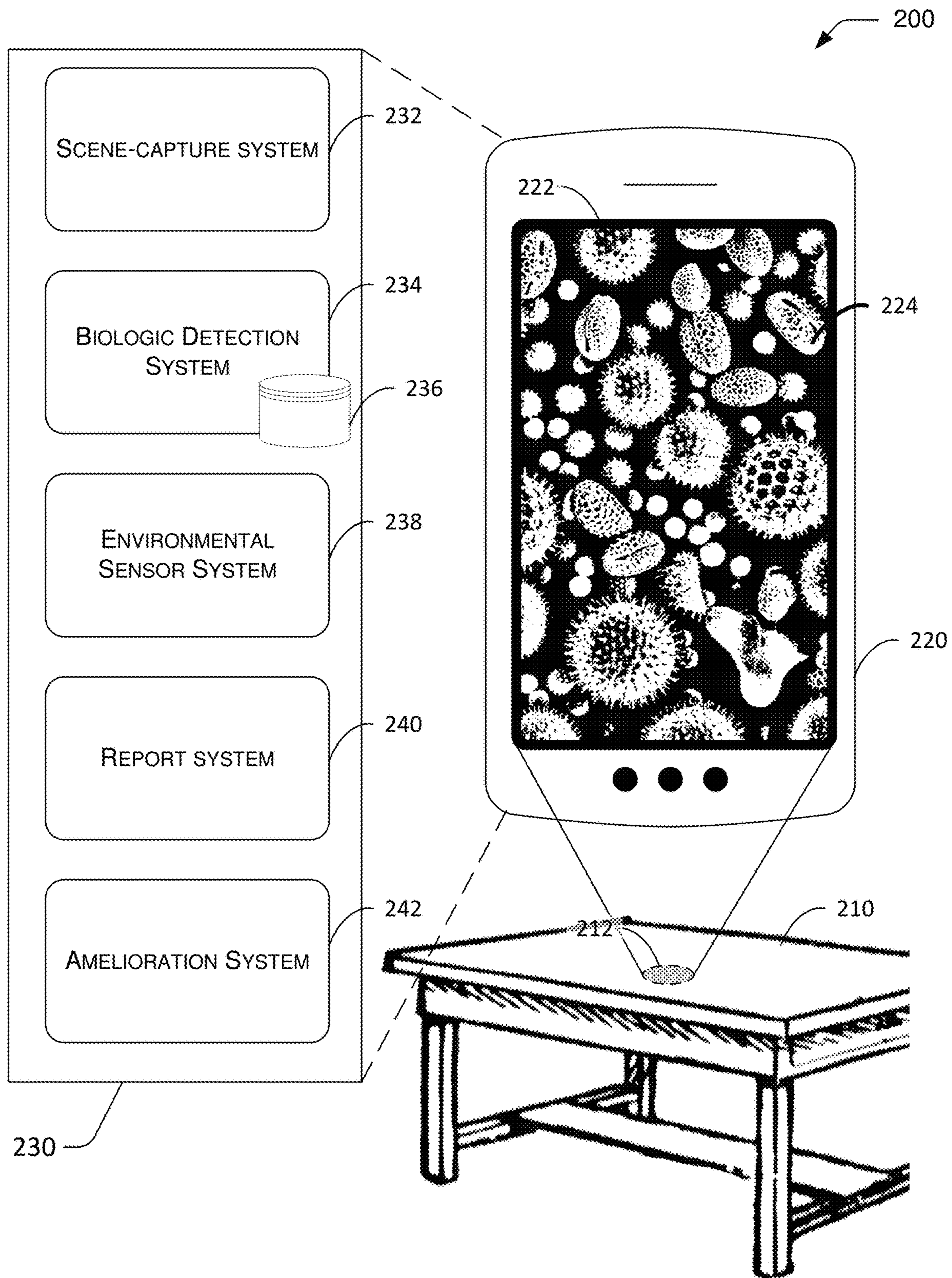


Fig. 2

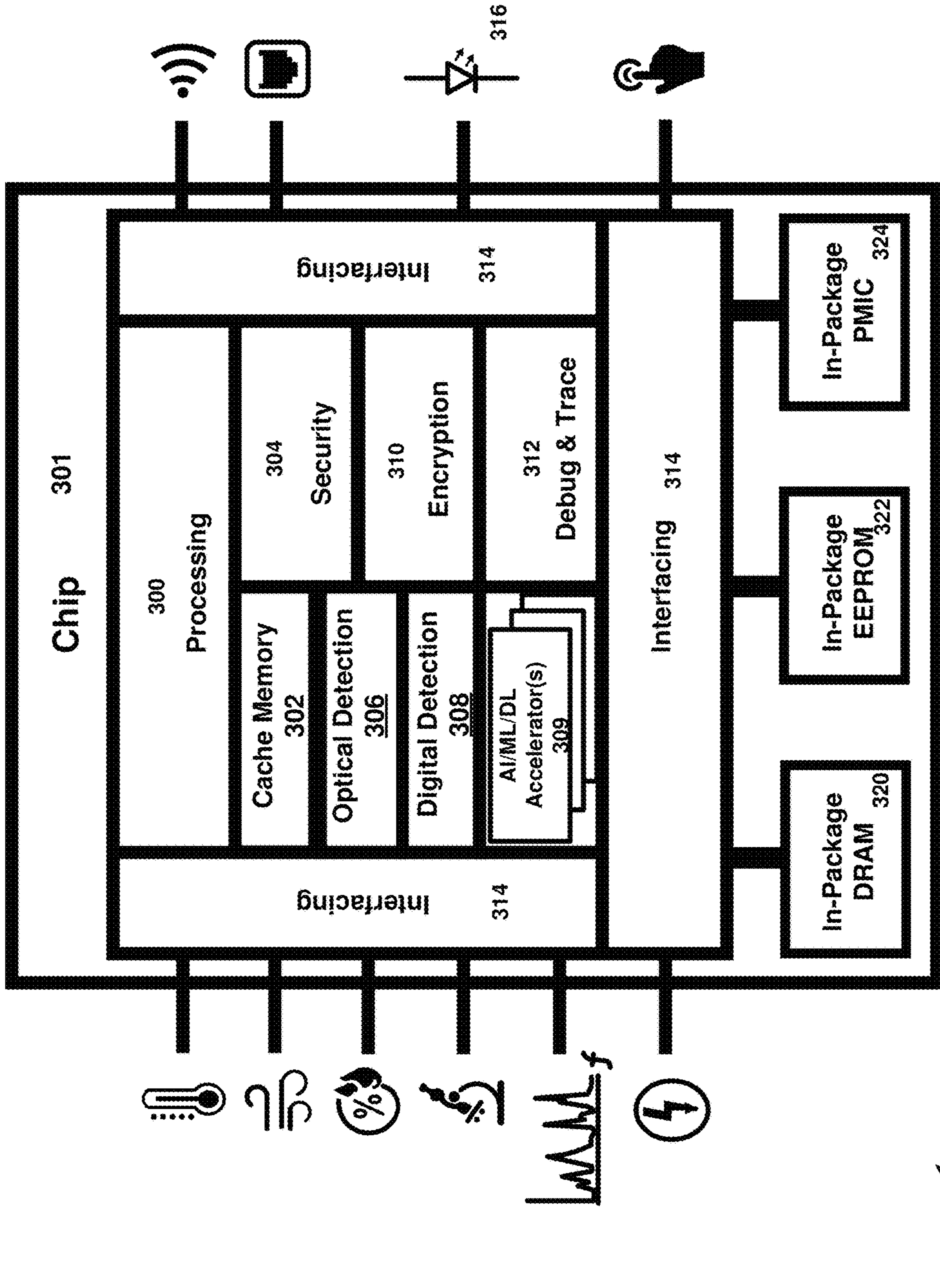


Fig. 3

300

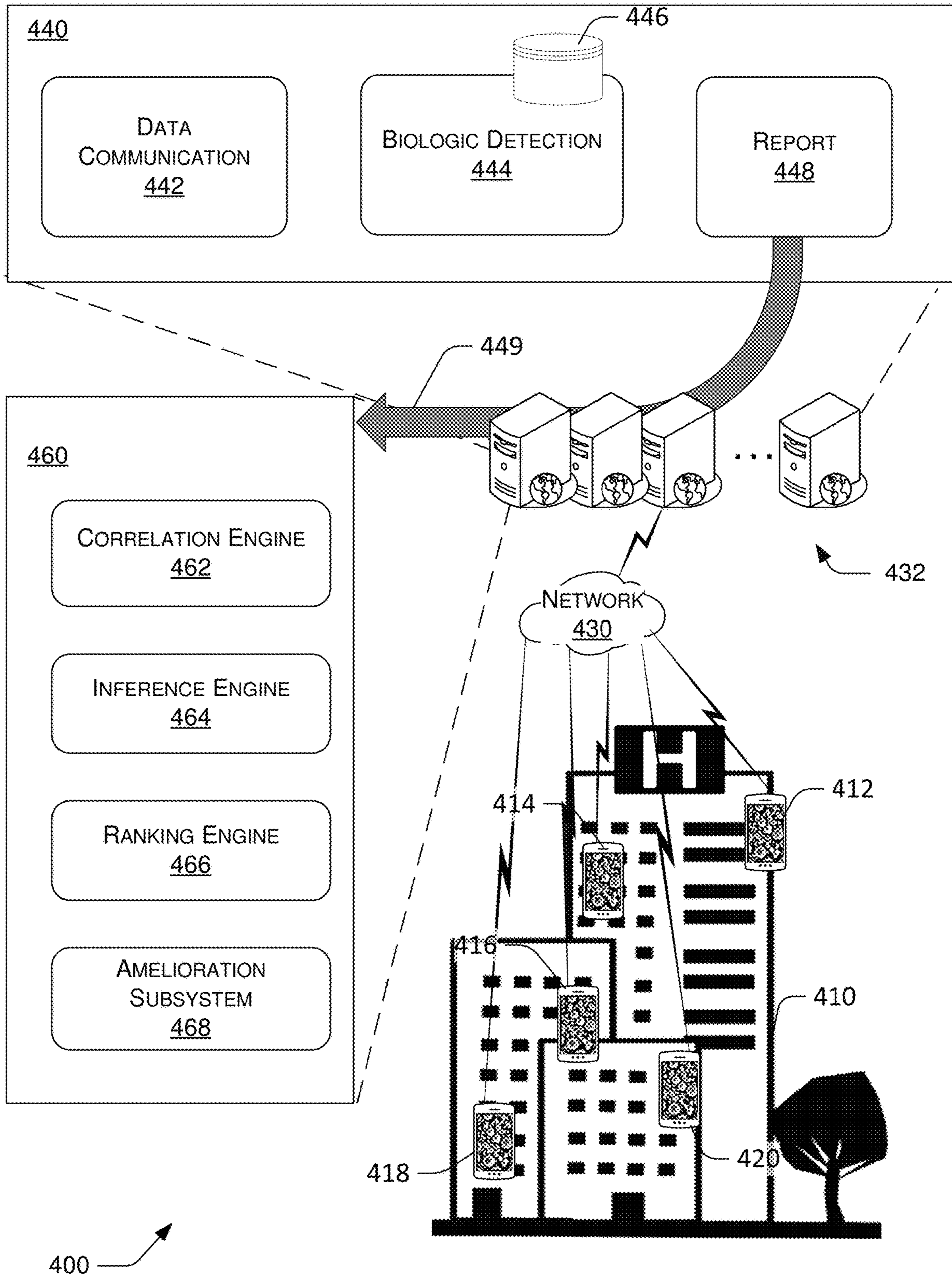


Fig. 4

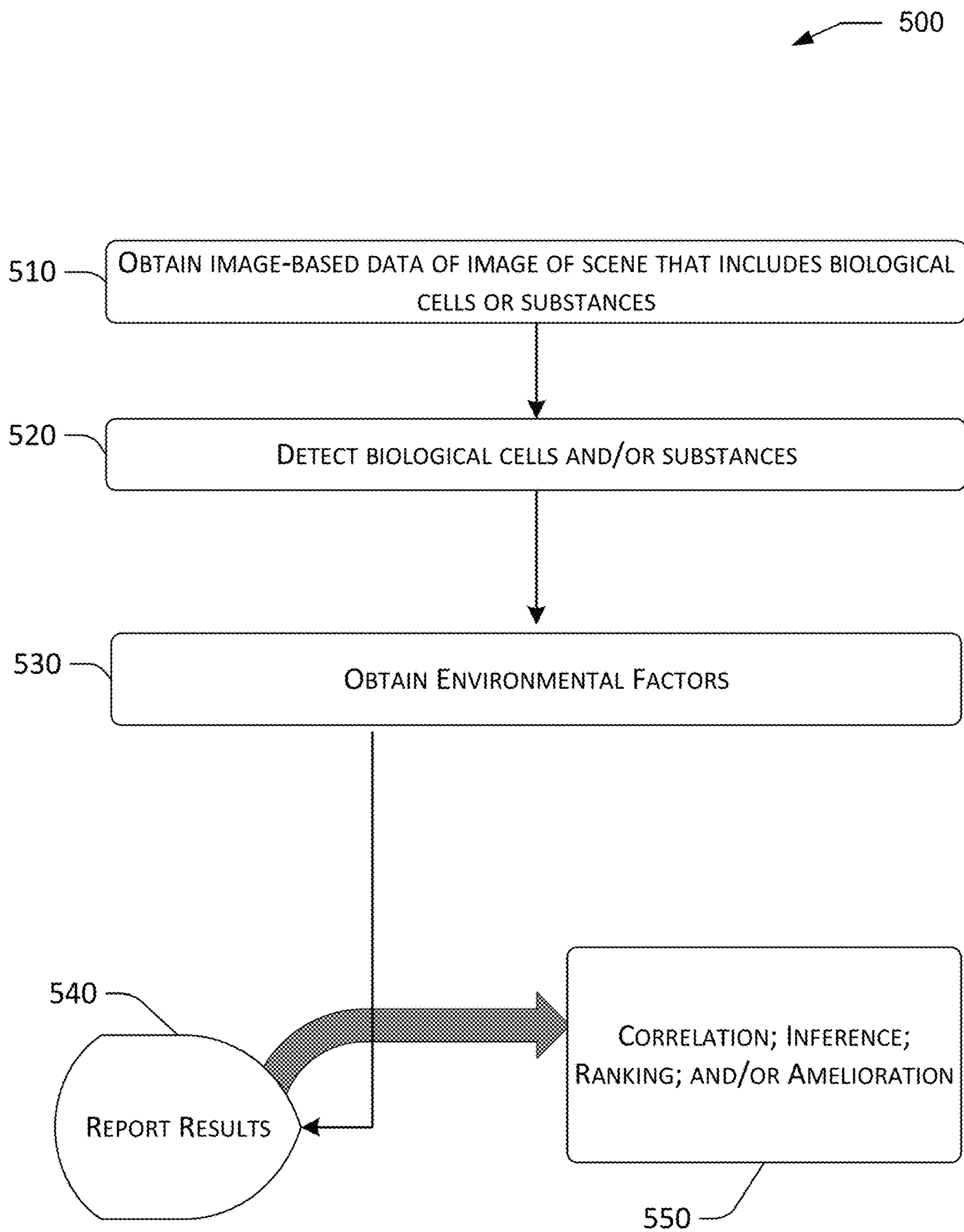


Fig. 5

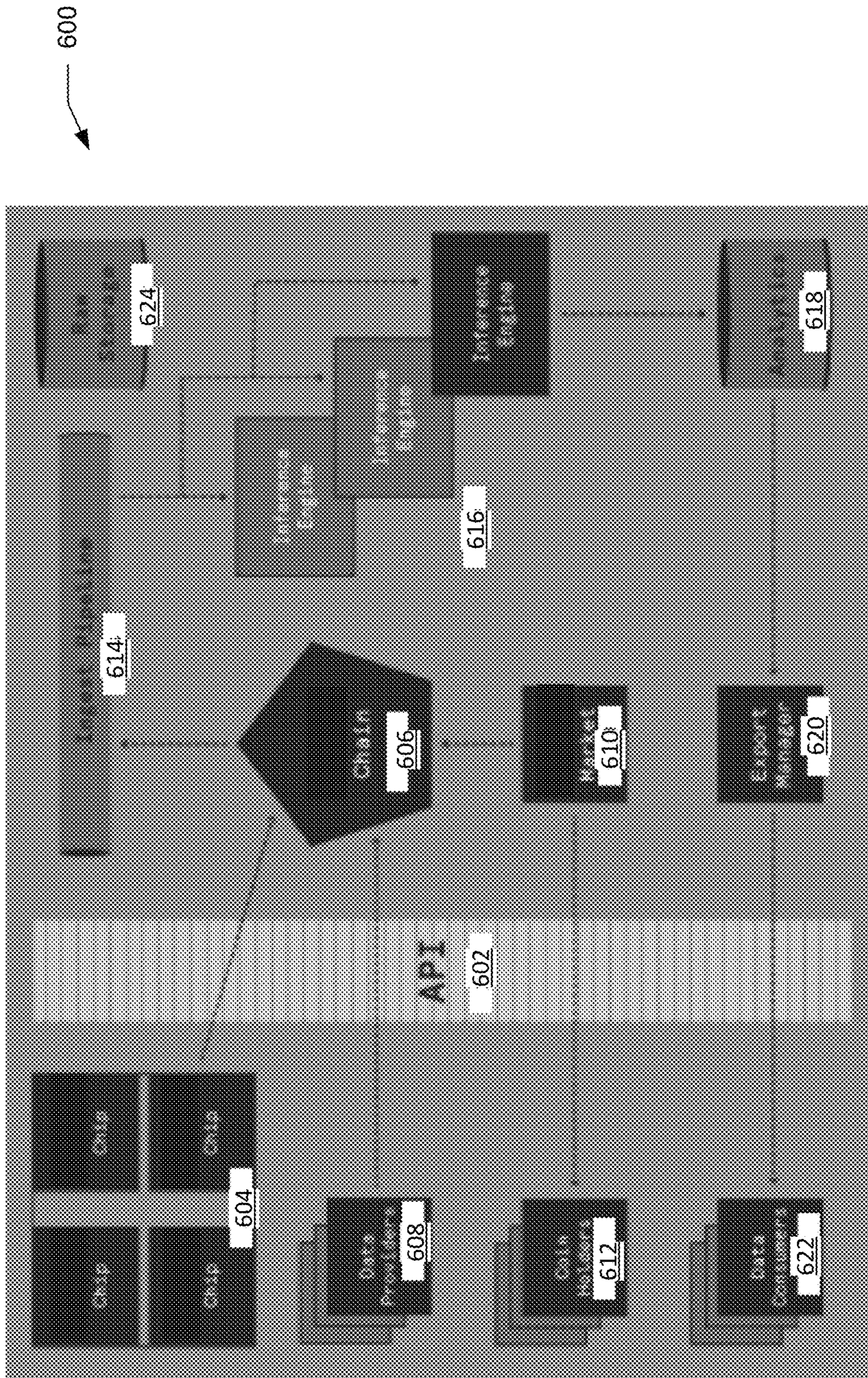


Fig. 6

700

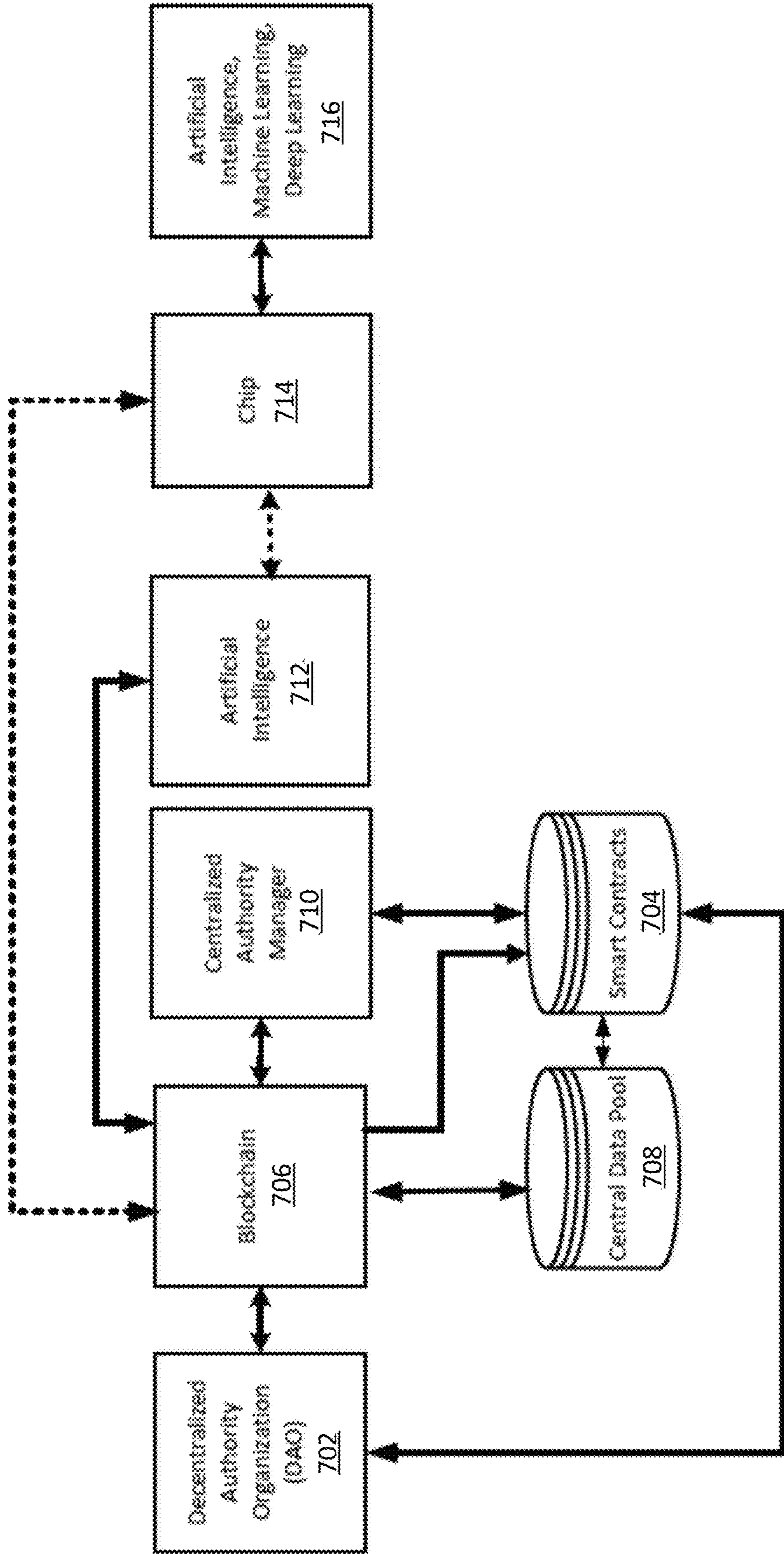


Fig. 7

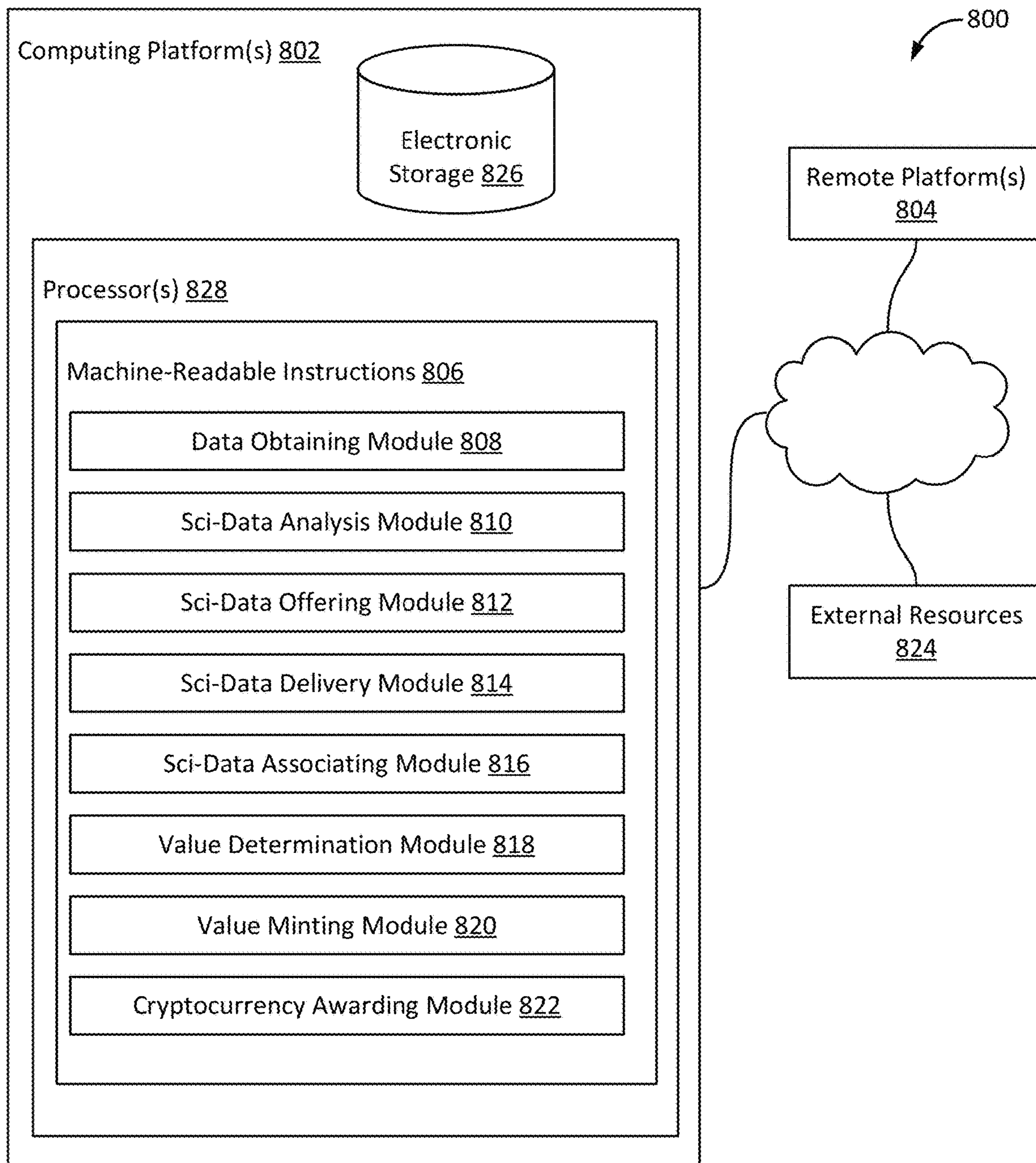


Fig. 8

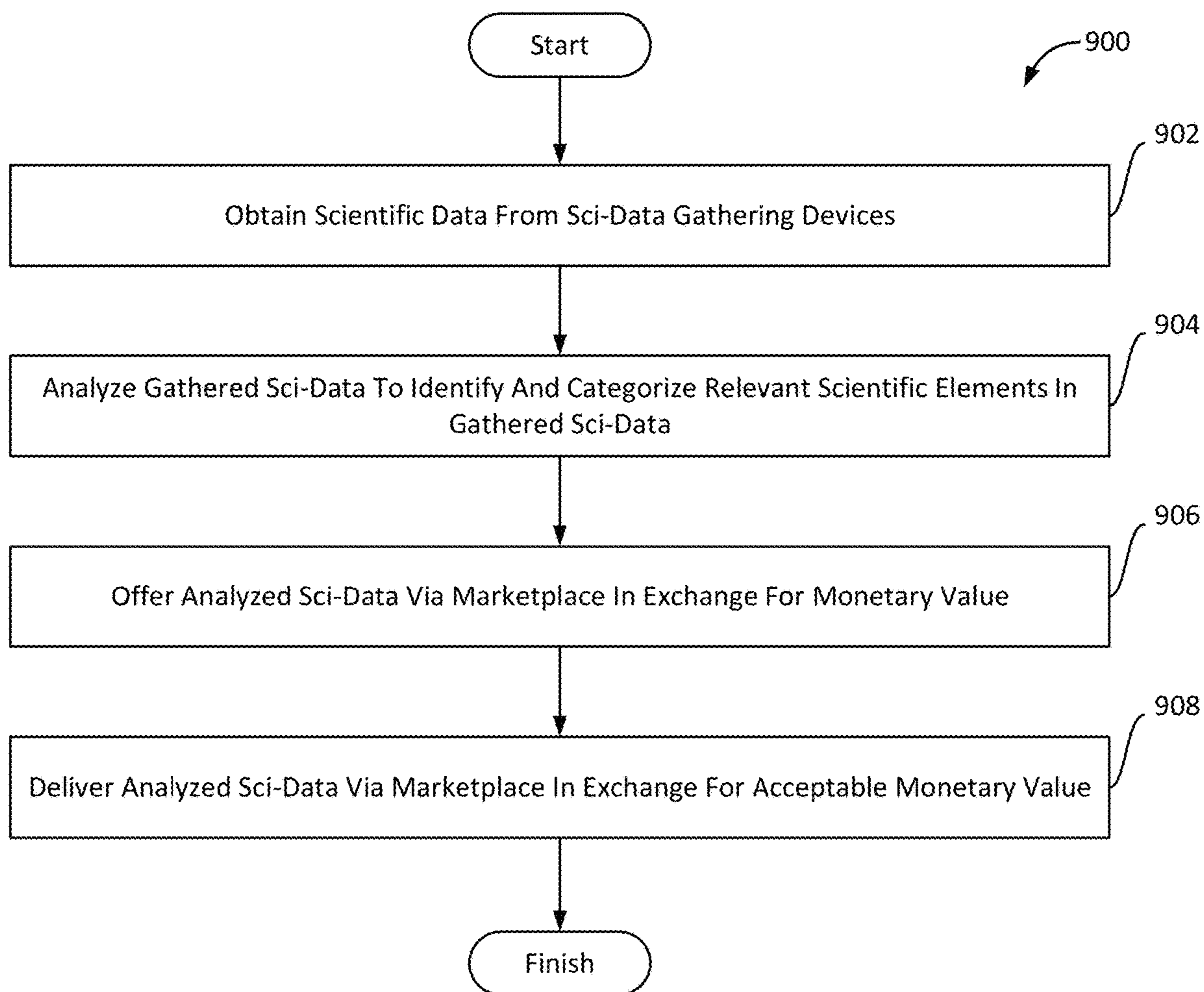


Fig. 9A

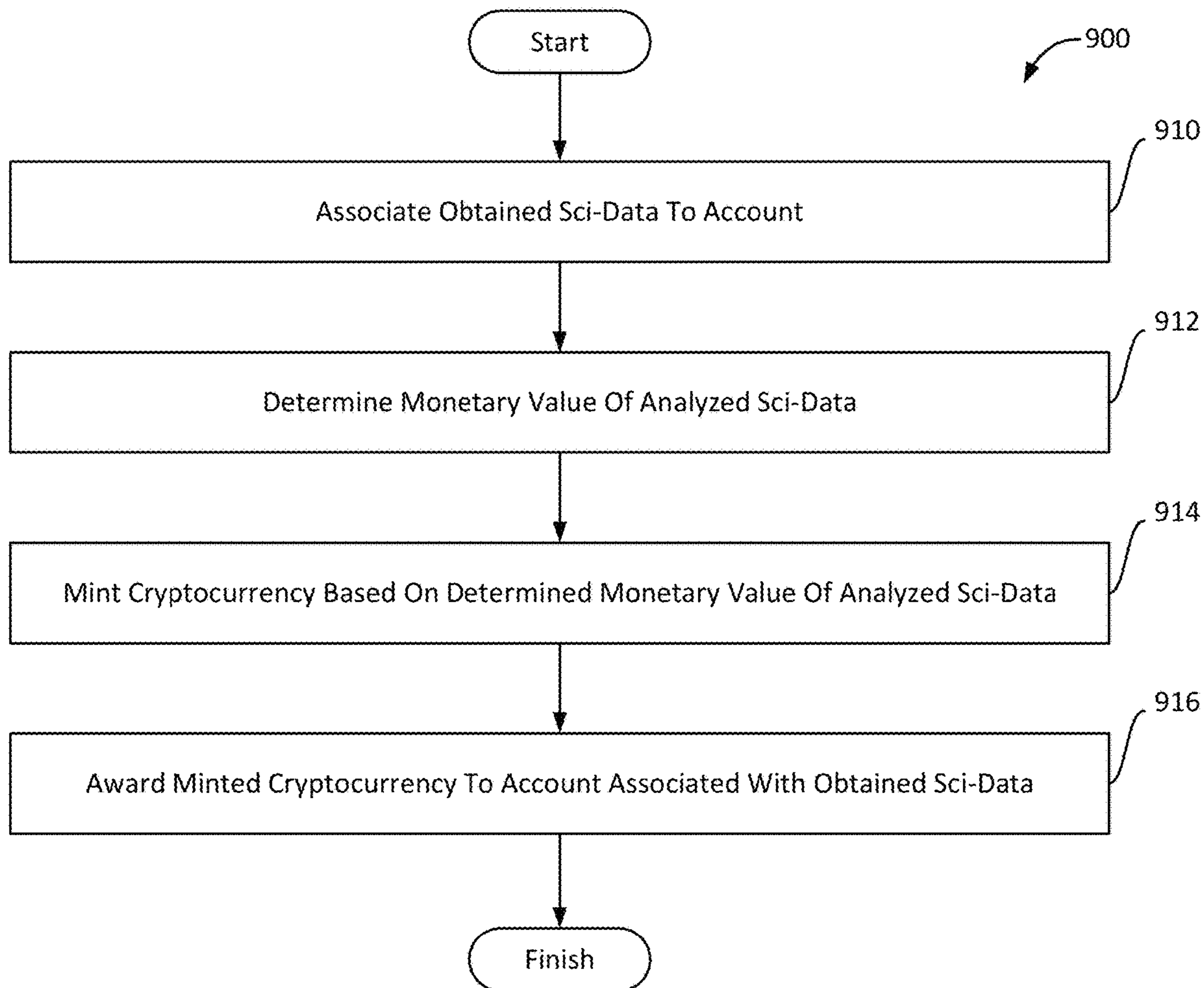


Fig. 9B

FACILITATION OF THE DISTRIBUTION OF SCIENTIFIC DATA

CROSS-REFERENCE TO RELATED APPLICATIONS

The present application claims priority to U.S. Provisional Patent Application No. 62/644,331, filed Mar. 16, 2018, entitled Data Collection and Analytics Based on In-Situ Biological Cell Detection and Identification, U.S. Provisional Patent Application No. 62/549,543, filed Aug. 24, 2017, entitled In-The-Field Pathobiological Cell Detection, U.S. Provisional Patent Application No. 62/588,754, filed Nov. 30, 2017, entitled Data Collection and Analytics Based on In-Situ Biological Cell Detection and Identification, U.S. Provisional Patent Application No. 62/797,703, filed Jan. 28, 2019, entitled Amelioration Based on Detection of Biological Cells or Biological Substances, PCT Patent Application No. PCT/US18/47568, filed Aug. 22, 2018, entitled Detection of Biological Cells or Biological Substances, and PCT Patent Application No. PCT/US18/61841, filed Nov. 19, 2018, entitled Data Collection and Analytics Based on Detection of Biological Cells or Biological Substances, the contents of which are incorporated herein by reference in their entirety.

BACKGROUND

Finding a cure for cancer. Preventing the spread of infectious disease. Defeating superbugs. These are some of the most critical challenges facing our world today, and we are losing ground on each of them.

Today, our approach to fighting global health threats is not working. Many dangerous illnesses still lack efficient prevention or treatment options and continue to affect millions of people every day. Globally, \$7 trillion each year is spent on healthcare. Considering that 90% of clinical trials end in failure, and a single clinical trial can cost between \$800 million and \$1.4 billion, there is a major imbalance in the system.

In the last century, a number of scientific breakthroughs have fundamentally improved global health. Today, people live longer and healthier than ever before. Many diseases that were once fatal or debilitating can now be prevented or treated due to monumental breakthroughs in areas like antibiotics and vaccines. Infections that were once the leading causes of mortality have been largely eliminated. Vaccines have led to the eradication or control of many devastating infectious diseases, including polio, smallpox, diphtheria and measles. Even diseases that still need our attention, like HIV, can be effectively controlled through multi-drug regimens that prevent progression.

Despite these major breakthroughs, many of the most common diseases are not effectively treated by existing therapies. Lung, colon, breast and prostate cancers are all still incurable once they have metastasized. The vast majority of orphan diseases, including rare cancers, lack effective therapies. Heart disease and stroke remain leading causes of mortality, while treatments for psychiatric diseases remain elusive. We have also seen the advent of entirely new issues that we are ill-equipped to tackle, including antibiotic-resistant bacteria (or "superbugs") and large-scale infectious disease prevention for illnesses like Ebola or H1N1.

The world has outgrown the system that has worked so well in the past, and it is time to adapt to our present challenges by creating an entirely new system. New threats demand new solutions.

Before a clinical trial or research project can begin, it requires an enormous amount of funding. Because the entities that have access to funds include major institutions, governments and pharmaceutical companies, these are the groups that end up determining research priorities by choosing which ones to pursue. Many times, these priorities don't align with actual global health threats, but rather with the best interests of the institution or are in response to a market size that would make a certain drug lucrative.

Today, there are significant borders between institutions and governments that mean important research is siloed instead of being widely accessible to all. A group of researchers working to understand a complicated protein as part of a cancer treatment research project in Australia might have no idea that a scientist in Germany has already done the legwork that they need. Meanwhile, that German researcher might have no idea that her protein research could be a critical component to the development of an important cancer treatment. Essentially, scientists around the world don't presently have all the puzzle pieces they need to solve these grand human health challenges. The current peer review process and communication structures in the scientific community make it very difficult for researchers to connect with one another and improve upon each other's work, making it impossible to make real progress toward any of our important, collective goals in global health.

Globally, Clinical trials can cost between \$800 million and \$1.4 billion each to conduct and have an exceptionally high (90%) failure rate. Clearly, the way clinical trials are conducted today is economically inefficient, but the rate of failure also speaks to inefficiencies in our scientific approach.

In a clinical trial to test a potential new drug, for example, scientists start by studying their drug candidate in a sterile, isolated laboratory environment. Drug candidates often appear to work well in these pre-clinical trials, where they are studied in test tubes, Petri dishes or animals like mice. When the drug candidate is moved into human trials, the story changes.

Because the human body is highly complex, it makes sense that a drug candidate would act differently in a human than it would in an artificial, controlled setting. To take the guesswork out of clinical trials and make important discoveries, what is needed is a new system that gives a better understanding of how molecules act in the real world, not just in laboratories, to give scientists a higher chance of success.

SUMMARY

One aspect of the present disclosure relates to a system configured that facilitates distribution of scientific data. The system may include one or more hardware processors configured by machine-readable instructions. The processor(s) may be configured to obtain scientific data from sci-data gathering devices. The processor(s) may be configured to analyze gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data. The processor(s) may be configured to offer the analyzed sci-data via a marketplace in exchange for a monetary value. The processor(s) may be configured to deliver the analyzed sci-data via the marketplace in exchange for an acceptable monetary value.

Another aspect of the present disclosure relates to a method that facilitates distribution of scientific data. The method may include obtaining scientific data from sci-data gathering devices. The method may include analyzing gath-

ered sci-data to identify and categorize relevant scientific elements in the gathered sci-data. The method may include offering the analyzed sci-data via a marketplace in exchange for a monetary value. The method may include delivering the analyzed sci-data via the marketplace in exchange for an acceptable monetary value.

Yet another aspect of the present disclosure relates to a non-transient computer-readable storage medium having instructions embodied thereon, the instructions being executable by one or more processors to perform a method that facilitates distribution of scientific data. The method may include obtaining scientific data from sci-data gathering devices. The method may include analyzing gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data. The method may include offering the analyzed sci-data via a marketplace in exchange for a monetary value. The method may include delivering the analyzed sci-data via the marketplace in exchange for an acceptable monetary value.

These and other features, and characteristics of the present technology, as well as the methods of operation and functions of the related elements of structure and the combination of parts and economies of manufacture, will become more apparent upon consideration of the following description and the appended claims with reference to the accompanying drawings, all of which form a part of this specification, wherein like reference numerals designate corresponding parts in the various figures. It is to be expressly understood, however, that the drawings are for the purpose of illustration and description only and are not intended as a definition of the limits of the invention. As used in the specification and in the claims, the singular form of 'a', 'an', and 'the' include plural referents unless the context clearly dictates otherwise.

BRIEF DESCRIPTION OF THE DRAWINGS

FIG. 1 illustrates conventional techniques to detect and identify pathogens.

FIG. 2 illustrates an example system in accordance with the present disclosure.

FIG. 3 illustrates a chip architecture in accordance with the present disclosure.

FIG. 4 illustrates an example system in accordance with the present disclosure.

FIG. 5 is a flow chart illustrating an example method in accordance with the present disclosure.

FIG. 6 illustrates an example of platform architecture in accordance with the present disclosure.

FIG. 7 illustrates an example system in accordance with the present disclosure.

FIG. 8 illustrates a system configured that facilitates distribution of scientific data, in accordance with one or more implementations.

FIGS. 9A and/or 9B illustrates a method that facilitates distribution of scientific data, in accordance with one or more implementations.

The Detailed Description references the accompanying figures. In the figures, the left-most digit(s) of a reference number identifies the figure in which the reference number first appears. The same numbers are used throughout the drawings to reference like features and components.

DETAILED DESCRIPTION

A technology that facilitates distribution of scientific data are disclosed. Exemplary implementations may: obtain sci-

entific data from sci-data gathering devices; analyze gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data; offer the analyzed sci-data via a marketplace in exchange for a monetary value; and deliver the analyzed sci-data via the marketplace in exchange for an acceptable monetary value.

Disclosed herein is technology to accelerate the development of cures for global health threats, including preventive treatments, therapeutics, vaccines and drug development. Taking creative advantage of cutting-edge technologies, a secure, borderless economy is created that mints its own currency to incentivize data collection and empower every organization, government, professional scientist or global citizen scientist to participate. The model is anchored by a platform, which includes a microprocessor chip to be used in various devices, such as Internet of Things (IoT) develops. The chip may be based on various architectures, such as RISC architectures, for example Arm Semiconductor Company based architectures. The chip can collect molecular data in real time, increasing the amount of available data by orders of magnitude. It is to be understood that platform and system are interchangeable terminology. The environment can include the platform or system, and a marketplace that are further described below.

Furthermore, blockchain technology may be used along with cryptocurrency or a Coin, to secure and authenticate data, attribute each piece of data to its correct source and compensate participants for their contributions.

With blockchain technology and the Coin serving as the foundation, the economy can include the following four fundamental concepts:

1) Aggregation of research-based data from organizations and institutions on a global scale;

2) The collection of real-time, real-world molecular level data from our devices implementing the chip and Internet of Things (IoT) devices;

3) The application of artificial intelligence and machine learning systems to identify trends and draw conclusions, providing scientists and researchers a tool to create life-saving solutions;

4) Inclusion of everyone on a global scale, from the sick or threatened to the research scientist or citizen scientist.

Example Scenario

Listeria monocytogenes is a pathogen that causes listeriosis, which is an infection with symptoms of fever, vomiting, and diarrhea. This pathogen is an example of a pathobiological cell. *Listeria* can spread to other parts of the body and lead to more serious complications, like meningitis. *Listeria* is often transmitted by ready-to-eat foods, such as milk, cheese, vegetables, raw and smoked fish, meat, ice cream, and cold cuts. This early and rapid detection is desirable so that cross-contamination can be avoided and any problems immediately addressed.

These ready-to-eat foods are often mass produced in food factories. In these factories, there is little to no time to stop production to test to determine if a harmful pathogen (like *listeria*) exists on the food-production surfaces. Depending on the comprehensiveness and desired accuracy of the test, conventional techniques to detect the bacteria take as long as a week to as short as hours. Regardless of the particulars of the test, these conventional tests involve the manual collection of samples from various surfaces, cataloging these samples, and performing invasive testing (e.g., culturing, chemical reaction, antibodies, etc.).

FIG. 1 illustrates conventional techniques **100** to detect and identify pathogens. Table **110** has a surface that, of course, has microbes thereon. This table **110** represents anything with a surface area that might have microbes living on it. For this discussion, assume that table **110** is in a commercial kitchen of a ready-to-eat food manufacturer. This manufacturer is concerned about *Listeria* in this kitchen. To detect the existence of *Listeria* in its kitchen, the manufacturer orders spot tests be performed in the kitchens.

To that end, a spot **112** on table **110** is selected for sampling. Using a sample-collection swab **120**, a tester swipes the spot **112**. Following arrow **130**, a sample-collected swab **122** is carefully transported to a testing station **140** so as to avoid collateral and inadvertent collection of samples from other sources.

Typically, this testing station **140** is physically separated and distant from the sample spot **112** of the commercial kitchen where the sample was collected. The testing station **140** is often in a laboratory of a testing facility. With traditional methods, the sample microbes **124** of the sample-collected swab **122** are transferred to Petri dishes **144** for cultivation. At some point, chemicals **142** may be added to the cultured microbes of the Petri dishes **144** for various reasons, such as dyes to make them more visible.

Following arrows **132** and **134** and perhaps weeks or months, a petri dish **146** with the adulterated (e.g., dyed) cultured microbes is ready to be examined under a microscope **150**. Typically, a human examines the cultures under the microscope **150** and identifies pathogens amongst the cultured microbes based on many factors, but mostly the human's professional experience in identifying such microbes.

Traditional methods of testing like that demonstrated in FIG. 1, where sample microbes are cultivated in labs, are flawed. 'Stressed' cells will not grow in cultures (and will, therefore, produce negative results) despite the bacteria being present, live and potentially harmful. Also, this is the slowest form of testing.

Alternative conventional techniques, based on molecular/chemical methods, detect all cell types, but don't differentiate between live and harmless dead cells, which can remain after disinfection. Thus, these molecular/chemical methods may indicate a false positive for the presence of a pathogen when only dead cells of the pathogen are present.

Still, other conventional techniques use antibodies to test biofilms, which are groups of microbes where cells stick together on a surface. This technique requires the biofilms to be removed from the surface, treated with a particular antibody, and then tested to see if the biofilm fluoresces. This type of technique only tests for the particular pathogen that the introduced antibody interacts with.

Example Electronic Device

FIG. 2 illustrates an example scenario **200** in accordance with the technology described herein. The example scenario **200** includes a table **210**. That table has a scene **212**, which is an area of a surface in view of a camera (not shown) of a smartphone **220**. Indeed, the camera captures an image **222** of scene **212**. Just like reality, the scene **212** includes multiple microbes, but these microbes are not visible in scene **212**.

For the example scenario **200**, the microbes are described as in situ (i.e., in place) because they are examined, tested, etc. where they naturally live, inhabit, or exist. That is, the microbes are not in the lab. Herein, "in the lab" indicates that the microbes have been moved, relocated, or expatriated in

order to perform the examination, testing, or the like. Other implementations of the technology described herein may involve microbes that are in the lab.

That image **222** is depicted as being viewed on a display of the smartphone **220**. The image **222** has been sufficiently magnified to be able to see various in situ microbes of scene **212**. And while not yet detected, one of these microbes is a pathogen **224**.

The smartphone **220** is one example of an electronic device **230** in accordance with the technologies described herein. However, in other example scenarios, the electronic device **230** may be, for example, a tablet computer, a smartdevice, a standalone device, a collection of cooperative devices, a button-sized device, a device on a chip, an accessory to a smartphone or smartdevice, an ambulatory device, a robot, swallowable device, an injectable device, embedded within medical lab equipment, or the like.

As depicted, the electronic device **230** includes a scene-capture system, a biologic detection system (and object detection system) **234**, a database **236**, an environmental sensor system **238**, a report system **240**, and an amelioration system **242**. These systems of the electronic device **230** are constructed from hardware, firmware, special-purpose components (e.g., sensors), and/or some combination thereof. These systems may, in some instances, include software modules as well.

The scene-capture system **232** is designed to obtain an image (e.g., image **222**) of a scene (e.g., scene **212**) that includes in situ biological cells therein. That is, there are biological cells located in a place (i.e., in-the-field) in the scene that is being captured by the scene-capture system. In some implementations, the scene-capture system **232** includes a camera to capture the visible part of the electromagnetic spectrum that is emitting or reflecting from the matter contained in the scene. In some implementations, the scene-capture system **232** includes components designed to capture non-visible parts of the electromagnetic spectrum (e.g., x-rays, infrared, gamma rays, ultraviolet, etc.) that is emitting or reflecting from the matter contained in the scene.

Examples of the action of obtaining (as performed by the scene-capture system **232**) include measuring, collecting, accessing, capturing, procuring, acquiring, and observing. For example, the scene-capture system **232** may obtain the image by capturing the image using the charge-coupled device (CCD) of the digital camera. In another example, the scene-capturing system **232** may obtain the image by measuring the electromagnetic spectrum of the scene.

The obtained image is micrographic, spectrographic, digital, or some combination thereof. The obtained image is micrographic because it captures the elements in the scene that are on a microscopic scale. The obtained image is spectrographic because it captures elements in the scene by using equipment sensitive to portions of the electromagnetic spectrum (visible and/or non-visible portions). The obtained image is digital because it formats and stores the captured information as data capable of being stored in a machine, computer, digital electronic device, or the like.

While the thing that is captured is called an image, this image is not necessarily displayable as a two-dimensional depiction on a display screen (as shown in image **222**). Rather, the image is an array of data that represents the quantitative and qualitative nature of the electromagnetic spectrum (or some portion thereof) received by the components of the scene-capture system **232** when it was exposed to the scene (e.g., scene **212**).

The biologic detection system **234** is designed to analyze the obtained image and detect the presence of one or more

pathobiological cells amongst the in situ biological cells of the captured scene. In some implementations, the biologic detection system **234** may actually identify one or more particular cells and/or substances in the scene. In that case, it may be called a biologic identification system. Such a system may identify the particular pathobiological cells amongst the in situ biological cells. Thus, depending on the implementation, this system **234** may be referred to as biological-cell detection system or the pathobiological detection system.

To accomplish detection, the biologic detection system **234** may employ on and/or employ a database **236**. This database **236** may be a database of pathobiologic-cellular signatures or training corpus. The biologic detection system **234** is a particular example of a biologic-cellular detection system. A training corpus is a database of numerous application-specific samples from which the AI/ML/DL engine “learns” and improves its capabilities and accuracy.

The biologic detection system **234** employs an AI/ML/DL engine to perform or assist in the performance of the detection and/or identification of one or more pathobiological cells. AI/ML/DL is short for artificial intelligence/machine learning/deep learning technology. Particular implementations may employ just an AI engine, just an ML engine, just a DL engine, or some combination thereof.

The AI/ML/DL engine may be implemented just on the smartphone **220** itself. In that case, the smartphone **220** need not communicate in real time with the platform (e.g., a remote computing system). In another implementation, the AI/ML/DL engine may be implemented just on the platform (thus remotely). In that case, the smartphone **220** communicates in real time (or nearly so) with the platform (e.g., a remote computing system). In still other implementations, the AI/ML/DL engine is implemented partially in both the smartphone **220** and the platform. In this way, the intensive processing is offloaded to the platform.

Some implementations of the biologic detection system **234** may perform its data analysis solely on the device without assistance from other devices, servers, or the cloud. Other implementations of the biologic detection system **234** may farm out all or nearly all of the data analysis to other devices, servers, or the cloud. In still other implementations, the data analysis may be shared amongst multiple devices and locations.

On its own or working with other devices or computer systems, the biologic detection system **234** analyzes the image of the scene to detect, determine, and/or identify the type or class of biological cells therein. It may do this, at least in part, by using distinguishing molecules of the cells that are observable using the electromagnetic spectrum. In some implementations, other data (such as chemical reactions or excitation) may be included in the analysis.

Some of these molecules are indicative of certain classes, types, or particular cells. Such molecules are called marker biomolecules herein. The electronic device can determine which cell types or class are present in a captured scene by the on the particular ones of, the types of, and the proportions of the biomolecules detected therein. This may be accomplished, at least in part, by calculating probabilities of objects detected in the image.

The environmental sensor system **238** is designed to measure one or more environmental factors associated with the in situ biological cells or the environment surrounding the in situ biological cells. In some instances, the environmental sensor system **238** may be simply described as a sensor.

The report system **240** is designed to report detection and/or identification of the one or more pathobiological cells in the obtained image and in some implementations, the report system **240** is designed to associate the measured environmental factor with the obtained image and/or with the detected pathobiological cell.

The amelioration system **242** is designed to respond to the detection and/or identification in a manner that ameliorates the pathobiological nature of the detected/identified pathobiological cells.

The electronic device **230** may have a communications system to send or receive data from other similar electronic devices or centralized/distributed servers. The electronic device **230** may have enhanced processor or co-processor to perform image-capture and processing functions.

Of course, the example scenario **200** described above is one implementation that detects pathobiological cells. Other implementations of the technology described herein may detect or identify pathobiological substances rather than cells.

One or more of the systems of electronic device **230** may be characterized as a non-transitory computer-readable storage medium comprising instructions that when executed cause one or more processors of a computing device to perform the operations of that electronic device.

Pre-Processed Data

Typically, images of the same scene are captured over-time. That may be over a few seconds, few minutes, or perhaps a few hours. In so doing, a massive amount of raw image data is produced. So much data that it may quickly overwhelm the local storage capacity of the smartphone **220** and often overwhelms the data transfer rate between the smartphone **220** and any network-based storage solution.

To address this issue, the smartphone **220** may be designed to pre-process or process the raw scene-captured data before storing locally or transferring it across the network. For pre-processing, the smartphone **220** may derive just the most critical or key data that helps identify or reconstruct the scene.

For example, the smartphone **220** may store the marker biomolecule information from each captured image or scene. The marker biomolecule information includes just the data regarding the type, amount, proportions, etc. of the marker biomolecules or substances detected, determined, identified, etc. in a particular image or scene. Thus, any other data from the image capture is discarded.

Along with associated environmental factors, this pre-processed information is stored or transferred across the network. This reduces the data storage/transfer requirements by multiple orders of magnitude. The particular cell type or class is determined from this stored or transferred pre-processed data. This may be done later or by different systems.

In some instances, the smartphone **220** may fully process the image/scene captured data to determine, detect, and/or identify the cell type or class. In this scenario, the electronic device stores or transfers its conclusion about the cell type or class with its associated environmental factors.

Scale of Amount of Data

Since the images being captured are on a microscopic scale, it may take many images to capture a small surface area of an object. In addition, even a short sequence of images adds up quickly to a great multitude of images. Thus, in only a short space (e.g., of just a few seconds), the sequence of microscopic-scale images of a very small area quickly overwhelms the internal data transfer, data storage, and/or data processing capability of a typical electronic

device (such as a smartphone). In addition, the typical external data transfer rates (e.g., of wireless communication) is not capable of accepting the data tsunami of this technology.

Two example approaches may be employed to address these issues. One involves the increased capacity of the electronic device, and the other involves the processing of the data into a manageable form.

First, this technology is implemented in such a way to employ special-purpose hardware to perform the pre-processing and processing of the incoming real-time data. That is, the specially designed hardware is built directly into the processors and electronics of the device to enable the device to quickly process the massive amount of incoming real-time data into a representative portion thereof without losing important aspects of the data.

Second, this technology employs a particular mechanism to produce a representative portion thereof without losing important aspects of the data. In short, that involves saving the deltas (i.e., changes) between the measurements (e.g., marker biomolecules) over time. These deltas are stored and/or transferred. In addition, data compression schemes may be used.

Imaging

The technology described herein utilizes an image-capturing system, such as the scene-capture system 232. In some instances, the image-capturing system may be called a camera. This is particular so when the system captures the visible part of the electromagnetic spectrum that is emitting and/or reflecting from the matter being observed. In some implementations, the image-capturing system may capture non-visible parts of the electromagnetic spectrum (e.g., x-rays, gamma rays, ultraviolet, etc.) that are emitting or reflecting from the matter being observed.

With some implementations, the image-capturing system may employ hyperspectral imaging and, in particular, snapshot hyperspectral imaging. Hyperspectral imaging collects and processes information from across a portion of the electromagnetic spectrum. With hyperspectral imaging, the spectrum is captured for each pixel in the image of a scene. Snapshot hyperspectral imaging uses a staring array (rather than a scanning array) to generate an image in an instant.

With some implementations, the image-capturing system may employ light-field imaging, which is also called plenoptic imaging. A plenoptic imaging system captures information about the light field emanating from a scene. That is, it captures the intensity of light in a scene, and also the direction that the light rays are traveling in space. This contrasts with a conventional camera, which records only light intensity.

Using plenoptic imaging enables the simultaneous capture of pictures at different focal points, allowing the device sensor to capture a two-dimensional image in multiple 3rd dimension planes (i.e., capture a volume of space vs. a plane). Capturing a volume facilitates faster detection on non-flat surfaces or when fluids or gases are observed.

In addition, a combination of both these hyperspectral and plenoptic technologies may be used. That is, the image-capture system may incorporate both snapshot hyperspectral imaging with plenoptic imaging.

Nanophotonics

In some instances, an agent is purposefully introduced into the scene, environment, or in the lab to enhance or improve the observations or measurements. For example, photonic nanostructures may be spread in the environment where measurements and observations may be made.

These photonic nanostructures are part of a field called nanophotonics or nano-optics, which involves the study of the behavior of light on the nanometer scale, and of the interaction of nanometer-scale objects with light. It is a branch of optics, optical engineering, electrical engineering, and nanotechnology. It often (but not exclusively) involves metallic components, which can transport and focus light via surface plasmon polaritons.

The term “nano-optics,” just like the term “optics,” usually refers to situations involving ultraviolet, visible, and near-infrared light (free-space wavelengths from 300 to 1200 nanometers).

Using nanophotonics to create high peak intensities: If a given amount of light energy is squeezed into a smaller and smaller volume (“hot-spot”), the intensity in the hot-spot gets larger and larger. This is especially helpful in nonlinear optics; an example is surface-enhanced Raman scattering. It also allows sensitive spectroscopy measurements of even single molecules located in the hot-spot, unlike traditional spectroscopy methods which take an average over millions or billions of molecules.

One goal of nanophotonics is to construct a so-called “superlens”, which would use metamaterials or other techniques to create images that are more accurate than the diffraction limit (deep subwavelength).

Near-field scanning optical microscope (NSOM or SNOM) is another nanophotonic technique that accomplishes the same goal of taking images with resolution far smaller than the wavelength. It involves raster-scanning a very sharp tip or very small aperture over the surface to be imaged.

Near-field microscopy refers more generally to any technique using the near-field to achieve nanoscale, subwavelength resolution. For example, dual polarization interferometry has picometer resolution in the vertical plane above the waveguide surface.

Environmental Factors

As indicated above, sensors obtain environmental factors related to, about, or near the scenes being observed. These may be called ambient factors. The sensors may measure or sense to obtain the environmental factors. In other instances, the factors may be accessed, acquired, procured, etc. via another source, sensor, memory, machine, etc.

The environmental factors are abiotic or biotic. However, there are other datapoints that may be gathered, but which are not expressly related to the environment. These may be called associated or observation-related factors.

An abiotic environmental factor is associated with non-biological sources. That is, the source of the thing being measured is not related to a living or recently living thing.

Examples of abiotic environmental factor include ambient temperature, timestamp (e.g., time and date), moisture, humidity, radiation, the amount of sunlight, and pH of a water medium (e.g., soil) where a biological cell lives. Other examples of abiotic environmental factors include barometric pressure; ambient sound; indoor location; ambient electromagnetic activity; velocity; acceleration; inertia; ambient lighting conditions; WiFi fingerprint; signal fingerprints; GPS location; geolocation; airborne particle counter; chemical detection; gases; radiation; air quality; airborne particulate matter (e.g., dust, 2.5 PPM, 10 PPM, etc.); atmospheric pressure; altitude; Geiger counter; proximity detection; magnetic sensor; rain gauge; seismometer; airflow; motion detection; ionization detection; gravity measurement; photoelectric sensor; piezo capacitive sensor; capacitance sensor; tilt sensor; angular momentum sensor; water-level (i.e., flood) detection; flame detector; smoke detector; force

gauge; ambient electromagnetic sources; RFID detection; barcode reading; or some combination thereof.

A biotic environmental factor is one having a biologic source. Example of such include the availability of food organisms and the presence of conspecifics, competitors, predators, and parasites.

While it is not an environment factor, per se, the observation-related or associated factor is described here. The associated or observation-related factor may be a measurement of quality, quantity, and/or characteristic of the environment about the observation itself or related to the environment from which the subject is observed or was obtained. They may also be data that a human or computer associated with other environmental factors or the scene.

Examples of the observation-related or associated factor include nearby traffic patterns or noises; tracking the movements of particular individuals (e.g., via employee badges or security cameras); visitors; patients, budgets of related departments; and the like.

Herein, a known sensor or measurement device may be listed as an example of an environmental factor. For example, Geiger counter and seismometer are listed as examples. It should be understood that the relevant factor for least listed examples is the measurements typically made by such devices. Thus, the obtained factor for example Geiger counter is radiation and the obtained factor for the example seismometer is the motion of the ground.

Artificial Intelligence, Machine Learning, and Deep Learning Technology

Herein, the term AI/ML/DL technology refers to a technology that employs known or new artificial intelligence (AI) techniques, machine learning (ML) techniques, deep learning (DL) techniques, and/or the like.

By applying AI/ML/DL technology such as convolutional neural networks (CNNs), some implementations of the technology described herein is capable of identifying pathobiological cells and substances within microscopic images and spectrographic signatures that environment and systems ingest from either existing datasets or streams of real-time sensor data.

By training its neural networks against libraries of high-quality pathobiological cells/substances images and signatures, the technology described herein can reliably identify specific cells/substances. Upon discovery, the technology described herein may take advantage of a sophisticated queueing system to retroactively “replay” historical data with a greatly increased sampling rate, enabling it to build a high-resolution model of the outbreak. This model is then added to the chain, fully secure, attributed and available to researchers who can use it to help contain the outbreak or to advance the understanding of its transmission model.

For example, the technology described herein can provide for real-time and real-world data. The chip is used in the data ingest pipeline and marketplace. Deployed throughout a building and/or across a region and using the sensor technology to pick up environmental (e.g. temperature, humidity, etc.), visible, and spectrographic data which, coupled with ambient other (e.g., location, time, etc.) data, the numerous chips in the system can together stream enormous volumes of valuable data into the platform for processing by the artificial intelligence insights engine described herein. As used herein, a platform includes a remote device with massive storage capacity and processing power.

The technology described herein may utilize AI to detect objects that have already been learned by a device or platform implementing the technology. This minimizes the amount of data transmitted, efficiently utilizing communi-

cation bandwidth. The sensed and other data associated with objects that the technology described herein detects, but cannot identify, will be sent to the platform which would, in turn, trigger an investigation that gathers real-world samples and take those samples to a lab for controlled analysis and identification using machine learning with deep learning. Once identified in the lab, the platform can send a specific detection approach to an implementation of the technology described herein so that it can then confidently identify the new object is going forward.

The technology described herein will either contain or connect to sensors that will enable novel abilities to detect pathobiological cells and substances at low concentrations, in noisy environments (e.g. objects in the midst of dust, dirt, or other uninteresting objects), in real-time (e.g. without significant delay from when object was in the field of view), in some cases without disturbing the environment detected objects (i.e. passive observation), and with the ability to search large three dimensional spaces so as to reduce the probability of not observing an interesting object which is especially important when the objects are infrequently present.

Having some or all of these qualities, coupled with detection assisted by AI/ML/DL engines will facilitate detection and identification that is much faster than present technology, more accurate, and possible outside of a lab environment.

Some implementations of the technology described herein utilize AI/ML/DL technology in the detection or identification of pathobiological cells or substances from collected data. In other implementations, the technology described herein may utilize AI/ML/DL technology in the analysis of the collected data with metadata such as environmental factors collected therewith.

According to Apr. 4, 2018, cloudmayo.com article (“Difference between Artificial Intelligence, Machine Learning and Deep Learning”), there is much confusion between three different but interrelated technologies of AI, ML, and DL. The article defines AI as a “technique that allows a computer to mimic human behavior,” ML as a “subset of AI techniques that uses a statistical method to enable a machine to improve with experiences,” and DL as a “subset of ML which makes the computation of multi-layer neural networks feasible.”

The electronic component or collection of components that employs an AI/ML/DL technology for training and/or data analysis is called an AI/ML/DL technology herein.

System-on-a-Chip

FIG. 3 illustrates an example system-on-a-chip 400, which is an implementation of the technology described herein. As shown, the system-on-a-chip 400 include a semiconductor chip integrated into a single package or as a chipset 301. Although a particular chip architecture is described, it is to be understood that other semiconductor chip architectures may be implemented.

The chip 301 can be resident in different devices, such a smart phones, laptop/table computers, dedicated medical/research devices, etc. Such devices are used for detection of pathobiological cells or substances. In certain implementations, computing may be performed off-chip via the cloud, or performed on the chip.

The chip 301, and in particular devices including the chip 301, may use AI/ML/DL engines to process data. In particular, AI/ML/DL engines may be used by the chip 301 in the accelerated processing of collected data, such as environmental factors and other data to detect/identify pathobiological cells and substances. In addition, doing so reduces data bandwidth of communication (e.g., to the platform).

Also, distributed processing at the device reduces the cost of the device and reduces communication bottlenecks.

The chip **301** includes a processor(s) or processing component **300**, cache memory **302**, security component **304**, optical detection **306**, and digital detection **308**. Depending on the implementation, the digital detection **308** may be used for one or more of the following: digital enhancement, object detection, or object identification.

Chip **301** can include one or more AI/ML/DL engines or accelerators **309**. The AI/ML/DL accelerators **309** can implement edge and/or cloud computing. Chip **301** further can include encryption **310**, and debug and trace component **312**.

Interfaces **314** are provided/included in chip **301**. In particular interfaces **314** provide the ability to receive sensory input, such as environmental factors (e.g., temperature, air pressure, wireless signals, capacitance, etc.) and capture image of a scene (e.g., via a microscope, camera, or spectrometer). The interface **314** also allows for transmission of data, network connections, user input, status indicators, and control illumination circuitry.

Interfaces **314** may further connect, for example, to an in-package dynamic random-access memory (DRAM) **320**, in-package electrically erasable programmable read-only memory (EEPROM) **322**, and power management integrated circuit (PMIC) **324**.

Network of Monitoring Devices

FIG. 4 is an illustration of an infrastructure **400** that facilitates data collection and data analysis of the technology described herein. The data collection, for example, may occur across a network of various widths (i.e., horizontally) and depths (i.e., vertically). Consider for this example a single building. That building is called the Hope Hospital **410**.

The Hope Hospital **410** has many different floors, rooms, departments, etc. For illustration purpose, the Hope Hospital **410** has four floors of five rooms each. And each floor is a department.

The Hope Hospital **410** has installed a variety of electronic monitoring devices. These devices are like smartphone **220** of example scenario **200**. However, these electronic monitoring devices may be stationary and special-purpose. These electronic monitoring devices are called NANOBOT™ devices herein. However, these NANOBOT™ devices are special-purpose devices rather than a smartphone. That is, the NANOBOT™ devices are designed and built for the specialized and particular purpose of the functionality described herein.

In this example, there are multiple NANOBOT™ devices placed throughout the hospital. Indeed, there could be multiple devices in each room. For example, a NANOBOT™ device may be installed on the bed, the bathroom counter, hand sanitizer dispenser, the faucet handle, the air vent, and the like. In addition, other NANOBOT™ devices may be installed in the ductwork of the HVAC.

As depicted, stationary device **412** is mounted on the wall in a room of patient X, device **414** is on the ceiling of a room of patient Y, device **416** is on the bed of patient Z, device **418** is mounted in the ducting of the third floor, and device **420** is a portable device carried by a nurse.

Each of these NANOBOT™ devices are installed to monitor biological cells and/or substances and environmental factors in their proximity. The devices will detect and/or identify the particular cells and/or substances in their proximity. In addition, the ongoing monitoring of these devices enable the tracking of changes in the detected and/or identified microscopic lifeforms, for example, in their proximity.

In addition, each of these NANOBOT™ devices include a sensor or sensors for monitoring one or more environmental factors, such as ambient temperature, humidity, indoor location, and the like. Each device tracks its proximate factors and microscopic lifeforms over time.

In addition to the stationary NANOBOT™ devices, other electronic devices may be used in the Hope Hospital **410**. For example, there may be mobile or ambulatory devices that are specially designed to do the same functions. These devices may be affixed to a mobile or portable platform. Alternatively, these devices may have the ability to relocate on their own power or volition.

For example, NANOBOT™ device may be affixed to a cart, robotic medication dispenser, a chair, and the like. As such, the NANOBOT™ device tracks the environmental factors and biological cells proximate the device as the thing to which it is affixed is moved throughout the hospital. As such, the indoor location of the device changes as well.

Similarly, a NANOBOT™ device may have its own mechanism for self-propulsion. It may have electric motors, wheels, and self-navigational capability to travel the hallways, walls, and ducts of the building. In addition, some self-propelled NANOBOT™ devices may travel in the air, for example, like a so-called drone (i.e., unmanned aerial vehicle). In some instances, a self-propelled NANOBOT™ device may travel on and/or within the liquid.

The self-propelled NANOBOT™ device may wander around the hospital or other space to generate a comprehensive amount of monitoring data for such space or a portion thereof. Alternatively, the self-propelled NANOBOT™ device may travel a pre-determined path or navigate its own path. In doing so, the device is tracking data as it travels and/or at particular points along the path.

In addition, humans may carry a smartphone or some form of smartphone accessory (e.g., watch) that is capable of performing these functionalities. Indeed, this type of mobile device may perform these functions actively and/or passively. That is, the user may actively choose when and where to perform measurements and/or the device may choose when and where to perform measurements.

These various devices in the Hope Hospital **410** may be interconnected with each other and/or connected to a common or interconnected network **430**. For example, the stationary NANOBOT™ devices may be connected to each other via a peer-to-peer mesh wireless network. Devices may be connected via a wireless access point (WAP) or via short-range wireless signal (e.g., BLUETOOTH). In turn, these networks may be connected to other networks (such as the so-called Internet) and to centralized or distributed computing center. For example, all of the stationary NANOBOT™ devices in a room may be connected at a single nearby WAP that is connected to the so-called cloud where the acquired data is stored in a cloud-based storage system. The network **430** represents any and all of these suitable communication networks.

Data Collection

Each device in the Hope Hospital **410** is configured to monitor its proximate area. The devices monitor for biological cells and various environmental factors, such as location, temperature, and humidity. Each device may be configured to collect data in a manner like that smartphone **220** of example scenario **200**.

Once the data is collected by a monitoring device, it may be locally analyzed or the raw data may be transferred or uploaded to a centralized and/or distributed computing sys-

tem. Supersystem **432** is depicted as a several servers. This represents the centralized and/or distributed computing system.

If analyzed locally, the collected data may be fully or partially analyzed locally. With local analysis, some or all of the processing that is necessary to detect and/or identify a class or type of biological cell is performed on the electronic monitoring device. Indeed, if fully processed locally, the monitoring device may form a conclusion regarding the type/class of cell in a monitored scene.

In some instances, the raw monitor data or partially processed data may be uploaded or transferred to a computing system. For example, each floor of the Hope Hospital **410** may have its own dedicated computer to store monitor data of the devices of that floor. That floor computer may perform some or all of the calculations needed to determine that type/class of the cells in the monitored scenes. Alternatively or in addition, each building may have its own computer; each campus has its own computer; each city has its own computer; each region has its own computer; etc. Alternatively or in addition, all of this data is transferred to the “cloud” for storage and processing overall or at each level.

In addition, individuals may collect data for their own personal reasons. For example, a visitor may collect data in the cafeteria of the hospital so that she knows how clean the surfaces are. This data and/or its analysis may be uploaded to the so-called cloud. That is, the data collection may be crowdsourced or available to the individual alone.

This data may be collected in a coordinated fashion by the Hope Hospital **410** or an agency working on their behalf. The collection and analysis of this data may be performed by the Hope Hospital **410**. In addition, the collection and analysis of the data of Hope Hospital **410** may be a service that the Hope Hospital **410** subscribes to.

Furthermore, a service may collect data from many and various locations in an anonymous fashion to protect the private data of each identifiable customer. With the data collected from many different locations and customers, the service may analyze the data to find meta-trends and meta-correlations.

The supersystem **432** includes one or both systems **440** and **460**. System **440** is primarily for the collection and analysis of image-based data. System **460** is primarily for the inferences or conclusions of the collected and analyzed data. The supersystem may be called the “platform” herein.

System **440** includes a data communications subsystem **442**, a biologic detection subsystem **444**, and a report subsystem **448**.

The data communication subsystem **442** obtains (e.g., via wireless communication) image-based data from one or more of multiple remote monitoring devices. The image-based data from each monitoring device is based on (e.g., derived from) one or more images of a scene proximate to that monitoring device. The proximate scene includes biological cells and/or substances therein.

The data communication subsystem **442** also obtains (e.g., via wireless communication) environmental data from one or more of the multiple remote monitoring devices. The environmental data being based on an environmental factor associated with the in-scene biological cells and/or substances of each device or the environment surrounding the in-scene biological cells and/or substances of each device.

The biologic detection subsystem **444** analyzes the image-based data and/or environmental data. Based on that analysis, the biologic detection subsystem **444** detects and/or identifies a type or class of biological cells and/or substance

amongst the in-scene biological cells and/or substances of each device or amongst several of the devices.

To accomplish detection, the biologic detection subsystem **444** may employ on and/or employ a database **446**. This database **446** may be a database of biologic-cellular or biologic-substantive signatures. This may be called a training corpus. A training corpus is a database of numerous application-specific samples from which the AI/ML/DL engine “learns” and improves its capabilities and accuracy.

The biologic detection subsystem **444** employs an AI/ML/DL engine to perform or assist in the performance of the detection and/or identification of one or more biological cells and/or substances.

The AI/ML/DL engine functionality may be split across the platform. That is, the devices may perform pre-processing of the image using AI/ML/DL engine and send the results as the image-based data to the system **440** for further processing herein. In that case, the device **414** communicates in real time (or nearly so) with the platform. In this way, the intensive processing is offloaded from the devices to the platform.

The biologic detection subsystem **444** analyzes the image-based data of the scene to detect, determine, and/or identify the type or class of biological cells and/or substances therein. It may do this, at least in part, by using distinguishing molecules of the cells that are observable using the electromagnetic spectrum. In some implementations, other data (such as chemical reactions or excitation) may be included in the analysis.

The report subsystem **448** reports the detection and/or identification of the type of biological cells and/or substances in the scene proximate to the monitoring device. The report subsystem **448** may send **449** its results and associated data (image-based data and/or environmental data) to the system **460**.

Data Analysis

As discussed herein, an electronic device captures a scene that has biological cells therein. In some instances, these biological cells may be described as in situ (i.e., in place) because they are monitored, examined, tested, etc. where they naturally live, inhabit, or exist. That is, the biological cells have not been moved, relocated, or expatriated in order to perform the examination, testing, or the like.

Using a camera or digital or other imaging technology, the electronic device captures a portion of the electromagnetic spectrum that is emitting or reflecting from the matter contained in the scene. The obtained image is micrographic, spectrographic, digital, or some combination thereof. The obtained image is micrographic because it captures elements in the scene that are on a microscopic scale. The obtained image is spectrographic because it captures elements in the scene by using equipment sensitive to portions of the electromagnetic spectrum (visible and/or non-visible portions). The obtained image is digital because it formats and stores the captured information as data capable of being stored in a machine, computer, digital electronic device, or the like.

While the thing that is captured is called an image, this image is not necessarily displayable as a two-dimensional depiction on a display screen. Rather, the image is an array of data that represents the quantitative and qualitative nature of the electromagnetic spectrum (or some portion thereof) received by the camera of the electronic device when it was exposed to the scene.

On its own or working with other devices or computer systems, the electronic device analyzes the image of the scene to detect, determine, and/or identify the type or class

of biological cells therein. It may do this, at least in part, by using distinguishing molecules of the cells that are observable using the electromagnetic spectrum. That is, the electronic device captures the observable electromagnetic spectrum (e.g., visible and/or non-visible) that is reflected, scattered, emitting, etc. from the in-situ cells of a captured scene to determine the molecules of those cells.

Some of these molecules are indicative of certain classes, types, or particular cells. Such molecules are called marker biomolecules herein. The electronic device can determine which cell types or class are present in a captured scene by the on the particular ones of, the types of, and the proportions of the biomolecules detected therein.

In addition, the electronic device may includes or may have one or more environmental sensors that are designed to measure one or more environmental factors associated with the in situ biological cells or the environment surrounding the in situ biological cells.

The electronic device may have or connect to a report system that is designed to report a detection and/or identification of the one or more cell types or classes in the obtained image and in some implementations, the report system is designed to associate the measured environmental factor(s) with the obtained image and/or with the detected cell.

Correlation Engine

The system **460** includes a correlation engine **462**. Based on the image-based data from multiple devices, environmental data from multiple devices, any other associated data, and the results from the report subsystem, the correlation engine **462** finds hidden patterns and ultimately discovers underlying causes of activity (or lack thereof) of biological cells and/or substances. The correlation engine **462** includes one or more AI/ML/DL engines.

Two of the categories of data are supplied to the correlation engine include cellular/molecular observations and environmental observations. One of the categories is based on the cellular and/or molecular measurements/observations of the scene itself. It may be the full conclusion about the detection/identification of the type/class of cells and/or substances found in the scene or something less than the full conclusion.

The other category is any other environmental factor. Of course, there is a myriad of choices and the mass amount of data available here. Because of this, the tools of the so-called Big Data are employed to address this.

Inference Engine

The system **460** includes an inference engine **464**, which may be implemented with AI/ML/DL engines. Based on the image-based data from multiple devices, environmental data from multiple devices, any other associated data, the results from the report subsystem, and the correlations of the correlation engine **462**, the inference engine **464** can draw inferences based on the patterns or links detected by the correlation engine. For example, there may be a direct correlation between ambient humidity and the proliferation of a specific type of biological cell. The inference is that humidity directly affects that type of cell's growth.

If the cell is deemed to be bad, then a human may decide to implement a solution to control the humidity more closely to control the growth more closely. Alternatively, with sufficient automation in place, a computer-controlled HVAC system may be directed to lower the humidity of a room to lessen the chances of the growth of that type of cell.

If strong inferences are formed from the analysis of the data, then more accurate predictions can be made. For example, with sufficient information gathered over a large

enough area in real time, an epidemic of an infectious disease may be detected and its spread predicted so early the spread of the epidemic may be halted long-before the epidemic could take hold.

To further the goal of making better inferences and predictions, the tools of the so-called Big Data may be employed. Big Data is an evolving term that describes the tools used to work with a voluminous amount of structured, semi-structured and unstructured data that has the potential to be mined for information.

Big Data works on the principle that the more you know about anything or any situation, the more reliably you can gain new insights and make predictions about what will happen in the future. By comparing more data points, relationships will begin to emerge that were previously hidden, and these relationships will enable us to learn and inform our decisions.

Most commonly this is done through a process which involves building models, based on the data we can collect, and then running simulations, tweaking the value of data points each time and monitoring how it impacts our results. This process is automated—today's advanced analytics technology will run millions of these simulations, tweaking all the possible variables until it finds a pattern—or an insight—that helps solve the problem it is working on.

Increasingly, data is coming to us in an unstructured form, meaning it cannot be easily put into structured tables with rows and columns. Much of this data is in the form of pictures and videos—from satellite images to photographs uploaded to social networking sites—as well as email and instant messenger communications and recorded telephone calls. To make sense of all of this, Big Data projects often use cutting edge analytics involving artificial intelligence and machine learning. By teaching computers to identify what this data represents—through image recognition or natural language processing, for example—they can learn to spot patterns much more quickly and reliably than humans.

A strong trend over the last few years has been a move towards the delivery of Big Data tools and technology through an “as-a-service” platform. Businesses and organizations rent server space, software systems and processing power from third-party cloud service providers. All of the work is carried out on the service provider's systems, and the customer simply pays for whatever was used. This model is making Big Data-driven discovery and transformation accessible to any organization and cuts out the need to spend vast sums on hardware, software, premises and technical staff.

Distributed Ledgers

With the platform, users may upload and share their own data related to biological tests, experiments, etc. Their data becomes part of the Big Data collection of data and may help form better inferences for others.

As part of this service, the data may be stored in a blockchain fashion to ensure that the data is not altered, deleted, or manipulated. A blockchain is a digitized, decentralized, distributed, public ledger of data transactions. Constantly growing as ‘completed’ blocks (the most recent transactions) are recorded and added to it in chronological order, it allows participants to keep track of data transactions without central recordkeeping. Each node (a computer connected to the network) gets a copy of the blockchain, which is downloaded automatically.

Example Processes

FIG. 5 is a flow diagram illustrating example process **500** that implement the techniques described herein for the data

collection and/or analysis for the detection and/or identification of biological cells and/or biological substances. For example, the example process 500 may detect and/or identify pathobiological cells and/or substances.

The example process 500 may be performed, at least in part, by the electronic device 230, by the system-on-a-chip 301, and/or system 440 as described herein. The example process 500 may be implemented by other electronic devices, computers, computer systems, networks, and the like. For illustration purposes, the example process 500 is described herein as being performed by a "system."

At block 510, the system obtains (e.g., via wireless communication) image-based data from one or more of multiple remote monitoring devices. The image-based data from each monitoring device is based on (e.g., derived from) one or more images of a scene proximate to that monitoring device. The proximate scene includes biological cells and/or substances therein. Indeed, the system may obtain image-based data based on a sequence of images of the scene. If any particular type or class of biological cells and/or substances are detected, then the scene included biological cells and/or substances.

A scene may include, for example, one or more surfaces on which the in-scene biological cells and/or substances inhabit; a liquid in which the in-scene biological cells and/or substances inhabit; a bodily fluid in which the in-scene biological cells and/or substances inhabit; an area in which the in-scene biological cells and/or substances inhabit; a volume in which the in-scene biological cells and/or substances inhabit; an area or volume with its dimensions falling below 0.1 mm; or a combination thereof.

In some implementations, the scene includes biologic cells or biologic substances. But, often, the scene includes both. The in-scene biological cells and/or substances may be characterized as: physically located on a surface; physically located in a medium (e.g., blood, bodily fluids, water, air, etc.); undisturbed in their environment; undisturbed and unadulterated; physically located on a surface in a manner that is undisturbed and unadulterated; not relocated for the purpose of image capture; unmanipulated for the purpose of image capture; or on a surface that is unaffected by the scene-capture system.

In some implementations, the biologic cells and/or substances are in situ and in other implementations, they are in the lab.

The obtained image is micrographic, spectrographic, and/or digital. In some implementations, the obtained image is micrographic because the image of the scene is captured at least in part: on a microscopic scale; using microscope like magnification; includes microscopic structures and features; includes structures and features that are not visible to a naked human eye; or a combination thereof.

In some implementations, the obtained image is spectrographic at least in part because the image of the scene is captured using some portion of the electromagnetic spectrum (e.g., visible spectrum of light, infrared, x-rays, gamma rays, ultraviolet) as it interacts with matter, such interactions include, for example, absorption, emission, scattering, reflection, and refraction.

The image may be obtained by capturing a digital image of the scene, and that scene may include in-scene biological cells and/or substances therein. In addition, digital enhancement of a captured digital image may be employed to better reveal the in-scene biological cells and/or substances in the captured image. The obtained image is digital at least in part because the image of the scene has handled as a set of machine-readable data.

At block 520, the system analyzes the image-based data and/or environmental data. Based on that analysis, the system detects and/or identifies a type or class of biological cells and/or substance amongst the in-scene biological cells and/or substances of each device or amongst several of the devices.

The system may employ an AI/ML/DL engine to perform or assist in the performance of the detection and/or identification of one or more biological cells and/or substances.

The AI/ML/DL engine functionality may be split across the platform. That is, the devices may perform pre-processing of the image using AI/ML/DL engine and send the results as the image-based data to the system 440 for further processing herein. In that case, the device 414 communicates in real time (or nearly so) with the platform. In this way, the intensive processing is offloaded from the devices to the platform.

The biologic detection subsystem 444 analyzes the image-based data of the scene to detect, determine, and/or identify the type or class of biological cells and/or substances therein. It may do this, at least in part, by using distinguishing molecules of the cells that are observable using the electromagnetic spectrum. In some implementations, other data (such as chemical reactions or excitation) may be included in the analysis.

Examples of one or more of the types or classes of biological cells that may be detected and/or identified at block 520 includes (by way of example, but not limitation): cells of a multicell biological organism; cells of a tissue or organ of a multicell biological organism; cells of a tumor or growth multicell biological organism; single-celled organism; microbes; microscopic organisms; single-celled organism; living things that are too small to be seen with a human's naked eye; a biological creature that can only be seen by a human with mechanical magnification; microscopic spores; a combination thereof. In addition, the biological cells have a size range that is selected from a group consisting of: 10-100 nanometers (nm); 10-80 nm; 10-18 nm; 15-25 nm; and 50-150 nm.

Furthermore, biological cells and/or substances may be typed or classified as pathobiological, not pathobiological, pathobiology unknown, or pathobiology not-yet-known. The pathobiological biological cells and/or substances may be classified or typed as (by way of example and not limitation): pathobiological cells; pathobiological substances; toxic; poisonous; carcinogenic; diseased cells; cancer cells; infectious agents; pathogens; bioagents; disease-producing agents; or combination thereof.

Of those biological cells that are characterized as microbes, they may be further typed or classified as one or more of the following (by way of example and not limitation): single-celled organisms; bacteria; archaea; fungi; mold; protists; viruses; microscopic multi-celled organisms; algae; bioagents; spores; germs; prions; a combination thereof.

In some instances, the operation at block 520 may include an identification of one or more particular biologic cells and/or substances in the scene. Rather than just detecting the type or class (e.g., pathogen), the operation may identify the member of that type or class (*listeria*). Examples of particular members of a class or type that this operation may identify include: *Clostridium botulinum*, *Streptococcus pneumoniae*, *Mycobacterium tuberculosis*, *Escherichia coli* o157:h7, *Staphylococcus aureus*, *Vibrio cholerae*, ebola, hiv, influenza virus, norovirus, zika virus, *aspergillus* spp, and *entamoeba histolytica*.

At block **520**, one or more implementations of the detection and/or identification includes operations to:

access a database of signatures of biological cells and/or substances;

isolate a biological cell and/or substance in the obtained image;

correlate the isolated biological cell and/or substance to at least one signature in the database;

determine that the correlation is significant enough to indicate a sufficient degree of confidence to identify the isolated biological cell and/or substance as being a biological cell and/or substance;

in response to that correlation determination, label the isolated biological cell and/or substance as being the determined biological cell and/or substance.

An example of a “sufficient degree of confidence” includes more likely than not. A confidence factor for the “sufficient degree of confidence” may be weighted relative to a perceived degree of threat. For example, a pathogen that is unlikely to cause a human infection may have a very high confidence factor (e.g., 80%). Thus, a detection may only be noted if it is at least 80% likely to be that particular pathogen. Conversely, a pathogen may be particularly dangerous (e.g., small pox) and have only a small confidence factor (e.g., 20%). In this way, the dangerous pathogen is detected even it is more likely that the pathogen was misdetected.

Other implementations of the detection and/or identification include operations to:

provide the obtained image to a trained biological detection/identification (detection and/or identification) engine, the trained biological detection/identification engine being an AI/ML/DL engine trained to detect and/or identify biological cells and/or substances based on a training corpus of signatures of biological cells and/or substances;

receive a positive indication from the biological detection/identification engine that the obtained image includes a biological cell and/or substance therein and/or the identity of that biological cell and/or substance.

Still other implementations of the detection and/or identification include operations to:

provide the obtained image to a trained biological detection/identification engine, the trained biological detection/identification engine being an AI/ML/DL engine trained to detect and/or identify pathobiological cells and/or substances based on a training corpus of signatures of pathobiological cells and/or substances;

receive a positive indication from the biological detection/identification engine that the obtained image includes a pathobiological cell and/or substance therein and/or the identity of that pathobiological cell and/or substance, wherein the obtained image includes data captured from the visible and/or non-visible electromagnetic spectrum of the scene.

Other variations of the detection and/or identification operations described above may be focused on pathobiological cells and/or substances in particular.

At block **530**, from each of the multiple devices, the system gathers one or more environmental factors associated with the in-scene biological cells and/or substances or the environment surrounding the in-scene biological cells and/or substances. The environmental data being based on an environmental factor associated with the in-scene biological cells and/or substances of each device or the environment surrounding the in-scene biological cells and/or substances of each device. In some implementations, the system may acquire information related to or associated with the scene.

The measured environmental factors include (but are not limited to):

temperature; timestamp (e.g., time and date, local time, incremental time, etc.), humidity; barometric pressure; ambient sound; location; ambient electromagnetic activity; ambient lighting conditions; WiFi fingerprint; signal fingerprints; GPS location; airborne particle or chemical detector/counter; gases; radiation; air quality; atmospheric pressure; altitude; Geiger counter; proximity detection; magnetic sensor; rain gauge; seismometer; airflow; motion detection; ionization detection; gravity measurement; photoelectric sensor; piezo capacitive sensor; capacitance sensor; tilt sensor; angular momentum sensor; water-level (i.e., flood) detection; flame detector; smoke detector; force gauge; ambient electromagnetic sources; RFID detection; barcode reading; or a combination thereof.

At block **540**, the device reports a detection and/or identification of the type of biological cell and/or substances in the scene proximate to the monitoring device. For example, the device may provide a report or notification via a user interface to a user. A messaging system (e.g., email or SMS) may be used for such notification.

In some implementations, the system may report that the type of biological cell and/or substances in the obtained image is a category flagged for further research and inquiry. For example, the device may be unable to detect the type of cell or substance. In that case, the device flags this as a something worthy of further inquiry. This category may be the default when there is a failure to detect or identify a cell or substance. In some instances, this category is only triggered with particular markers (e.g., chemicals or structures) are detected.

The report or notification may include the following (by way of example and not limitation) operations: send a communication (e.g., message, postal mail, email, text message, SMS message, electronic message, etc.) to a human or machine that is designated to receive such communications via wired or wireless communications mechanisms; send a notification (e.g., message, postal mail, email, text message, SMS message, electronic message, push notices, etc.) to a human or machine that is designated to receive such notification via wired or wireless communications mechanisms; update a database designated to receive such updates via wired or wireless communications mechanisms; store in memory (e.g., local or remote) the detection; or a combination thereof.

In addition, at block **540**, the device associates the measured environmental factor and/or the associated factor with the obtained image and/or with the detected type or class of biological cell and/or substance. This association may perform in one more database. Indeed, such a database may take the form of a distributed ledger (DL) technology.

As part of block **550**, the report operation may send **449** its results and associated data (image-based data and/or environmental data) to the system **460**.

At block **560**, the system **460** performs correlation, inference, ranking, and/or amelioration operations in response to the results and associated data (image-based data and/or environmental data) sent **449** to system **460**.

Distribution of Scientific Data

The environment and systems can be made available and remove borders to attract the largest network of professional and citizen scientists from around the world to solve the grand challenges of human and environmental health through a sincere, transparent and purpose-driven system.

Using the blockchain and the Coin, an incentive and reward-based economic system can be established to enable the formation of a large aggregated molecular dataset.

Artificial Intelligence, machine learning, and deep learning for healthcare may be used to accelerate development on many fronts. Throughout the industry, biologists and computer scientists are working together to design experiments that incorporate artificial intelligence to identify elements of data that might otherwise be overlooked to predict and treat disease, classify disease types, understand disease subpopulations, find new treatments and match them with the appropriate patients. This is progress, but the model is limiting because it requires us to have access to the right datasets first.

Even with relatively limited aggregated datasets, the application of artificial intelligence, machine learning, and deep learning by computational biologists, researchers, scientists and medical professionals can provide positive outcomes. It is the intent of have the environment and systems described herein to increase the scale of collected and created data by orders of magnitude more and applied it to all global health threats.

The environment and system can accelerate the rate of innovation in human and environmental health by incentivizing the aggregation of global datasets and creating new real-time datasets, enabling scientists, researchers and artificial intelligence systems to make new discoveries on an unprecedented scale.

By converging key health care technologies, a global system can provide a decentralized economy that incentivizes data sharing on a global scale using blockchain and the cryptocurrency or Coin, to enhance the speed and efficiency of medical discovery. The Coin or coins will be minted in real time to convert global data to tangible value.

Although scientists and researchers are taking advantage of all the tools available to them by working hand-in-hand with computer scientists to design experiments that harness the power of artificial intelligence, machine learning and deep learning, but these tools are limited because they require access to the right datasets first. By incentivizing institutions, organizations, governments or individuals globally to pool datasets into one shared ecosystem, we will create an enormous, multi-dimensional, global dataset.

By collecting both research-based and real-world molecular data from both partner institutions and marketplace contributors, all of whom will be compensated for their contributions with the Coin, creating a decentralized economy can be created in pursuit of better health. The blockchain technology means that each piece of data collected is attributed to its correct source, so if a scientist contributes data that is ultimately used by an institution to develop a critical health solution, that scientist will get credit and compensation for their important contribution to the cause, just like a musician would receive royalty compensation for the use of his music.

All of the data contributed to the environment will filter through an artificial intelligence, machine learning and deep learning system, which will be used by marketplace participants to identify patterns, find new correlations and draw conclusions at a scale never before reached. These insights will be added back into the shared database for anyone to access. We firmly believe that by providing the world's brightest minds with this tool, we can dramatically accelerate the speed and efficiency of cure development.

The more data that is collected, the faster and more effective discoveries can be made. Additionally, a particular chip architecture is provided. The chip will have the ability

to collect data in real time, either from real-world environments or laboratories, meaning that all data ranging from a spectroscopy machine in a lab to tabletop germs in a household can be added to the environment in real time.

For example, in a decentralized economy an initial task may be to collect data and build a dataset. Just as the Human Genome Project decreased the cost of sequencing a human genome by roughly one-million fold between 1990-2003, the use of described technologies can decrease the cost and time associated with scientific discovery.

The environment can be used as a tool by professional scientists and institutions around the world to remove borders and encourage collaboration and efficiency around major global health challenges. In addition to these key players in today's science ecosystem, there is another person we hope to empower: the citizen scientist.

It is expected that the environment is open for participation to people around the world, regardless of institutional affiliation or geographical location. A major institution might tap into the platform's database of lab-based and real-world superbug data to develop a critical solution to antimicrobial resistance, or an individual might contribute important molecular data about the conditions and materials that promote black mold growth by placing a chip (device using chip) in their home. Citizen scientists can also participate in the economy simply by providing funding, or purchasing the Coin(s), to target particular disease areas. For example, someone with a family history of Alzheimer's disease might want to purchase Coins to direct funding toward further research to have a tangible part in the mission to find a treatment. The goal is to unleash both the professional scientist from his or her institutional boundaries and encourage the citizen scientist to participate directly in the process of developing cures.

In particular scenarios, the environment exists to accelerate the development of cures for the global health threats that impact all of us. An initial list of key disease areas to target may be created, with new disease areas added in time.

Anyone in the world can contribute to the development of cures (which could include preventive medicine, therapeutics, vaccines or new drug development) for any of these disease areas by simply purchasing Coins and choosing which area to fund. As we continue to add more categories to the list, anyone will be able to have a direct impact on the development of cures for the diseases that matter most to them. Eventually, all global illnesses may be represented.

An example of initial disease targets, which collectively affect billions of people globally, can include: Oncology, Pancreatic cancer, Brain cancer, Virology, Hepatitis C, Influenza, Respiratory syncytial virus, Neglected tropical disease, Bacteriology, Methicillin-resistant *Staphylococcus aureus* (MRSA), *Clostridium difficile* (*C. Diff*), Neurology, Alzheimer's Disease, etc.

Each of these illnesses needs new treatments and effective preventive measures. In today's world, the development of a new treatment takes 10 to 15 years, and even then, 90% of drugs tested in clinical trials fail, effectively resetting the clock.

By mobilizing "citizen scientists" (all parties, including researchers and scientists) contribute to the environment and its economy, the progress of finding solutions and cures can be shortened that timeframe to a fraction of the time and reducing failure rates to ensure that those suffering from these illnesses are considered and cared for.

The environment, system (platform) may use an inference engine, artificial intelligence, machine learning and deep learning. By forming a large global dataset, this continuous

and massive collection of molecular data will enable artificial intelligence, machine learning, and deep learning systems to act as an insight engines by identifying trends and drawing conclusions, giving scientists and researchers an entirely new tool to create life-saving solutions. Use of particular chip architectures, designs, open source libraries, data sets, machine learning, and other technologies allows for an artificial intelligence system's capabilities to derive unique perspectives and insights from this novel, global dataset.

In certain implementations, blockchain is used; blockchain with artificial intelligence; blockchain with artificial intelligence and the chip; and blockchain with artificial intelligence, the chip, and artificial intelligence, machine learning, deep learning. Using the chip, the artificial intelligence system will have access to a very large scale of data that will enable the system's correlations and patterns to drive scientific research priorities and cure development.

The datasets attached to environment and systems have innate value, but also have additional value derived from applied machine learning. The environment enhances the utility of the datasets by attaching contextual metadata, helping the learning routines understand and correlate them. The platform mines for insights within the marketplace by aggressively interrogating the ingest pipeline. Incoming data is passed to individualized handlers which feed them to a series of machine learning algorithms custom-built for the dataset that is processed, exposing new and perhaps unexpected insights. These insights are minted into new, carefully attributed assets on the chain, increasing the marketplace's overall value and offering researchers new avenues for accelerating their work.

By applying deep learning tools such as convolutional neural networks (CNNs) the platform is capable of identifying pathogens within microscopic images and spectrographic signatures that environment and systems ingest from either existing datasets or streams of real time sensor data. By training its neural networks against libraries of high-quality bacteriological images and signatures, the platform can reliably identify specific pathogens. Upon discovery, the platform takes advantage of a sophisticated queueing system to retroactively "replay" historical data with a greatly increased sampling rate, enabling it to build a high-resolution model of the outbreak. This model is then added to the chain, fully secure, attributed and available to researchers who can use it to help contain the outbreak or to advance the understanding of its transmission model.

The chip (chip implemented device) can provide for real-time and real-world data. The chip is used in the data ingest pipeline and marketplace. Deployed throughout a building and/or across a region and using the sensor technology to pick up environmental (e.g. temperature, humidity, etc.), visible, and spectrographic data which, coupled with other (e.g. location, time, etc.) data, the numerous chips in the system can together stream enormous volumes of valuable data into the platform for processing by the artificial intelligence insights engine described above.

The chip will utilize artificial intelligence to detect objects that have already been learned by the platform, minimizing the amount data transmitted to the platform, efficiently utilizing communication bandwidth. The sensed and other data associated with objects that the chip detects, but can't identify, will be sent to the platform which would in will turn trigger an investigation that gathers real-world samples and take those samples to a lab for controlled analysis and identification using machine learning with deep learning. Once identified in the lab, the platform can send a specific

detection algorithm to the chip so that it can then confidently identify the new object going forward.

The chip will either contain or connect to sensors that will enable novel abilities to detect biological objects at low concentrations, in noisy environments (e.g. objects in the midst of dust, dirt, or other uninteresting objects), in real-time (e.g. without significant delay from when object was in field of view), in some cases without disturbing the environment detected objects (i.e. passive observation), and with the ability to search large 3 dimensional spaces so as to reduce the probability of not observing an interesting object which is especially important when the objects are infrequently present. Having some or all of these qualities, coupled with artificial detection algorithms will facilitate detection and identification that is much faster than present technology (typically a sample is collected and cultivated/grown in a lab environment over days), more accurate, and possible outside of a lab environment. Sensors that may be employed to this end are snapshot hyperspectral imaging (to capture a frequency spectrum over many points in a two-dimensional area) and/or light-field/plenoptic imaging (to facilitate the capture, both with w/fast frame rate. A novel combination of both these technologies may be used.

The chip has the ability to interact with the platform, allowing remote configuration of chip-based devices. Some examples of configuration capabilities are a) updating the imbedded code for the chip, b) adjusting sensor configurations and interfaces, 3) downloading improved or new detection algorithms, and d) changing the detection polling duty cycle. In order to facilitate heightened awareness during critical periods, the platform might shorten the polling cycle of devices in affected regions. To optimize power management, the platform might lengthen polling cycles or turn off less pertinent sensors. In order to diminish scalability concerns, the chip will have processing capabilities that will scale the frequency and detail level of outbound communications to the platform.

Vitaly, the chip leverages its sophisticated encryption and anti-spying security capabilities to ensure the ownership and value of the data and/or its privacy are secure. The platform, at the chip and/or elsewhere in the system will further protect the data and provide a marketplace for it using blockchain technology.

The environment and systems make use of distributed ledger of the blockchain. The distributed ledger provides the ability to mint the tokens or coins that the marketplace is built upon, as well as the security needed to safely store the transactions that accompany the buying and selling of healthcare and medical data. The marketplace can be both centralized and decentralized, granting participants control over their own transactions.

In addition to features that environment and systems derive from the blockchain, smart contracts may be implemented as well to ensure secure and immutable attribution. This is a feature that ensures that the author of a dataset is eternally associated with it, and that relationship will persist even if it is bought and used by multiple generations of consumers. This mechanism is part of the incentive system which drives liquidity within the marketplace.

The blockchain can make use of and leverage other blockchain platforms, consensus models, and design models that provide or enable functionality that a) accommodates the needs of its marketplace for certification and attribution of data contributions and transactions, b) support very large numbers of IoT devices powered by its Nano Sense chips contributing a continual real-time stream of molecular data,

environmental, and other data, and c) has the ability to enable users to securely access and leverage a big data system.

The following describes research-based data. Traditional molecular data includes high throughput molecular data such as, Omics: Genomics, Lipidomics, Proteomics, Glycomics, Metabolomics, Transcriptomics Molecular composition: spectroscopy, UV/Vis Circular Dichroism Molecular structure: Mass spectroscopy, Hydrodynamic methods, X-ray crystallography, Nuclear magnetic resonance, Light scattering, Dumas Method, Enzyme-linked immunosorbent assay (ELISA), Immunohistochemistry Molecular Function: Protein affinity chromatography, Co-immunoprecipitation, Fusion of protein with a molecular tag, such as green fluorescent protein (GFP), DNA footprinting, Gene expression. Cellular Metabolism includes: Microplate based assays, ATP assays, Mitochondrial assays, Reactive oxygen species (ROS), oxidative stress, and metabolism assays Cellular Phenotype: Microscopy, Flow cytometry, Western Blot, Genetic screen Measure of Cellular Metabolism.

The following are resulting correlations derived from systems biology. Through the use of experimentally derived data, computational biology and biostatistics, systems biology develops systems, frameworks and models for integrative analysis of omics data, imaging and clinical data to identifying key molecules driving pathophysiology.

Real-world data can include epidemiological data, clinical data such as healthcare databases and clinical data outcomes. Real-world data can also include patient data from patient registries, pharmacy and health insurances databases; social media from patient-powered research networks. Real-world data can also include real-world molecular data from: Chemical Data, Spectroscopy, Imaging, Microscopy, Environmental, Pressure, Humidity, Light, Gas, Temperature, and pH.

The following discusses impact of infectious disease on chronic disease. It is known that various infectious diseases determinately lead to non-communicable chronic diseases such as cancer, immune-mediated syndromes or neurodevelopmental disorders. Advances in epidemiological reporting and research-based molecular technologies have enabled these correlations to be made, but require assumptions due to the size of the data sets used. Recognizing the relationship between infectious and chronic disease states can lead to ground-breaking health advances worldwide.

The environment and systems, and particularly the marketplace enables essential aggregation, scaling and correlation of critical molecular datasets such as genomics, proteomics, transcriptomics and metabolomics and real-world datasets including clinical, epidemiological, and real-world molecular data sets specific to various pathogens chronic disease states. Such a marketplace would enable the merging of currently siloed or individual data pools toward the common effort of chronic and infectious disease prevention, diagnosis and treatment.

Biological systems are complex. When an issue arises, multiple variables should be considered to identify the most effective solutions. While current therapeutic and diagnostic technologies are improving over time, their development is only incremental due to this single target, passive approach. Antimicrobial resistance (AMR) is a severe issue in human and environmental health. When drugs and preventative measures are designed to target specific proteins critical to the health and metabolism of the harmful bacterial cell, the cell can mutate overtime and resist the desired effect. There are multiple mechanisms of AMR, involving a variety of targets within the cell.

The marketplace takes on these complicated issues more holistically, through the value of aggregated lab-based and real-world data at scale, factoring in and connecting all relevant and known markers in a given system. This allows for data to be aggregated, scaled and correlated to enable all constitutes of each critical pathway to be monitored, evaluated and targeted, leading to more robust and effective solutions to this detrimental issue.

Described blocks/components are resident in a distributed network, such as the Internet. The blocks or components, unless otherwise particularly described, are functionally described. In other words, a particular block/component may not necessarily be resident at one or more hardware devices, and may operate over several devices in the network.

Now in reference to FIG. 6, an example platform architecture 600 is shown. An application program interface or API 602 connects chips 604 to a blockchain 606. Data providers 608 through API 602 are also connected to the blockchain 606. A marketplace 610 is connected to the blockchain 606. Chips 604, data providers 608 and marketplace 610 provide data to the blockchain 606. The marketplace 610 provides data to coin holders 612 through API 602.

The blockchain 606 provides data to an ingest pipeline 614, which further provides data to inference engines 616. Inference engines provide data to analytics 618. The analytics 618 provides data to export manager 620. Through the API 602, export manager provides data to data consumers 622. The platform architecture 600 can further include raw storage 624.

Now in reference to FIG. 7, an example system 700 is shown. A Decentralized Authority Organization or DAO 702 is a decentralized authority. Global Citizen Scientists, or any party, are allowed to vote through the DAO 702. Example of voting include determine what percentage of crypto currency (i.e., coin) that they own/control is to be used. For example, percentage of crypto currency used for medical research, non-profit organizations, charities, etc. Various voting algorithms may be implemented by the DAO 702. DAO 702 is connected to a Smart Contracts bucket/database or Smart Contracts 704.

DAO 702 is further connected to blockchain 706. Blockchain 706 provides crypto currency, or tradeable tokens, or as described above, the coin. Example of tradeable tokens include ERC20 defined tokens as provided by the Ethereum network protocol. Blockchain 706 communicates with a Central Data Pool 708. The Central Data Pool 708 also communicates with Smart Contracts 704.

Blockchain 706 communicates with both the DAO 702 and a Centralized Authority Manager 710. Blockchain 706 may be managed by the DAO 702 and the Centralized Authority Manager 710. The Centralized Authority Manager 710 further communicates with Smart Contracts 704, and can create and manage contracts in Smart Contracts 704. Smart Contracts 704 can include hyper ledgers.

The Blockchain 706 communicates with Artificial Intelligence 712. In certain implementations, Artificial Intelligence 712 may be cloud-based and provide/manage data storage.

The system 700 further includes a chip 714. The Chip 714 can be resident in different devices, such as smart phones, laptop/table computers, dedicated medical/research devices, etc. Such devices (i.e., chip) are used for detecting biodata. The devices can also be used for minting coins. The Chip 714 communicates with Blockchain functional block, which will be located within the Chip 714 or elsewhere in the Platform, to mint the data detected by it. In certain imple-

mentations, computing may be performed off-chip via the cloud, or performed as edge computing.

The Chip **714**, and in particular devices implementing the Chip **714**, will use artificial intelligence and all or some of the Chip versions may use artificial intelligence machine learning w/deep-learning (AI/ML/DL), performed by the block Artificial Intelligence, Machine Learning, Deep Learning **716**, to process data. In particular, AI/ML/DL **716** may be used by the Chip **714** in the accelerated processing of bio-data (AI to detect and identification the bio-objects, AI ML w/DL to train and thus optimize AI detection and identification algorithms in the controlled environment of a lab), bio currency, etc. AI/ML/DL **716** may be cloud-based.

FIG. **8** illustrates a system **800** configured that facilitates distribution of scientific data, in accordance with one or more implementations. In some implementations, system **800** may include one or more servers **802**. Server(s) **802** may be configured to communicate with one or more client computing platforms **804** according to a client/server architecture and/or other architectures. Client computing platform(s) **804** may be configured to communicate with other client computing platforms via server(s) **802** and/or according to a peer-to-peer architecture and/or other architectures. Users may access system **800** via client computing platform(s) **804**.

Server(s) **802** may be configured by machine-readable instructions **806**. Machine-readable instructions **806** may include one or more instruction modules. The instruction modules may include computer program modules. The instruction modules may include one or more of data obtaining module **808**, sci-data analysis module **810**, sci-data offering module **812**, sci-data delivery module **814**, sci-data associating module **816**, value determination module **818**, value minting module **820**, cryptocurrency awarding module **822**, and/or other instruction modules.

Data obtaining module **808** may be configured to obtain scientific data from sci-data gathering devices. The sci-data gathering devices may include bio-detection device configured to detect type of biological cells and/or substances proximate to the bio-detection device based on one or more images captured by the bio-detection device. A sci-data gathering device includes electronic instruments and equipment typically used in the gathering or observation of empirical data. This includes instruments and equipment used in medical, laboratory, testing, and research facility. Examples include microscopes, scales, thermometers, glucose meters, etc. The smartdevice **220** and chip **301** are examples of a bio-detection device.

Sci-data analysis module **810** may be configured to analyze gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data. As used herein, a scientific element is a particular piece of data or collection of associated data. The relevance may be determined by similarity to previous elements that were considered relevant. In addition, an AI/DL/ML engine may be used to determine relevance based on a training and/or existing data.

Sci-data offering module **812** may be configured to offer the analyzed sci-data via a marketplace in exchange for a monetary value. The marketplace may be implemented as e-commerce solution. Monetary value refers to a value based on some standard. For example, data may have a monetary value based on the U.S. dollar or the Euro.

Sci-data delivery module **814** may be configured to deliver the analyzed sci-data via the marketplace in exchange for an acceptable monetary value. In some

instance, a user of the marketplace may negotiate for the sci-data. Once there is a agree-upon value, an exchange takes place.

Sci-data associating module **816** may be configured to associate the obtained sci-data to an account. An account may be owed or connected to a person, company, or some legal entity.

Value determination module **818** may be configured to determine the monetary value of analyzed sci-data.

Value minting module **820** may be configured to mint cryptocurrency based on the determined monetary value of the analyzed sci-data. The determined monetary value may be based on previously determined values for similar relevant scientific elements.

Cryptocurrency awarding module **822** may be configured to award the minted cryptocurrency to the account associated with the obtained sci-data.

In some implementations, sci-data may include information collected via observation or experimentation using the scientific method which is the systematic pursuit of knowledge involving the recognition and formulation of a problem and testing of hypotheses. In some implementations, sci-data may include information categorized in accordance with known scientific disciplines (e.g., biology, chemistry, etc.). In some implementations, sci-data may include biological data.

In some implementations, server(s) **802**, client computing platform(s) **804**, and/or external resources **824** may be operatively linked via one or more electronic communication links. For example, such electronic communication links may be established, at least in part, via a network such as the Internet and/or other networks. It will be appreciated that this is not intended to be limiting, and that the scope of this disclosure includes implementations in which server(s) **802**, client computing platform(s) **804**, and/or external resources **824** may be operatively linked via some other communication media.

A given client computing platform **804** may include one or more processors configured to execute computer program modules. The computer program modules may be configured to enable an expert or user associated with the given client computing platform **804** to interface with system **800** and/or external resources **824**, and/or provide other functionality attributed herein to client computing platform(s) **804**. By way of non-limiting example, the given client computing platform **804** may include one or more of a desktop computer, a laptop computer, a handheld computer, a tablet computing platform, a NetBook, a Smartphone, a gaming console, and/or other computing platforms.

External resources **824** may include sources of information outside of system **800**, external entities participating with system **800**, and/or other resources. In some implementations, some or all of the functionality attributed herein to external resources **824** may be provided by resources included in system **800**.

Server(s) **802** may include electronic storage **826**, one or more processors **828**, and/or other components. Server(s) **802** may include communication lines, or ports to enable the exchange of information with a network and/or other computing platforms. Illustration of server(s) **802** in FIG. **8** is not intended to be limiting. Server(s) **802** may include a plurality of hardware, software, and/or firmware components operating together to provide the functionality attributed herein to server(s) **802**. For example, server(s) **802** may be implemented by a cloud of computing platforms operating together as server(s) **802**.

Electronic storage **826** may comprise non-transitory storage media that electronically stores information. The electronic storage media of electronic storage **826** may include one or both of system storage that is provided integrally (i.e., substantially non-removable) with server(s) **802** and/or removable storage that is removably connectable to server(s) **802** via, for example, a port (e.g., a USB port, a firewire port, etc.) or a drive (e.g., a disk drive, etc.). Electronic storage **826** may include one or more of optically readable storage media (e.g., optical disks, etc.), magnetically readable storage media (e.g., magnetic tape, magnetic hard drive, floppy drive, etc.), electrical charge-based storage media (e.g., EEPROM, RAM, etc.), solid-state storage media (e.g., flash drive, etc.), and/or other electronically readable storage media. Electronic storage **826** may include one or more virtual storage resources (e.g., cloud storage, a virtual private network, and/or other virtual storage resources). Electronic storage **826** may store software algorithms, information determined by processor(s) **828**, information received from server(s) **802**, information received from client computing platform(s) **804**, and/or other information that enables server(s) **802** to function as described herein.

Processor(s) **828** may be configured to provide information processing capabilities in server(s) **802**. As such, processor(s) **828** may include one or more of a digital processor, an analog processor, a digital circuit designed to process information, an analog circuit designed to process information, a state machine, and/or other mechanisms for electronically processing information. Although processor(s) **828** is shown in FIG. **8** as a single entity, this is for illustrative purposes only. In some implementations, processor(s) **828** may include a plurality of processing units. These processing units may be physically located within the same device, or processor(s) **828** may represent processing functionality of a plurality of devices operating in coordination. Processor(s) **828** may be configured to execute modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822**, and/or other modules. Processor(s) **828** may be configured to execute modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822**, and/or other modules by software; hardware; firmware; some combination of software, hardware, and/or firmware; and/or other mechanisms for configuring processing capabilities on processor(s) **828**. As used herein, the term “module” may refer to any component or set of components that perform the functionality attributed to the module. This may include one or more physical processors during execution of processor readable instructions, the processor readable instructions, circuitry, hardware, storage media, or any other components.

It should be appreciated that although modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822** are illustrated in FIG. **8** as being implemented within a single processing unit, in implementations in which processor(s) **828** includes multiple processing units, one or more of modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822** may be implemented remotely from the other modules. The description of the functionality provided by the different modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822** described below is for illustrative purposes, and is not intended to be limiting, as any of modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822** may provide more or less functionality than is described. For example, one or more of modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822** may be eliminated, and some or all of its functionality may be provided by other ones of modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822**. As another example, processor(s) **828** may be configured to execute one or more additional modules that may

perform some or all of the functionality attributed below to one of modules **808**, **810**, **812**, **814**, **816**, **818**, **820**, and/or **822**.

FIGS. **9A** and/or **9B** illustrates a method **900** that facilitates distribution of scientific data, in accordance with one or more implementations. The operations of method **900** presented below are intended to be illustrative. In some implementations, method **900** may be accomplished with one or more additional operations not described, and/or without one or more of the operations discussed. Additionally, the order in which the operations of method **900** are illustrated in FIGS. **9A** and/or **9B** and described below is not intended to be limiting.

In some implementations, method **900** may be implemented in one or more processing devices (e.g., a digital processor, an analog processor, a digital circuit designed to process information, an analog circuit designed to process information, a state machine, and/or other mechanisms for electronically processing information). The one or more processing devices may include one or more devices executing some or all of the operations of method **900** in response to instructions stored electronically on an electronic storage medium. The one or more processing devices may include one or more devices configured through hardware, firmware, and/or software to be specifically designed for execution of one or more of the operations of method **900**.

FIG. **9A** illustrates method **900**, in accordance with one or more implementations.

An operation **902** may include obtaining scientific data from sci-data gathering devices. Operation **902** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to data obtaining module **808**, in accordance with one or more implementations.

An operation **904** may include analyzing gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data. Operation **904** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to sci-data analysis module **810**, in accordance with one or more implementations.

An operation **906** may include offering the analyzed sci-data via a marketplace in exchange for a monetary value. Operation **906** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to sci-data offering module **812**, in accordance with one or more implementations.

An operation **908** may include delivering the analyzed sci-data via the marketplace in exchange for an acceptable monetary value. Operation **908** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to sci-data delivery module **814**, in accordance with one or more implementations.

FIG. **9B** illustrates method **900**, in accordance with one or more implementations.

An operation **910** may include associating the obtained sci-data to an account. Operation **910** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to sci-data associating module **816**, in accordance with one or more implementations.

An operation **912** may include determining monetary value of analyzed sci-data. Operation **912** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as

or similar to value determination module **818**, in accordance with one or more implementations.

An operation **914** may include minting cryptocurrency based on the determined monetary value of the analyzed sci-data. Operation **914** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to value minting module **820**, in accordance with one or more implementations.

An operation **916** may include awarding the minted cryptocurrency to the account associated with the obtained sci-data. Operation **916** may be performed by one or more hardware processors configured by machine-readable instructions including a module that is the same as or similar to cryptocurrency awarding module **822**, in accordance with one or more implementations.

Glossary

The following is a list of relevant terms used herein. Unless the context in which the term is used indicates differently, the terms of this glossary may be understood as being described in this glossary in accordance with the technology described herein.

Electronic Device: An apparatus that includes one or more electronic components designed to control or regulate the flow of electrical currents for the purpose of information processing or system control. An electronic device may include some mechanical, optical, and/or otherwise non-electronic parts in addition to its electronic components. Examples of such electronic components include transistors, diodes, capacitors, integrated circuits, and the like. Often such devices have one or more processors that are capable of executing instructions, memories, input/output mechanisms (e.g., display screens, touchscreens, cameras, etc.), and communication systems (e.g., wireless networking and cellular telephony). Examples of an electronic device contemplated herein includes a smartphone, a tablet computer, medical equipment, a microscope, a smartdevice, a computer, a standalone unit, a collection of cooperative units, a button-sized unit, system-on-a-chip, a device on a chip, an accessory to a smartphone or smartdevice, an ambulatory device, a robot, swallowability device, an injectable device, or the like. Depending on the implementation, the electronic device may be characterized as: portable; handheld; fits into a typical pocket; lightweight; portable and with fixed (non-aimable) optics—thus, the device must be moved to aim the optics; with aimable optics—thus, the device need not be moved to aim the optics; or a combination thereof. In addition, an implementation of an electronic device may be characterized as a smartdevice (e.g., smartphone or tablet computer) with its own processor and camera (as its scene-capture system); an accessory or case for a smartdevice that operatively attaches to the smartdevice and adds additionally processing capabilities and functionalities for its scene-capture system; stand-alone device with its own processor and camera (as its scene-capture system); ambulatory device that can move under its own power; a device-on-a-chip; system-on-a-chip; or a wireless device that is configured to interconnect with a wireless network of such devices, this device has its own processor camera (as its scene-capture system).

System: An assemblage or combination of things, parts, or components that form a unitary or cooperative whole. In some instances, a system and platform are used synonymously.

Scene: An area, place, location, scenario, etc. that is in view of the scene-capture system.

Image: An array (e.g., two-dimensional) of data derived from and mapped to a scene. An image may be an array of measured data regarding the electromagnetic spectrum emanating from, reflected off, passing through, scattering off of, etc. the contents (e.g., matter) of the scene. The image has an inherent frame or bound around or surrounding the subject scene.

In situ: Describes something that is situated in the original, natural, or existing place or position. Something that is in place or position. It is undisturbed.

In-the-field: A synonym for in situ.

In the lab: Describes the opposite of in situ. That is, it describes something that has been removed from its original or natural place or position. It is something that is not in place. It has been repositioned.

Biological cell: In biology, a cell is the basic structural, functional, and biological unit of all known living organisms. Typically, biological cells consist of cytoplasm enclosed within a membrane, which contains many biomolecules such as proteins and nucleic acids. Organisms can be classified as single-celled or unicellular (consisting of a single cell; including bacteria) or multicellular (such as plants and animals). While the multicellular plants and animals are often visible to the unaided human eye, their individual cells are visible only under a microscope, with dimensions between 1 and 100 micrometers.

Biological substance: As used herein, a biological substance is not itself a biological cell. Rather, this is a substance is strongly associated with biological cells or lifeforms. In particular, a biological substance may be part of or produced by a biological cell or lifeform. In other instances, a biological substance is capable of affecting a lifeform (or some portion thereof).

Biological cells and/or substances: As used herein, this refers to both “biological cells” and “biological substances.”

Type or class of biological cell: The cells may be classified, categorized, or typed based on identifiable characteristics (e.g., physical, chemical, behavioral, etc.). For example, some cells may be classified as pathological because they cause disease. Some may be a diseased typed because they are malfunctioning and/or infected.

Micrographic: An image is classified as micrographic when it captures content that is on a microscopic scale. Such content includes things that are less than 100 micrometers in size. More generally, it includes items smaller than a macroscopic scale (which are visible to the unaided human eye) and quantum scale (i.e., atomic and subatomic particles).

Spectrographic: An image is classified as spectrographic when it captures the interaction between matter and some portion of the electromagnetic spectrum. Examples of such interactions included absorption, emission, scattering, reflection, refraction, translucency, etc.

Optical: Physics that involves the behavior and properties of light, including its interactions with matter and instruments that use or detect it. However, optics involve more than just the visible spectrum.

Visible Spectrum: This is part of the spectrographic image but specifically includes some portion of the visible spectrum (i.e., light) and excludes the non-visible portions.

Digital: This describes data that is formatted and arranged so as to be managed and stored by a machine, computer, digital electronic device, or the like. A data in the form of a digital signal uses discrete steps to transfer information.

Disease: Any disordered or malfunctioning lifeform or some portion thereof. A diseased lifeform is still alive but ill, sick, ailing, or the like.

Pathological: Something that is capable of causing disease or malfunction in a lifeform (or a portion thereof). A pathogen is pathological.

Pathobiological: Something is pathobiological if it is either capable of causing the disease to a lifeform (or some portion thereof) or is a diseased lifeform (or some portion thereof).

Pathobiological cell: A biological cell that is pathobiological.

Pathobiological substance: This is a substance that is either capable of causing the disease to a lifeform (or some portion thereof) or is associated with a diseased lifeform (or some portion thereof). The substance is not itself a biological cell.

Pathobiological cells and/or substances: As used herein, the term “pathobiological” modifies both “cell” and “substance.”

Pathogen: A biological cell (e.g., unicellular organism) that is capable of causing a disease. More generally, anything that can cause or produce disease.

Diseased cell: A biological cell (e.g., cancerous cell) that is alive but diseased.

Lifeform: The body form that characterizes an organism. Examples of lifeforms include:

Plants—Multicellular, photosynthetic eukaryotes

Animals—Multicellular, eukaryotic organisms

Fungus—Eukaryotic organisms that include microorganisms such as yeasts and molds

Protists—a Eukaryotic organism that is not an animal, plant or fungus

Archaea—Single-celled microorganisms

Bacteria—Prokaryotic microorganisms

Organism: An organism may generally be characterized as containing different levels of the organization; utilizing energy; responding to stimuli/environment; maintaining homeostasis; undergoing metabolism; growing; reproduction; and adapting to its environment.

Environmental factor: It is anything measurable that is capable of affecting the scene or that is associated with the scene. Such things can be abiotic or biotic. Abiotic factors include, for example, ambient temperature, moisture, humidity, radiation, the amount of sunlight, and pH of the water medium (e.g., soil) where a microbe lives. Examples of biotic factors include the availability of food organisms and the presence of conspecifics, competitors, predators, and parasites.

Smartphone: Generally, this term refers to a portable electronic device with features that are useful for mobile communication and computing usage. Such features include the ability to place and receive voice/video calls, create and receive text messages, an event calendar, a media player, video games, GPS navigation, digital camera and video camera.

Smartdevice: The concept of a smartdevice includes a smartphone, but it also includes any other portable electronic device that might not have all of the features and functionality of a smartphone. Examples of a smartdevice include a tablet computer, portable digital assistant, smart watches, fitness tracker, location trackers, a so-called internet-of-things device, and the like. A smartdevice is an electronic device that is generally connected to other devices or networks via different wireless protocols such as Bluetooth, NFC, Wi-Fi, 3G, etc., that can operate to some extent interactively and autonomously.

Accessory: As used herein, this is an accessory to an electronic device (such as a smartphone or smartdevice). It adds additional functionality and/or capabilities to the electronic device. Examples of such accessories include a smart-watch or electronically enabled phone case.

Additional and Alternative Implementation Notes

In the above description of example implementations, for purposes of explanation, specific numbers, materials configurations, and other details are set forth in order to better explain the present disclosure. However, it will be apparent to one skilled in the art that the subject matter of the claims may be practiced using different details than the examples ones described herein. In other instances, well-known features are omitted or simplified to clarify the description of the example implementations.

The terms “techniques” or “technologies” may refer to one or more devices, apparatuses, systems, methods, articles of manufacture, and/or executable instructions as indicated by the context described herein.

As used in this application, the term “or” is intended to mean an inclusive “or” rather than an exclusive “or.” That is, unless specified otherwise or clear from context, “X employs A or B” is intended to mean any of the natural inclusive permutations. That is, if X employs A; X employs B; or X employs both A and B, then “X employs A or B” is satisfied under any of the foregoing instances. In addition, the articles “a” and “an” as used in this application and the appended claims should generally be construed to mean “one or more,” unless specified otherwise or clear from context to be directed to a singular form.

These processes are illustrated as a collection of blocks in a logical flow graph, which represents a sequence of operations that may be implemented in mechanics alone, with hardware, and/or with hardware in combination with firmware or software. In the context of software/firmware, the blocks represent instructions stored on one or more non-transitory computer-readable storage media that, when executed by one or more processors or controllers, perform the recited operations.

Note that the order in which the processes are described is not intended to be construed as a limitation, and any number of the described process blocks can be combined in any order to implement the processes or an alternate process. Additionally, individual blocks may be deleted from the processes without departing from the spirit and scope of the subject matter described herein.

The term “computer-readable media” is non-transitory computer-storage media or non-transitory computer-readable storage media. For example, computer-storage media or computer-readable storage media may include, but are not limited to, magnetic storage devices (e.g., hard disk, floppy disk, and magnetic strips), optical disks (e.g., compact disk (CD) and digital versatile disk (DVD)), smart cards, flash memory devices (e.g., thumb drive, stick, key drive, and SD cards), and volatile and non-volatile memory (e.g., random access memory (RAM), read-only memory (ROM)).

What is claimed is:

1. A system configured that facilitates distribution of scientific data, the system comprising:

one or more hardware processors configured by machine-readable instructions to:

obtain scientific data from sci-data gathering devices; analyze through artificial intelligence, machine learning, and deep learning (AI/ML/DL) technology implemented by AI/ML/DL engines, gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data;

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offer the analyzed sci-data via a marketplace in exchange for a monetary value;
 deliver the analyzed sci-data via the marketplace in exchange for an acceptable monetary value;
 associate the obtained sci-data to an account;
 determine monetary value of analyzed sci-data;
 mint cryptocurrency based on the determined monetary value of the analyzed sci-data; and
 award the minted cryptocurrency to the account associated with the obtained sci-data.

2. The system of claim 1, wherein sci-data includes information collected via observation or experimentation using the scientific method which is the systematic pursuit of knowledge involving the recognition and formulation of a problem and testing of hypotheses.

3. The system of claim 1, wherein sci-data includes information categorized in accordance with known scientific disciplines.

4. The system of claim 1, wherein sci-data includes biological data.

5. The system of claim 1, wherein the sci-data gathering devices include bio-detection device configured to detect type of biological cells and/or substances proximate to the bio-detection device based on one or more images captured by the bio-detection device.

6. The system of claim 1, wherein the determined monetary value is based on previously determined values for similar relevant scientific elements.

7. A method that facilitates distribution of scientific data, the method comprising:

obtaining scientific data from sci-data gathering devices; analyze through artificial intelligence, machine learning, and deep learning (AI/ML/DL) technology implemented by AI/ML/DL engines, gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data;

offering the analyzed sci-data via a marketplace in exchange for a monetary value;

delivering the analyzed sci-data via the marketplace in exchange for an acceptable monetary value;

associating the obtained sci-data to an account;

determining monetary value of analyzed sci-data;

minting cryptocurrency based on the determined monetary value of the analyzed sci-data; and

awarding the minted cryptocurrency to the account associated with the obtained sci-data.

8. The method of claim 7, wherein sci-data includes information collected via observation or experimentation using the scientific method which is the systematic pursuit of knowledge involving the recognition and formulation of a problem and testing of hypotheses.

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9. The method of claim 7, wherein sci-data includes information categorized in accordance with known scientific disciplines.

10. The method of claim 7, wherein sci-data includes biological data.

11. The method of claim 7, wherein the sci-data gathering devices include bio-detection device configured to detect type of biological cells and/or substances proximate to the bio-detection device based on one or more images captured by the bio-detection device.

12. The method of claim 7, wherein the determined monetary value is based on previously determined values for similar relevant scientific elements.

13. A non-transient computer-readable storage medium having instructions embodied thereon, the instructions being executable by one or more processors to perform a method that facilitates distribution of scientific data, the method comprising:

obtaining scientific data from sci-data gathering devices; analyze through artificial intelligence, machine learning, and deep learning (AI/ML/DL) technology implemented by AI/ML/DL engines, gathered sci-data to identify and categorize relevant scientific elements in the gathered sci-data;

offering the analyzed sci-data via a marketplace in exchange for a monetary value;

delivering the analyzed sci-data via the marketplace in exchange for an acceptable monetary value;

associating the obtained sci-data to an account;

determining monetary value of analyzed sci-data;

minting cryptocurrency based on the determined monetary value of the analyzed sci-data; and

awarding the minted cryptocurrency to the account associated with the obtained sci-data.

14. The computer-readable storage medium of claim 13, wherein sci-data includes information collected via observation or experimentation using the scientific method which is the systematic pursuit of knowledge involving the recognition and formulation of a problem and testing of hypotheses.

15. The computer-readable storage medium of claim 13, wherein sci-data includes information categorized in accordance with known scientific disciplines.

16. The computer-readable storage medium of claim 13, wherein sci-data includes biological data.

17. The computer-readable storage medium of claim 13, wherein the sci-data gathering devices include bio-detection device configured to detect type of biological cells and/or substances proximate to the bio-detection device based on one or more images captured by the bio-detection device.

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