•REVIEW•



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Natural products for infectious microbes and diseases: an overview of sources, compounds, and chemical diversities

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As coronavirus disease 2019 (COVID-19) threatens human health globally, infectious disorders have become one of the most challenging problem for the medical community. Natural products (NP) have been a prolific source of antimicrobial agents with widely divergent structures and a range of vast biological activities. A dataset comprising 618 articles, including 646 NP-based compounds from 672 species of natural sources with biological activities against 21 infectious pathogens from five categories, was assembled through manual selection of published articles. These data were used to identify 268 NP-based compounds classified into ten groups, which were used for network pharmacology analysis to capture the most promising lead-compounds such as agelasine D, dicumarol, dihydroartemisinin and pyridomycin. The distribution of maximum Tanimoto scores indicated that compounds which inhibited parasites exhibited low diversity, whereas the chemistries inhibiting bacteria, fungi, and viruses showed more structural diversity. A total of 331 species of medicinal plants with compounds against *C. neoformans*, the family Ancardiaceae has compounds against *Salmonella typhi*, the family Cucurbitacea against the human immunodeficiency virus (HIV), and the family Ancistrocladaceae against *Plasmodium*. This review summarizes currently available data on NP-based antimicrobials against refractory infections to provide information for further discovery of drugs and synthetic strategies for anti-infectious agents.

natural product, infectious pathogen, drug discovery, drug development, in silico analysis

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Introduction

Infectious diseases are a significant challenge to public

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health, producing the second highest number of deaths from disease globally. Natural products (NP), which have properties evolutionarily optimized for different biological functions (Atanasov et al., 2015), have been a prolific source of antimicrobial agents with widely divergent structures and

biological activities (Atanasov et al., 2015). With the advent of combinatorial chemistry and high-throughput screening (HTS) over the last few decades (Atanasov et al., 2015), there has been a surge in the identification of new molecular architectures and their precise mechanism of action (MOA), including an ever-increasing number of NPs (Teijaro et al., 2018). Currently, approximately 60% of approved smallmolecule medicines are related to NPs, and 69% of all antibacterial agents are derived from NPs (Zuo et al., 2012). Antimalarial drugs such as quinine and artemisinin are examples of widely used, effective drugs derived from NPs (Chen et al., 2020).

The plant kingdom serves as a unique resource for antimicrobial compounds (Atanasov et al., 2015). The first written records of plant medicine date from 2,600 BC, in Egypt (Cragg and Newman, 2013). Similar to traditional Chinese medicine (TCM), which had systematically been applied for over two thousand years (Winder, 1988), the documentation of the Indian Ayurveda system dates back to the first millennium BC (Patwardhan, 2005). The herbal medicine of Europe is largely derived from Greek and Roman medicine. The Arabs preserved many Greco-Roman medical books during the Dark and Middle Ages in the fifth to twelfth centuries (Cragg and Newman, 2013). The search for antimicrobial agents, in particular, has largely been focused on the exploration of NPs, as almost all the antibiotic scaffolds were derived from natural sources (Newman and Cragg, 2016). It has been estimated that 350 agents, including NPs, semi-synthetic antibiotics, and synthetic chemicals, have so far reached the world market as antimicrobials (Salam and Quave, 2018). The successful discovery of artemisinin was due to the exploration of antique Chinese herbal medicine books.

The adaptive capability of plant eukaryotic antibiotics is still uncharted territory, although these antibiotics have historically been used against viral, bacterial, fungal, and parasitic infections (Dvorkin-Camiel and Whelan, 2008). Microorganisms are a prolific source of structurally diverse bioactive medicines. The discovery of penicillin from Penicillium notatum by Alexander Fleming in 1928 marked a significant shift from plants to microorganisms as a source of natural antimicrobial agents (Atanasov et al., 2015). During the following decades, various bioactive NPs, including quinine, caffeine, nicotine, codeine, atropine, colchicine, cocaine, and capsaicin, were isolated from their natural sources (Atanasov et al., 2015). The twentieth century witnessed the development of compounds with apparently miraculous properties, mainly in the form of antibiotics (Dvorkin-Camiel and Whelan, 2008). Many of these drugs were developed from diverse groups of fungi, in which twothirds of antibiotics are produced by members of the bacterial order actinomycetales (Cragg and Newman, 2013). The marine environment is another rich source of bioactive compounds, many of which belong to totally novel chemical classes, not previously discovered in terrestrial sources (Akram et al., 2018). During the decade from 1977 to 1987, about 2,500 new metabolites were reported from various marine organisms, and in 2010 alone, 1,003 new compounds were published (Blunt et al., 2012).

NPs continue to serve as essential sources of chemical entities, supporting drug discovery for infectious diseases. A dataset comprising a relatively complete list of NPs which have been used against five categories of major infectious pathogens has been assembled through a manual selection of published articles (Pubmed, Google Scholar, Web of Science, and CNKI). In this review we use this resource to discover NP-based antimicrobials which may be active against refractory infections, attempt to predict the future for drug discovery in this arena, and endeavor to spark the curiosity of investigators who might enter this fascinating, complex landscape of NPs and their interactions with infection.

Classifications of infectious pathologies in conventional and traditional medicines

When infectious pathogens enter the human body, non-specific and specific immunity protects the body from the attack (Reller et al., 2001). Phagocytosis, a typical non-specific mechanism of protection, induced by neutrophils and macrophages, occurs. The remaining pathogens are cleared as completely as possible by antibody-mediated humoral immunity and cell-mediated immunity, with the help of other immune molecules (Reller et al., 2001). Immune reactions fall into two opposite yet cooperative systems, just like the Yin and Yang conception from Daoism in the basic theory of TCM: immune initiation and response (Yang), participated by activators and effectors such as T cells (Th₁, Th₁₇), B cells, DC, M ϕ , NK cells, NKG2D, and IFN- γ ; and immune regulation and tolerance (Yin), managed by regulators and controllers such as IL-10, NKG2A, Treg, Th2, Breg, DCreg, and M2. The detailed strategy of the human immune system against pathogens changes with the characteristics of each pathogen, from virus to parasite. The classification of infectious pathologies in conventional medicine is pathogenoriented, and usually includes viruses, bacteria, fungus, parasites, and chlamydia and mycoplasmas. We selected and summarized potential NP candidates for severe and refractory pathologies caused by infectious pathogens from these five categories.

From the perspective of TCM, infectious pathologies occur as a result of macroclimate or physical changes, leading to imbalance of the Yin-Yang of the body, followed by the development of various symptoms. Infections are caused by an external pathogenic attack-pestilent Qi or evil Qi-due to

climatic or environmental changes. A weakened immune system results in disease arising from the battle between healthy Qi (immunity) and pathogenic Qi (pathogens) (Luo et al., 2020). The former, functioning as a defending barrier, basically consists of four layers, from the exterior to the interior: defensive Oi, Oi, nutrient Oi, and Blood, Therefore in TCM, the classification of infections is always symptomoriented. As mentioned in Volume Plain Questions, Chapter Discussion on Acupuncture, Inner Cannon of Yellow Emperor (Huang Di Nei Jing), "all the five categories of pestilences are contagious in a similar way with similar symptoms" (Yao, 2010). Here "five categories of pestilences" refer to cold pestilence (with symptoms manifesting cold), warm pestilence (with symptoms manifesting heat), larvnx infection, infectious dysentery, and parasite infection, manifested by different syndromes. The reaction between antigen and antibody, or evil Qi and healthy Qi, induced by external attack, happens to correspond well in the two systems of medicines (Ma, 1981). The discovery of artemisinin is one typical example from the Chinese herbal database, built based on thousands of years of clinical experience (Figure 1).

Data sources and mining of NP against five categories of infectious pathogens

Both English databases, including PubMed (Canese and Weis, 2013), Google Scholar (Vine, 2006), and Web of Science (Analytics, 2017), and Chinese databases, including CNKI (Xia et al., 2008) and Wanfang (Wang and Shi, 2012) were searched to collect currently available literature. We conducted activity-based selection on NPs against 21 infectious pathogens covering five categories: viruses (hepatitis B virus (HBV), influenza virus (IAV), herpes simplex virus (HSV), human immunodeficiency virus (HIV), severe acute respiratory syndrome coronavirus (SARS-CoV), Middle East respiratory syndrome coronavirus (MERS-CoV), and severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2); bacteria (Salmonella typhi, Shigella, Mycobacterium tuberculosis, methicillin-resistant staphylococcus aureus (MRSA), Streptococcus pneumonia, and Helicobacter pylori); fungi (Cryptococcus neoformans and Malassezia furfur); parasites (Plasmodium, Schistosoma, Trichomonas vaginalis); and other categories (Leptospira, Treponema Pallidum, and Prion). All included articles were classified into five levels of evidence regarding the robustness of data and experiments conducted. Articles containing only in vitro experiments, conducted or primarily screened by molecular docking, were considered Level One evidence; those with in vitro experiments plus animal model testing or the identification of specific molecular interactions were considered Level Two; those involving signal pathways were considered Level Three; those involving examination of molecular effects by gene knockdown were considered Level Four; and those including molecular docking or other experiments to test the direct binding of the target based on pharmacodynamics and mechanism studies, were considered Level Five.

A total of 618 articles were collected. These articles included 646 NP-based compounds, from 672 natural sources, with biological activities against 21 pathogenic organisms, from five different categories. Of the articles, 91.59% (566) reported in vitro experiments, 2.75% (17) reported a combination of *in vivo* and *in vitro* research, 2.10% (13) were *in* vivo studies, 1.29% (8) were in silico studies, and 0.16% (1) was an in vitro study with in silico screening. With respect to level of evidence, 85.11% (526) of the studies available were categorized as providing level one evidence, 11.33% (70) provided level two evidence, 2.75% (17) had level-three evidence, 0.16% (1) had level-four evidence, and 0.64% (4) provided level-five evidence, indicating that most of the current studies into NPs are still at the preliminary stage, and the MOA and specific targets should be a focus of further exploration (Supplementary materials in Supporting Information).

We selected data from some of the most representative pathogens from each category for further discussion.

NPs against viruses

There are numerous, potentially useful, NPs to be evaluated and exploited for therapeutic applications against different virus families (Kitazato et al., 2007). Three major mechanisms of inhibition of HBV by compounds from natural sources have been reported: the inhibition of HBV covalently closed circular DNA (cccDNA) (Wang, 2008), regulation of HBV replication, and inhabitation of HBV replication by modifying JNK/MIAPK signal pathway. Studies have shown that some effective components from medicinal plants can inhibit the replication of HBV DNA as well as cccDNA (Su et al., 2017). Epigallocatechin gallate, a catechin compound extracted from green tea, has an antiviral effect on the HBV tolerant cell line Hep G2.117, and has an inhibitory effect on pre-core mRNA, cccDNA, and replication intermediate DNA, which are involved in HBV expression (Wang, 2008).

Treatments for acquired immune deficiency syndrome (AIDS) generally fall into the categories of HIV suppression or HIV reactivation (Cary and Peterlin, 2018). A promising area of research involves a deeper understanding of the identification of drug targets which prevent HIV transcription is the goal to achieve a functional cure (Chang et al., 2011; Deeks, 2012). These HIV reverse transcriptase inhibitors include *biflavonoids from Rhus succedanea* (Lin et al., 1997), michellamines from *Ancistrocladus korupensis* (Boyd et al., 1994), and lanostane-type triterpenes from

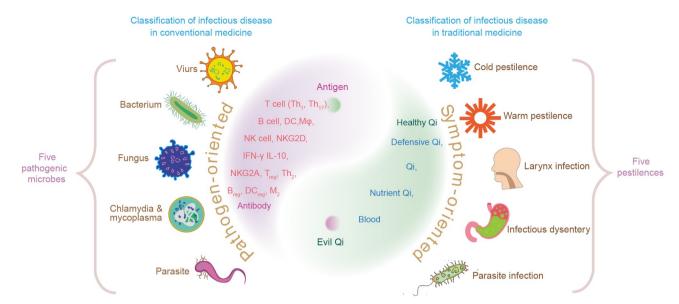


Figure 1 Classification of infectious diseases in conventional and traditional medicines. Left panel, The classification of infectious pathologies in conventional medicine is pathogen-oriented, including five categories: viruses, bacteria, fungi, parasites, and chlamydia and mycoplasmas. The immune reactions could be generally elucidated by two opposite yet cooperative systems, just like the Yin and Yang conception from Daoism in the basic theory of TCM: immune initiation and response (Yang), involving activators and effectors such as T cells (Th₁, Th₁₇), B cells, DC, M ϕ , NK cells, NKG2D, and IFN- γ ; and immune regulation and tolerance (Yin), managed by regulators and controllers such as IL-10, NKG2A, T_{reg}, Th2, B_{reg}, DC_{reg}, and M2. Right panel, From the perspective of traditional medicine, the classification of infections is always symptom-oriented. The diseases occur due to imbalance between healthy Qi (immunity) and pathogenic Qi (pathogen). The *Inner Cannon of Yellow Emperor (Huang Di Nei Jing)* classified five categories of pestilences, cold pestilence, warm pestilence, larynx infection, infectious dysentery, and parasite infection, manifested by different syndromes. The recovery from diseases depends on the battle between healthy Qi (immunity) and pathogenic Qi (pathogen). The former, functioning as a defending barrier, basically consists of four layers from the exterior to the interior: defensive Qi, Qi, nutrient Qi, and Blood.

Polyalthia suberosa (Li et al., 1993). Three of the active compounds identified are known to be HIV integrase inhibitors: flavonoid gallate ester from *Acer okamotaanum* (Kim et al., 1998), and dicaffeoylquinic acids from *Achyrocline satureioides* (Zhu et al., 1999). Some active compounds were found to be HIV protease inhibitors: watersoluble lignins from *Inonotus obliquus* (Ichimura et al., 1998), uvaol and ursolic acid from *Crataegus pinatifida* (Min et al., 1999), and maslinic acid from *Geum japonicum* (Xu et al., 1996; Chisembu and Hedimbi, 2009).

At present, no preventive vaccines or established antiviral therapies are available for coronaviruses (Sohrabi et al., 2020). However, several synthetic compounds have shown promise, including hydroxychloroquine and choloroquine phosphate (Cortegiani et al., 2020), which act through several mechanisms, including alkalinization of the host cell phagolysosomes. Despite the short time since the emergence of SARS-CoV-2 several studies have reported on the use of computer modeling for screening for the inhibition of this virus (Liu and Zhou, 2005). Typically, these models determine the free binding of energy between a ligand and a receptor, with a lower free binding energy indicating a stronger bond between the ligand and the receptor (Lung et al., 2020). Several researchers have utilized virtual computer docking models to screen for potential compounds that could bind to and inhibit key proteins present in SARS-CoV (Liu and Zhou, 2005), highlighting the potential antiviral activity

of compounds such as sabadinine and aurantiamide acetate. Only a handful of studies have investigated the potential of NPs as therapeutic agents against MERS-CoV (Richardson et al., 2020). Silvestrol, a phytochemical from *Aglaia* sp., has been found to be a potent inhibitor of MERS-CoV replication (EC_{50} of 1.3 nmol L⁻¹) (Müller et al., 2018). Some promising compounds for the treatments of coronavirus include scutellarein, Silvestro L, tryptanthrin, saikosaponin B₂, griffithsin, lycorine, and polyphenolics (Müller et al., 2018).

NPs against bacteria

NPs have played a vital role in the development of antimicrobial drugs, such as discovering of rifamycins (1957), quinolone (1962), trimethoprim (1968), oxazolidinone (2000), and antibacterial lipopeptides (2003), polyketides, non-ribosomal peptides, and aminoglycosides (Butler, 2005; Pham et al., 2019). *Staphylococcus aureus* infection is the leading cause of hospital-acquired pneumonia and infection of surgical wounds (Ambé et al., 2015). MRSA tends to be multi-drug resistant (MDR), exhibiting resistance not only to β -lactam antibiotics but also to a variety of antibiotic classes. Most of the active extracts contained tannins, (poly)phenols (including flavonoids, lignans and coumarins), terpenoids or alkaloids, which have been previously reported to be active compounds against MRSA. The considerable antibacterial activity against MRSA of grape seed extract (*Vitis vinifera* L; Vitaceae),

1127

which is rich in potent antioxidant polyphenolics, signified a major advancement in the treatment of MRSA diseases (Al-Mousawi et al., 2020). Aqueous extract of *Enantia polycarpa*, a tropical plant of the Annonaceae family (Ambé et al., 2015), had bactericidal activity against 75% of MRSA strains tested. It had minimum inhibitory concentration (MIC) and minimum bactericidal concentration (MBC) values of 3.125 and 12.5 mg mL⁻¹, respectively. Phytochemical analysis revealed the presence of saponins, alkaloids, and tannins in an aqueous extract (Ambé et al., 2015).

NPs against fungi

Many of the antibiotics currently used for the treatment of infections come from fungi. Fungi imperfecti, or *Penicillium*, and *Cephalosporium* were the two of the first fungi identified as having antibiotic activity (Dvorkin-Camiel and Whelan, 2008). There are five groups of antifungal agents approved for human consumption: the azoles, the polyenes, the echinocandins, the allylamines, and the antimetabolites (Howard et al., 2020).

Cryptococcal meningitis (CM) is a vital mycosis caused by Cryptococcus neoformans and Cryptococcus gattii, which leads to over one million cases and 600,000 deaths per year (Sloan and Parris, 2014; Liu et al., 2012; Park et al., 2009). Cryptococcosis primarily manifests after C. neoformans enters and colonizes the lungs (Saag et al., 2000; Samie et al., 2019). The successful colonization of the host by C. neoformans is attributed to its approximately 150 reported virulence factors (Malachowski et al., 2016). Pelargonium sidoides extract has been found to retain its antifungal activity for up to four years after preparation. The yeast cell wall and membrane are its first line of defense against its extracellular environment (Bahn et al., 2005). C. neoformans enlarges its capsule through distal growth by releasing and attaching capsular components stored within secreted vesicles to the outer edge of the capsule (Zaragoza et al., 2006). Preliminary investigation of the antifungal mechanism revealed that the active compounds induced apoptosis of C. neoformans cells and arrested the cell cycle at the G1/S phase. Berberine, a protoberberine-type isoquinoline alkaloid isolated from Berberis aquifolium, Berberis vulgaris, Berberis aristata, and Hydrastis canadensis, has been shown to have broad antibacterial and antifungal activity. Berberine causes damage to the cell membrane, and the possible impairment of cell function. Berberine may affect mitochondrial respiratory function, causing the breakdown of $\Delta \Psi m$ and a lack of accumulation of rhodamine 123 in the mitochondria (Ludovico et al., 2001).

NPs against parasites

Schistosomiasis, caused by flatworms of the genus Schisto-

soma, is one of the most neglected tropical diseases (Veras et al., 2012). Schistosoma mansoni is the major etiological agent of human schistosomiasis, and praziquantel (PZQ) is the only drug available to treat this neglected disease. There is therefore an urgent need for new drugs. Recent studies have indicated that extracts from *Piper aduncum* (Piperaceae) are active against adult worms of S. mansoni. Extracts of cardamonin from P. aduncum caused tegumental alterations, and reduction of oviposition and motor activity in worms of S. mansoni. Another two compounds, luteolin and (3R,6R)-linalool oxide acetate, showed anthelmintic activity against S. mansoni (IC₅₀ range from 5.8 to 36.9 μ g mL⁻¹). Luteolin induced tegumental damage in S. mansoni (Wangchuk et al., 2016). A bioguided phytochemical study identified 2-methoxy-6-pentyl-benzoquinone, also known as Primin, as the major bioactive metabolite producing toxicity against adult S. mansoni worms (Viegas et al., 2017). An in vivo study showed that the methanolic fraction of Clerodendrum umbellatum reduced the hepatosplenomegaly induced by infection with S. mansoni, and prevented the elevated malondialdehyde (MDA) level induced by the infection, while producing a significant increase in catalase activity and glutathione levels. Phytochemical analysis of an aqueous fraction from C. umbellatum leaf extract revealed the presence of alkaloids, flavonoids, cardiac glycosides, phenols, saponins, tannins, and terpenoids (Jatsa et al., 2015). Extract of Ramalina aspera shows a cercaricidal effect on S. mansoni at a concentration of 5.0 μ g mL⁻¹, with effective molluscicidal activity against the embryos and adult snails of the species B. glabrata and cercariae of S. mansoni. It could be a promising molluscicidal and cercaricidal agent (Silva et al., 2019).

NPs against other infectious pathogens

Prions are the infectious agents that cause neurodegenerative diseases, known as prion diseases or transmissible spongiform encephalopathies (TSEs). These diseases include Creutzfeldt-Jakob disease (CJD), kuru, fatal familial insomnia, scrapie, and bovine spongiform encephalopathy (Aguzzi and Liu, 2017). A variety of anti-prion compounds, including pentosane polysulfate, polyamines, amantadine, astemizole, dextran sulfate, Congo red, suramin, rapamycin, and quinacrine have been reported to bind to the prion protein PrP^C, acting as chemical chaperones to reduce the accumulation of PrP^{Sc} in cell culture (Ishibashi et al., 2016; Imberdis et al., 2016), with specific binding to the hotspot region of PrP^C (Ferreira et al., 2014). However, most of the compounds which were active in vitro have failed in vivo (Li and Weng, 2017). Two novel and potent NPs, BNP-03 ((1s,3R,4r,5S)-3,5-bis(((E)-3-(3,4-dihydroxyphenyl)acryloyl)oxy)-1,4-dihydroxycyclohexane-1-carboxylic acid) and BNP-08 ((1R,3R,4R,5S)-1,3-bis(((E)-3-(3,4-dihydroxvphenyl)acryloyl)oxy)-4,5-dihydroxycyclohexane-1-carboxylic acid) selected from self-constructed dabase by Choi et al. (2018), showed inhibitory effects, reducing PrP^{Sc} signals in a standard scrapie cell assay (Klöhn et al., 2003). Several naturally occurring polyphenols, phenothiazines, antihistamines, stating, and antimalarial compounds are potent prion inhibitors. Several of the new PrP^{Sc} inhibitors cross the blood-brain barrier, and thus have the potential to be active after TSE infection reaches the brain (Kocisko et al., 2003). In one study, a total of 500 marine invertebrate extracts were searched for yeast-prion inhibiting extracts. Five compounds exhibiting detectable anti-prion activities were isolated: aplysamine-1, aplysamine-2, purealidine-Q, 3,5dibromoverongiaquinol, and 3,5-dibromoverongiaquinol dimethyl ketal (Jennings et al., 2018). All active compounds identified in the extract of the sponge Suberea ianthelliformis contained an ethylaminodibromophenyl (EADP) moiety, and were identified as potent inhibitors of yeast prions (Hamann et al., 1993; Tsukamoto et al., 1996). The EADP structure class may serve as a useful lead for the future development and design of novel and improved anti-prion therapeutics (Tilvi et al., 2004; Ishibashi et al., 1991). The H₃ receptor, ApoE and AChE have all been identified as potential targets for treating various neurodegenerative diseases (Morisset et al., 2000; Torrent et al., 2015).

From the data collected, several different mechanisms appear to be involved in treating infectious diseases by natural products. Anti-microorganism mechanisms include enzyme inhibition and the disruption of the permeability or integrity of the membranes of microorganisms. The former acts through: (i) inhibition of integrase, reverse transcriptase, or RNA polymerase, such as treatments for HIV and IAV infection (Wang et al., 2018; Panthong et al., 2015; Huerta-Reves et al., 2004), which inhibit nucleic acid (DNA/RNA) replication of viruses in host cells; (ii) inhibition of essential enzymes for viral propagation like ATPase/MERS-3CLpro enzyme/SARS-3CLpro (Yu et al., 2012; Jo et al., 2019; Su et al., 2020); (iii) inhibition of the functional enzymes of microbes to reduce their pathogenicity, like GlmU/InhA/iNOS and COX-2 (Han et al., 2019; Hartkoorn et al., 2012; Alves-Silva et al., 2020). Additionally, NPs could affect the viability or activity of microorganisms (cell wall/cell membrane/mitochondrial membranes) (Silva et al., 2019 Zuzarte et al., 2011). Secondly, NPs act through the modulation of host cells by activating or inhibiting the inflammation-associated signaling pathways of host cells infected by microorganisms, such as NF-kB/JNK/MAPK/p53, to adjust the potential of enemy-destroying and self-destructive process to a better balance (Zhou et al., 2017; Han and Guo, 2012; Pang et al., 2011). Thirdly, NPs exert protective effects on cells through antioxidant mechanisms (Alam et al., 2012; Fankem et al., 2019; Reddy et al., 2010). Lastly, NPs can induce DNA damage (Singh et al., 2018), interfere with substance metabolism (Agarwal et al., 2008), exert immunomodulation (Basso et al., 2020) or autophagy modulation (Kaneda et al., 1991; Laconi et al., 2014).

Novel active skeleton compounds, such as Asperflavipine A, have become a topic of interest in NP research. Terpenoids, polyketones, alkaloids, and heterozygous KS-NRPS NPs have attracted considerable interest. Most of these NPs have anti-tumor, antibacterial, anti-inflammatory, and other biological activities. Novel NPs also have great potential because of their unique characteristics (Chen et al., 2020). The novel skeletons of heteroterpenes, such as Rhodomyrtusials derived from terpenoid biosynthesis, have shown great potential in studies of antibacterial activities. Due to their novel and diverse structural skeletons, they are widely distributed in microorganisms, including fungi and actinomycetes, sponges, Myrtaceae, Ericaceae, Lamiaceae, and some large fungi such as *Ganoderma lucidum* (Matsuda and Abe, 2016; Peng and Qiu, 2018; Zhao et al., 2020).

The hurdle remains in the researches of natural antimicrobial agent discoveries is that only the verification of their activity *in vitro* is effective, but insufficient research has been carried out *in vivo*, and the mechanisms of activity are difficult to explain, limiting the practical application of some effective NPs. The anti-microorganism mechanisms of NPs is the result of more than one mode of action, so the specific mechanism of an NP could be explored based on the integrated analysis of its effects on both the microorganism and the host cells, including studies of morphology, molecular mechanisms, and function. This is why the mechanisms of numerous conventional drugs have not been explored in depth. However, given the considerable potential of NPs for anti-microorganism activity, it is important to study their specific mechanisms of action (MOA).

Compounds from natural sources which inhibit the five groups of pathogens

In this section we provide a comprehensive analysis of the currently collected data using three approaches. To analyze the major classification of compounds selected and their chief MOAs, we used network pharmacology. To explore the structural relationships between NPs targeting different infectious pathogens, we calculated the maximum Tanimoto similarity scores between all molecule pairs of selected compounds. We also explored the diversity of their natural sources.

Identification of different classes of natural product-based compounds based on network pharmacology

From the collected data, we selected a total of 268 available NP-based compounds which have activity against 19

infectious pathogens, including HBV, IAV, HSV, HIV, SARS-CoV, SARS-CoV-2, MERS-CoV, Salmonella typhi, Shigella, Mycobacterium tuberculosis, MRSA, Helicobacter pylori, Cryptococcus neoformans, Malassezia furfur, Plasmodium, Schistosoma, Trichomonas vaginalis, Leptospira, and prions. Information about the MOAs of the compounds against microorganisms was obtained from the literature, and a network diagram was constructed using Cytoscape 3.8.0 (Shannon et al., 2003) (Figure 2). Target genes of compounds with a minimum required interaction score (indicating the prediction of compound targets) greater than 0.7 (high confidence) were obtained from the STITCH website (Szklarczyk et al., 2016). Functional enrichment analysis of target genes and the graphical display of the results of the analysis were conducted using the ClueGO v2.5.7 plug-in of Cytoscape 3.8.0 software (Bindea et al., 2009) (Figure 3).

The 268 NP-based compounds were classified into ten categories: sugars and glycosides, quinonoids, penylpropanoids, flavonoids, terpenoids, steroids, alkaloids, tannins, polyphenols, fatty acids, amino acids, and sulfur compounds. Based on our analysis, the compounds the most credible evidence of activity were agelasine D and dicumarol against *Mycobacterium tuberculosis*; dihydroartemisinin against *Plasmodium*; and pyridomycin against *Mycobacterium tuberculosis*. Other promising interactions, which would repay further investigation, are compounds 1,8-cineole, borneol, camphene, berberine, andα-pinene against *Cryptococcus neoformans*; artemisinin against *Plasmodium*; Baicalin, emodin, myricetin, and scutellarein against SARS-CoV; and methyl jasmonate against *Trichomonas vaginalis* (Figure 2).

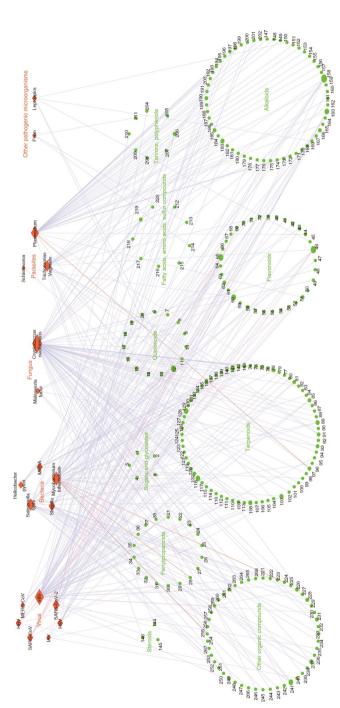
The possible MOAs of four compounds baicalin, emodin, berberine, and quercetin against the above infectious pathogens were illustrated in detail, as they have shown the most abundant inhibition effects. The most plausible pathways are those related to immunoregulation, anti-inflammatory activities, and modification of the replication of DNA and translation of the proteins of infectious pathogens. For instance, emodin negatively regulates DNA metabolism and silencing by miRNA (Figure 3). Emodin can protect infected cells by inhibiting the replication and maturation of the EV71 virus (Zhong et al., 2017). This observation also suggests that emodin may play a role against pathogens such as HIV and HSV by inhibiting their DNA replication.

These four compounds regulate similar signaling pathways, including (i) inflammation-immune-response-related signaling pathways: after pathogenic microorganisms infect body cells, they induce a series of inflammatory and immune responses to resist the invasion of pathogenic microorganisms (Häcker, 2018), which helps to regulate the response to infection. For example, emodin can regulate the production of chemokines and TNF- α ; berberine can regulate IL-1 β ; and quercetin can regulate the production of IL-8 and chemokines. (ii) Apoptosis-related signaling pathways, infection, and the subsequent excessive inflammation and immune response may induce apoptosis and cell damages (Zhang and Wang, 2014). These four compounds can regulate apoptosis-related signaling pathways, and may reverse cell damage through this common mechanism.

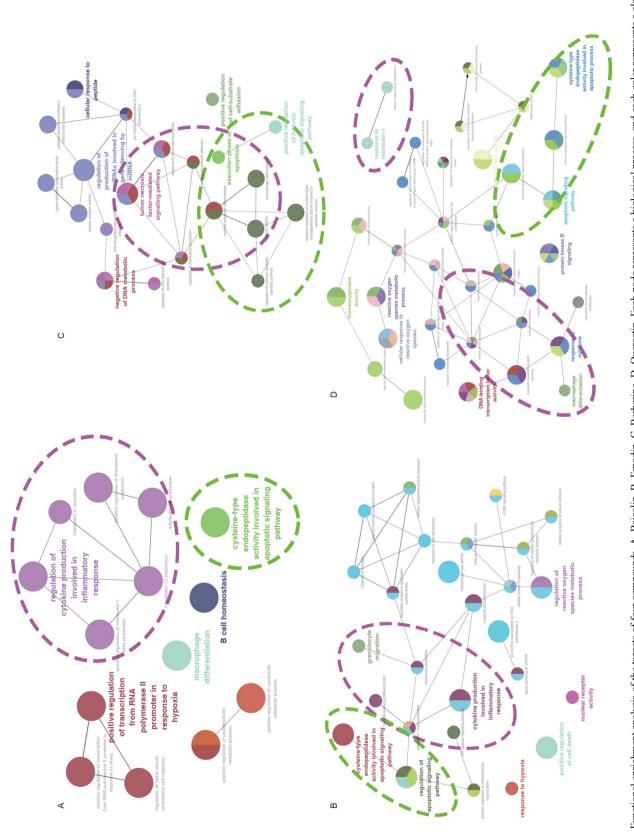
Chemical diversity of available natural product-based compounds

To explore the structural relationships of NPs targeting different infectious pathogens, we calculated the maximum Tanimoto similarity scores between all molecule pairs of the selected compounds. A total of 268 NP-based compounds targeting 19 pathogenic organisms from five categories were collected: virus (HBV, IAV, HSV, HIV, SARS-CoV, SARS-CoV-2, and MERS-CoV); bacteria (Salmonella typhi, Shigella, Mycobacterium tuberculosis, MRSA, and Helicobacter pylori); fungus (Cryptococcus neoformans, and Malassezia furfur); parasites (Plasmodium, Schistosoma, and Trichomonas vaginalis) and others (Leptospira and prions). The molfiles of the compounds in the Canonical SMILES format were downloaded from the ChEMBL online database (Gaulton et al., 2016). The Tanimoto similarity was calculated using a Perl program rewritten based on MolPrint 2D (Kong et al., 2011). To examine the relationship between the targeted pathogen types and the chemical diversity we subdivided the dataset into five subgroups (virus, bacterium, fungus, parasite, and others) (Pye et al., 2017).

Figure 4 shows that compounds, especially those inhibiting parasites, had high similarity, with a relatively large "hammerhead" distribution. Compared with the other subgroups, particularly the bacteria, fungi, and viruses, the distribution of the maximum Tanimoto scores is centered at relatively low values, suggesting that the chemistries of the compounds inhibiting these pathogenic organisms show high structural diversity. The targets and mechanisms of inhibition of parasites might be relatively less diversified comparing with those against bacteria and viruses, which contain a variety of targets and use different approaches. However, except for the "others" subgroup, the maximum Tanimoto scores of all infectious pathogens are considered to be high (T>0.4). These results indicate that, in the absence of significant innovation in discovery approaches, there are diminishing returns in terms of the discovery of fundamentally new chemical diversity from continued investigations into the same classes of natural sources. These conclusions agree with those of Pye et al. (2017). Therefore, more efforts should be paid to exploring novel targets, deciphering mechanisms, and discovering compounds. Recently, there have been some efforts to describe NP chemical space and to use this space as a boundary for designing NP-like synthetic screening libraries (Rosén et al., 2009; Pascolutti et al., 2015). For a given biosynthetic class of NPs, a huge number



Artemisinin, 84 Artesunate, 85 Betulinic acid, 86 Borneol, 87 Bornyl acetate, 88 Camphene, 89 Carvacrol, 90 Caryophyllene, 91 Caryophyllene oxide, 92 cis-fs-ocimene, 93 Citral, 94 Citronellol, 95 Cucurbitacin Figure 2 Compound-pathogen network of natural products based on a pharmacology network. Compounds are divided into ten groups and pathogens into five groups. Green circles and red rhombuses Cudraflavanone B, 50 Epicatechin, 51 Epigallocatechin gallate, 52 Epigallocatechin-(48-8,2B-O-7)-epicatechin, 53 Gallocatechin, 54 Helichrysetin, 56 Hesperidin, 57 Isobavaschalcone, 58 Kaempferol, 59 Lupinifolin, 60 Luteolin, 61 Mangiferin, 62 Myricetin, 63 Naringenin, 64 Prunin, 65 Quercetin, 66 Rutin, 67 Scutellarein, 68 Swertifrancheside, 69 Theaflavin, 70 Tiliroside, 71 Wogonin, 72 amangostin, 73 (+) Spathulenol, 74 (-)-bornyl acetate, 75 1,8-cineole, 76 2-carene, 77 3-O-acetyloleanolic acid, 78 3-carene, 79 3-epi-betulinic acid, 80 Alphitolic acid, 81 Andrographolide, 82 Arjunolic acid, 83 F, 96 Cuminaldehyde, 97 Dihydroartemisinin, 98 Epifriedelanol, 99 Eucalyptol, 100 Fenchone, 101 Friedelin, 102 Geraniol, 103 Glycyrrhizin, 104 Hederagenin, 105 Isodehydroluffariellolide, 106 Kaurenoic 119 Picroside, 120 Phytol, 121 Piperitone, 122 Pulegone, 123 Puupehenone, 124 Resinone, 125 Sabinene, 126 Salaspermic acid, 127 Sugiol, 128 Taraxerol, 129 Thymol, 130 Tripterifordin, 131 Ursolic acid, 132 Mallotochromene, 208 Mallotojaponin, 209 Proanthocyanidins, 210 Punicalagin, 211 Resveratrol, 212 Allicin, 213 Aplysamine-1, 214 Aplysamine-2, 215 Arginine, 216 Dihomo-y-linolenic acid, 217 Linoleic 11 Diospyrin, 12 Emodin, 13 Menadione, 14 Physcion, 15 Plumbagin, 16 Primin, 17 Rhein, 18 Shikonin, 19 Tanshinone Iia, 20 Thymoquinone, 21 (+)-calanolide A, 22 8-methoxypsoralen, 23 Arctigenin, 24 Calanolide A, 25 Calanolide B, 26 Chlorogenic acid, 27 Coumarins, 28 Dicumarol, 29 Gomisin J, 30 Helioxanthin, 31 Isoverbascoside, 32 Matairesinol, 34 Poliumoside, 35 Psoralen, 36 Soulattrolide, 37 Suksdorfin, 38 Verbascoside, 39 (-)-catechin, 41 5,7,2'-trihydroxyflavone, 42 Ampelopsin, 43 Apigenin, 44 Astilbin, 45 Astragalin, 47 Biochanin, 48 Catechin, 49 p-cymene, 133 a-bisabolol, 134 a-humulene, 135 a-pinene, 137 fb-bisabolene, 138 fb-caryophyllene, 139 fb-myrcene, 140 fb-phellandrene, 141 fb-pinene, 143 y-cymene, 144 fb-pinene, 144 fb-pinene, 145 y-cymene, 144 fb-pinene, 144 fb-pinene Dioscin, 145 Suberosol, 146 β-sitosterol, 147 (-)-O-O-dimethylgrisabine, 148 (-)-Roemrefidine, 149 (-)-tetrahydropalmatine, 150 Acetycorynoline, 151 Ancistrobarterine A, 152 Ancistrocongolines A, 153 Ancistrocongolines B, 154 Ancistrocongolines C, 155 Ancistrocongolines D, 156 Anonaine, 157 Asimilobine, 158 Berberine, 159 Boldine 3, 160 Buchenavianine, 161 Candimine, 162 Canthin-6-one, 163 phylline A, 174 Epiberberine, 175 Fangchinoline, 176 Fascaplysin, 177 Febrifugine, 178 Harman, 179 Isocorydine 1, 180 Jatrorrhizine, 181 Lanuginosine, 182 Laurolistine, 184 Lycorine, 185 Lysicamine, 186 Matrine, 187 Michellamine A, 188 Nitidine, 189 Oxoglaucine, 190 Oxymatrine, 191 Palmatine, 192 Plakinidine A, 193 Pseudopalmatine, 194 Quinine, 195 Roemerine, 196 Siamenol, 197 Stephanine, 198 Tetrahydroberberines, 199 Tetrahydropalmatine, 200 Triptonine B, 201 Voacamine, 202 Xylopinine, 203 p-coumaroyltyramine, 204 Chebulinic acid, 205 Corilagin, 206 Curcumin, 207 acid, 218 Oleic, 219 Palmitic acid, 220 Pentadecanoic acid, 221 2-amino-benzoic acid, 222 4-methoxycinnamic acid, 223 Acetovanillone, 224 Agelasine D, 225 Allantoin, 226 Ascorbic acid, 227 Atherosperminine, 228 Benzoic acid, 229 Benzyl benzoate, 230 Caffeic acid, 231 Camphor, 232 Cimamic acid, 233 Cordycepin, 234 Coumaric acid, 235 Cryptotanshinone, 236 Decane, 237 Ellagic acid, 238 Ethyl Nepetoidin B, 250 Nonanal, 251 Octyl gallate, 252 Paeonol, 253 Peroxyferolide, 254 Protocatechuic acid, 255 Pyridomycin, 256 Pyrocatechol, 257 Salicylic acid, 258 Sodium houttuyfonate, 259 Suramin, 260 correspond to compounds and infectious pathogens, respectively. The size of each node is proportional to its degree. The thickness and distance to the red line signifies the credibility of the supporting evidence. The numbers signify: 1 Astragaloside, 2 Forsythoside A, 3 Mannose, 4 Salzmannianoside A, 5 Salzmannianoside B, 6 Aloe-emodin, 7 Anthraquinone, 8 Chrysophanol, 9 Cudraxanthone S, 10 Dihydrotanshinone, acid, 107 Lancilactone C, 108 Limonene, 109 Linalyl acetate, 110 Lupenone, 111 Lupeol, 112 Lupulone, 113 Linalool, 114 Moronic acid, 115 Myrcene, 116 Madolin K, 117 Nigranoic acid, 118 Oleanolic acid, Cepharanthine, 164 Chelerythrine, 165 Columbarnine, 166 Coptisine, 167 Corydine, 168 Coumaroyttyramine, 169 Corynoline, 170 Decarine, 171 Dehydroroemerine, 172 Dioncophyllacine A, 173 Dioncoacetate, 239 Ethyl cinnamate, 240 Gallate, 241 Gallic acid, 243 Isocupressic acid, 244 Juliprosine, 245 Leonurine, 246 Methyl jasmonate, 247 N-cis-feruloyltyramine, 248 Neocnidilide, 249 Syringic acid, 261 Syringin, 262 Toxiusol, 263 Tryptanthrin, 264 Woodylide A, 265 Woodylide B, 266 n-butanol, 267 p-hydroxybenzyl alcohol, 268 *r*-fagarine.



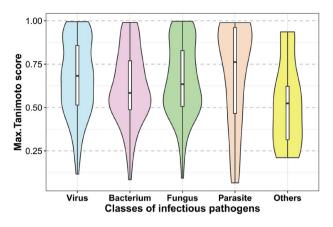


Figure 4 Violin plots of maximum Tanimoto scores of compounds among five categories of infectious pathogens.

of theoretical molecules can be created from primary building blocks such as amino acids, sugars, acetate, and propionate. Using these diverse and often chiral components, NP libraries should, therefore, exceed synthetic libraries in terms of the structural diversity of chemical scaffolds (Pye et al., 2017).

Classification of families of medicinal plants known to produce compounds inhibiting pathogenic microorganisms

Using the most reliable data, we collected information about compounds targeting four pathogenic organisms from four subgroups: C. neoformans (fungus), Salmonella typhi (bacterium), HIV (virus), and Plasmodium (parasite), to explore the diversity of their natural sources. A subset of 331 species of medicinal plants known to produce compounds with biological activities against the above four pathogenic microorganisms was collected from the literature. Data about 129 species producing NPs were collected for C. neoformans, 62 species for HIV, 26 for Plasmodium, and 114 for Salmonella typhi. In order to visualize the lineages selected above, four species trees with the different subgroup of compounds based on the chloroplast genome (Figure 5) were constructed in R 3.4.2 (The R Core Team, 2013) with the "phytools" package, using the function S.PhyloMaker (Oian and Jin, 2016). The backbone of the tree and nodes was obtained together with the S.PhyloMaker script (Qian and Jin, 2016). The tree was visualized and annotated in iTOL (Letunic and Bork, 2016).

NP-based compounds were relatively more abundant in the Asteraceae, Celastraceae, Lamiaceae, and Myrtaceae families for *C. neoformans*; the Anacardiaceae, Asteraceae, Euphorbiaceae, Lamiaceae, Leguminosae, and Polygonaceae families for *Salmonella typhi*; the Cucurbitaceae, Euphorbiaceae, Lamiaceae, Rutaceae, and Schisandraceae families for HIV; and the Ancistrocladaceae, Annonaceae,

Apocynaceae, Lauraceae, and Zingiberaceae families for Plasmodium (Figure 5). We concluded that, though the discovery of unexplored and unusual sources provides opportunities for finding novel NPs, NP-based compounds from the same family, genera or species may share similar secondary metabolic pathways, mechanisms, and targets, in which specific synthetic pathways could be further explored.

Based on the above analysis, sugars and glycosides, quinonoids, penylpropanoids, flavonoids, terpenoids, steroids, alkaloids, tannins, polyphenols, fatty acids, amino acids, and sulfur compounds from NP sources are the ten most promising compounds against infectious pathogens (Table 1). Inflammation and immune response-related signaling pathways and apoptosis-related signaling pathways are the two major pathways through which these compounds act against pathogenic microorganisms. The NP library is in need of further exploration, especially for the identification of novel targets and the deciphering of mechanisms, as the structural diversity of chemical scaffolds is of great abundance. The families Asteraceae, Celastraceae, Anacardiaceae, Cucurbitaceae. Ancistrocladaceae and associated genera and species are found to contain abundant compounds which act against specific pathogenic microorganisms. Synthetic versions of these pathways should be further explored, to discover effective compounds. The targets and mechanisms for inhibiting bacteria and viruses are diverse. More efforts should be put into exploring novel targets, deciphering mechanisms of action, and discovering compounds in NP-like synthetic screening libraries.

In silico methods for the identification of natural product-based compounds

NP drug discovery is a multidimensional problem, requiring the consideration of several factors for both natural and synthetic compounds, including safety, pharmacokinetics, and efficacy, to be evaluated during the selection of drug candidates (Chi et al., 2006). NP discovery is achieved using two main approaches. As the conventional "top-down" approach has already been largely abandoned (Jensen et al., 2014), the advent of latest technologies such as artificial intelligence (AI), "organ-on-chip", and microfluidics technologies (Chi et al., 2006) has bolstered the emergent "bottom-up" approach, offering the capacity to access the unexpressed genetic potential of microorganisms (Bachmann et al., 2014). Recent advances in analytical and computational techniques have made possible new approaches to the analysis of complex NPs, to design synthetic versions of compounds and to derive new and innovative drugs (Chi et al., 2006). Metabolic pathway engineering and synthetic biology have transformed NP discovery, production, and engineering (Pye et al., 2017).

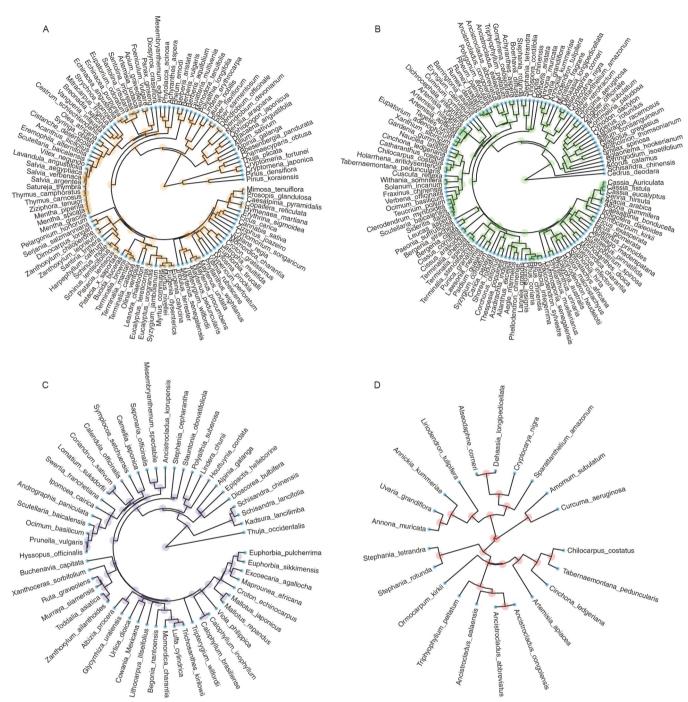


Figure 5 Four species trees of medicinal plants for the known compounds in (A) Crytococcus neoformans, (B) Salmonella typhi, (C) HIV, and (D) Plasmodium, based on the chloroplast genome.

The area of NP discovery *in silico* is becoming increasingly active, mainly due to the reduced time, risks, and resources taken by this approach compared to traditional experimental approaches (Luo et al., 2020) (Figure 6). Computational models can be used to identify inadequate knowledge about the targets of NP, undesirable pharmacokinetic expressions upon target interaction, or off-target effects. Therefore, it is necessary to implement *in silico* methods and powerful data resources to facilitate the design and redesign of NP-like molecules with desired bioactivity, and to predict and validate NP targets. There is an urgent need to find therapeutic NPs when infectious diseases are prevalent (Trosset and Cavé, 2019). NPs have a large quantity of lead-like molecules, which could be used as scaffolds to expand the chemical library (Gu et al., 2013). Chemical similarity search depends heavily on the existence of reference compounds for known targets, and the availability of data about existing NPs or preclinical NP

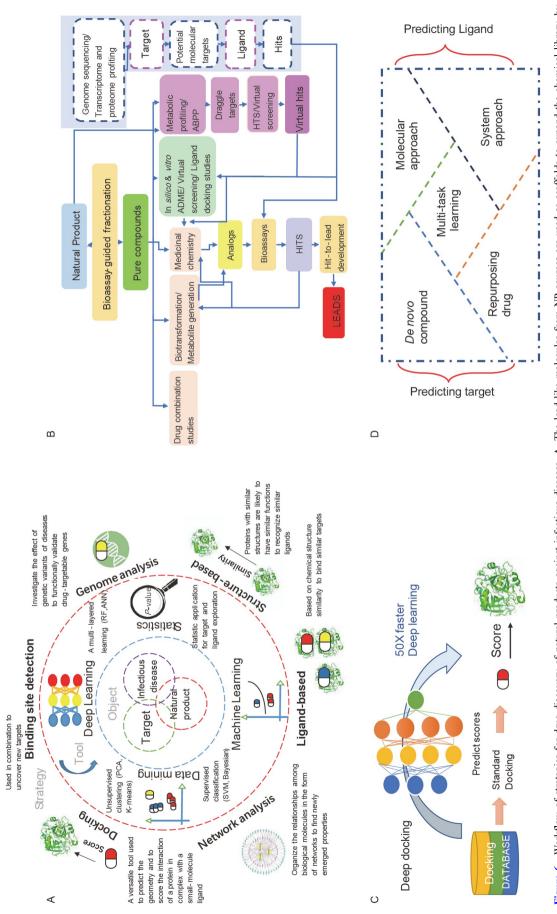
Table 1	Species of natural	products with most abun	dant compounds against infectious	s pathogens, taken from the literature

No.	Species of natural products	Representative compounds	Infectious pathogens
1	Sideritis italica	β-Pinene, β-Cubebene, β-Bisabolene, α-Bisabol, tricyclene, p-Metoxyacetophenone, palmitic acid, m-Metoxyacetophenone, limonene, carvacrol, camphor, benzyl benzoate, Δ3-Carene, nuciferol, nonanal, and kaur-15-ene	Salmonella typhi
2	Barringtonia asiatica	Uncineol, tetradecanol, humulene oxide, hexyl hexanoate, geranyl butyrate, ethyl valerate, ethyl lactate, eicosane, decane, acetovanillone, 4-Propyl-guaiacol, and (Z)-4- decenal, (-)-c-elemene	Salmonella typhi
3	Juglans regia	Quercetin 3-xyloside, quercetin 3-rhamnoside, quercetin 3-galactoside, quercetin 3- arabinoside, p-coumaric acid, palmitic acids, linoleic acid, 4-p-coumaroylquinic acids, 3-caffeoylquinic acids, 5-caffeoylquinic acids, 3-p-coumaroylquinic acids, and oleic	MRSA, Cryptococcus neoformans
4	Cannabis sativa	β-sitosterol, pentadecanoic acid, p-coumaroyltyramine, methyl hexadecanoate, friedelan-3-one, ergost-5-en-3-ol, epifriedelanol, 4-hydroxy-3-methoxybenzaldehyde, 10E-hexadecenoic acid, naringenin, and β-sitosterol-β-D-glucoside	MRSA, Cryptococcus neoformans, Helicobacter pylori
5	Ancistrocladus abbreviatus	Ancistrobrevine J, ancistrobrevine I, ancistrobrevine H, ancistrobrevine G, ancistrobrevine F, ancistrobrevine E, 6-O-Demethylancistrobrevine H, 5-epi-Ancistrobrevine F, 5-epi-Ancistrobrevine E, and 5'-O-Demethylancistrobrevine B	Plasmodium
6	Piper flaviflorum	Sarmentine, piperyline, piperolactam D, pellitorine, homopellitorine, demethoxypiplartine, brachyamide B, 1-[(2E,4E,9E)-10-(3,4-methylenedioxyphenyl)-2,4,9-unde- catrienoyl]pyrrolidine, (2E,6E)-sarmentosine, and (2E)-decenoylpiperidide	Cryptococcus neoformans
7	Ecdysanthera rosea	Pregnane glycoside ecdysantheroside A, C-21 pregnane glycoside ecdysosides H, C-21 pregnane glycoside ecdysosides G, C-21 pregnane glycoside ecdysosides F, C-21 pregnane glycoside ecdysosides E, C-21 pregnane glycoside ecdysosides D, C-21 pregnane glycoside ecdysosides C, C-21 pregnane glycoside ecdysosides B, and C-21 pregnane glycoside ecdysosides A	Cryptococcus neoformans
8	Guatteria multivenia	Dihydromadolin-K, guadiscine, guatterin A, lanuginosine, liriodenine, lysicamine, madolin-K, madolin-W, and o-methylpallidine	MRSA, Cryptococcus neoformans
9	Pinus koraiensis	Limonene, isolongifolene, caryophyllene, camphene, β-pinene, β-myrcene, alpha- pinene, and 3-carene, (-)-bornyl acetate	Cryptococcus neoformans
10	Stephania rotunda	Xylopinine, vireakine, tetrahydropalmatine, stephanine, roemerine, pseudopalmatine, palmatine, dehydroroemerine, and cepharanthine	Plasmodium
11	Liriodendron tulipifera	Peroxyferolide, oxoglaucine, norushinsunine, norglaucine, liriodenine, lipiferolide, asimilobine, and anonaine	Plasmodium
12	Tripterygium wilfordii	Triptonine B, tripteryols C, tripteryols B, tripteryols A, tripterifordin, salaspermic acid, (±)-5,4'-dihydroxy-2'-methoxy-6', 6"-dimethypyraro-(2",3":7,8)-6-methyflava- none, (2S)-5,7,4'-trihydroxy-2'-methoxy-8, and 5'-di(3-methyl-2-butenyl)-6-methyl- flavanone	HIV, Cryptococcus neoformans

candidates (Mujawar et al., 2018). We collected databases from the literature (Table 2). Virtual NP databases can be categorized into (i) encyclopedic and general NP databases, (ii) databases enriched with NPs used in traditional medicines, and (iii) specialized databases focused on specific habitats, geographical regions, organisms, biological activities, or even specific NP classes. However, the availability of materials for experimental evaluation produces a bottleneck in NP-based drug discovery, as many of those resources do not offer free downloads (Kinghorn et al., 2019).

In silico methods have strongly impacted the way in which new targets are identified for old ligands (Keiser et al., 2009; Cameron et al., 2013), the prediction of side effects (Lounkine et al., 2012), and the development of anatomical therapeutic indicators of approved ligands (Wu et al., 2013). Over the past few decades, several strategies have become important, including network-based analysis; molecular docking; ligand-based and structure-based approaches, including chemical structures and reactions; protein structure; protein-protein interactions; signal transduction; genetic interactions; and metabolic networks (Agamah et al., 2020). These strategies have been widely used to predict candidate targets and ligand-target interactions. Molecular docking is used to predict the geometry and score the interactions of proteins in complex with small-molecule ligands (Kitchen et al., 2004). These methods can be used to predict whether a given ligand is potentially able to bind other targets. Ligand-based analysis indicates that similar compounds tend to have similar biological properties, and have been extensively used to analyze and predict the activity of ligands for new targets (March-Vila et al., 2017). Structure-based analysis is based upon the idea that proteins with similar structures are likely to have the same functions, and to recognize related ligands. In the field of drug repurposing, protein comparison is used as a method to identify secondary targets of an approved ligand (Ehrt et al., 2016) (Figure 6A and B).

Machine learning (ML) or data mining (DM) approaches have permeated *in silico* methods, and play critical roles in uncovering significant patterns in chemical and pharmacological property space, which are essential for compound discovery. Advanced machine learning models and algorithms such as support vector machines (Burbidge et al.,



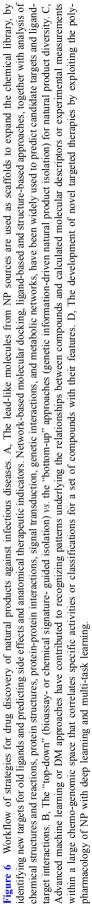


Table 2 Databases available for <i>in silico</i> analysis of natural products

Database	Description	URL	Reference
Binding database (BindingDB)	Protein binding database	https://www.bindingdb.org	Kirchmair et al., 2015
ChEMBL	Drug bioactivity data	https://www.ebi.ac.uk/ chembl/	Bento et al., 2014
Chinese Natural Product Database (CNPD)	Chinese Natural Product Database	https://www.neotrident. com/	Shen et al., 2003
Comparative Toxicogenomics Database (CTD)	Gene-Drug-Disease interactions	http://ctdbase.org/	Davis et al., 2017
The Drug Gene Interaction Database (DGIdb)	Gene-Drug interactions	http://dgidb.genome.wustl. edu	Wagner et al., 2016
Drug Repurposing	Drug repurposing	https://clue.io/repurposing	Corsello et al., 2017
DrugBank	Gene-Drug interactions and drug information	https://www.drugbank.ca/	Law et al., 2014
Hetionet	Combination of 29 public databases on genes, diseases, drugs, and side effects	https://het.io/	Himmelstein et al., 2017
Illuminating the Druggable Genome	Drug-targeted protein families	https://druggablegenome. net/	Rodgers et al., 2018
Kyoto Encyclopedia of Genes and Gen- omes (KEGG)	Drug Information on drugs and targets	https://www.genome.jp/ kegg/drug/	Kanehisa and Goto, 2000
Library of Integrated Network-Based Cel- lular Signatures (LINCS)	Gene expression and drugs	http://www.lincsproject. org/	Keenan et al., 2018
NCGC Pharmaceutical Collection (NPC)	A comprehensive, publically-accessible collection of approved and investigational drugs for high-through- put screening that provides a valuable resource for both validating new models of disease and better understanding the molecular basis of disease pathol- ogy and intervention	https://tripod.nih.gov/npc/	Huang et al., 2011
Orphanet	Rare diseases and orphan drugs	http://www.orpha.net	Pavan et al., 2017
Pharos	Knowledgebase for the druggable genome	https://pharos.nih.gov/idg/ index	Nguyen et al., 2017
PubChem	Chemical database	https://pubchem.ncbi.nlm. nih.gov/	Kim et al., 2016
repoDB	Clinical trial and repositioning database	http://apps.chiragjpgroup. org/repoDB/	Brown and Patel, 2017
Side Effect Resource	information on marketed medicines and their recorded adverse drug reactions	http://sideeffects.embl.de/	Kuhn et al., 2016
STITCH	Chemical-protein interaction networks	http://stitch.embl.de/	Szklarczyk et al., 2016
SuperTarget	Drug targets and side effects	http://insilico.charite.de/ supertarget/	Hecker et al., 2012
The Toxin and Toxin Target Database (T3DB)	Gene-toxin database	http://www.t3db.ca/	Wishart et al., 2015
TCM Database	Drug screenings of Traditional Chinese medicine	http://tcm.cmu.edu.tw/	Chen, 2011
Traditional Chinese Medicine database (TCMDb)	Chinese herbal medicines, components, targets, and diseases	http://166.111.57.233/	He et al., 2001
Traditional Chinese Medicine integrated da	http://www.megabionet. org/tcmid/	Xue et al., 2013	
Traditional Chinese Medicine Systems Pha	armacology (TCMSP)	http://www.tcmspw.com/ tcmsp.php	Ru et al., 2014
Transformer	Cytochrome-drug interactions	http://bioinformatics.char- ite.de/transformer/	Hoffmann et al., 2014
Therapeutic Target Database	Drug targets	http://bidd.nus.edu.sg/ group/cjttd/	Chen et al., 2002
UniProt	Protein database	www.uniprot.org/	Consortium, 2015
Universal Natural Product Database (UNPD)	a comprehensive resource of natural products for virtual screening	http://pkuxxj.pku.edu.cn/ UNPD	Gu et al., 2013
YaTCM	Linking traditional Chinese medicine to targets and diseases	http://cadd.pharmacy.nan- kai.Edu.cn/yatcm/home	Li et al., 2018
Super Natural	bio/cheminformatics and personalized medicine	http://bioinformatics.char- ite.de/supernatural/	Hoffmann et al., 2014

(To be continued on the next page)

Database	Description	URL	Reference
Super Natural II	2D structures; vendor information for over 215,000 compounds	http://bioinf-applied.char- ite.de/supernatural_new/in- dex.php	Tsai et al., 2011
Universal Natural Product Database	3D structures assembled from other available Chinese databases	http://pkuxxj.pku.edu.cn/ UNPD/introduction.php	Gu et al., 2013
Drug Discovery Portal	A collection of biological target and computational tests for potential drug candidates using the physics of molecular interactions	https://drugdiscovery.tacc. utexas.edu/#/	Clark et al., 2010
Medicinal plant Activities, Phytochemicals & structural database	Medicinal plant activities, phytochemical and structural database	http://www.mapsdatabase.	Schuster et al., 2010
AfroDb	Compounds from African medicinal plants	https://omictools.com/ afrodb-tool	Ntie-Kang et al., 2013
Universal Natural Products Database (UNPD)	A largest non-commercial and freely available database for natural products	http://pkuxxj.pku.edu.cn/ UNPD	Gu et al., 2013
Dictionary of natural products	A structured database holding information on chemical substances	http://dnp.chemnetbase. com	Petersen et al., 2011
MarinLit	A database dedicated to marine natural products research	http://pubs.rsc.org/marin- lit/	Valli et al., 2013
Global Natural Products Social Molecular Networking (GNPS)	A web-based mass spectrometry ecosystem sharing raw, processed or identified tandem mass (MS/MS) spectrometry data	https://gnps.ucsd.edu	Aron et al., 2020
KNApSAcK	information on the relationships between metabolites and their expressing organism(s)	http://kanaya.naist.jp/ KNApSAcK_Family/	Nakamura et al., 2013
Natural Product Activityand Species Source Database (NPASS)	Integrating detailed information of species sources and biological activities of natural products	http://bidd2.nus.edu.sg/ NPASS/	Zeng et al., 2018
Ambinter and Greenpharma	Natural products	http://www.ambinter.com; http://www.greenpharma. com	-
AnalytiCon Discovery	MEGx—Purified natural products of microbial and plant origin	http://www.ac-discovery. com	_
Chengdu Biopurify Phytochemicals	TCM compounds library	http://www.biopurify.com	-
Selleck Chemicals	Natural products	http://www.selleckchem. com	_
TargetMol	Natural compound library	http://www.targetmol.com	-
MedChem Express	Natural product library	http://www.medchemex- press.com	_
InterBioScreen	Natural compound collection	http://www.ibscreen.com	-
TimTec	Natural product library (NPL)	http://www.timtec.net	-
AK Scientific	Natural products	http://www.aksci.com	-
Developmental Therapeutic Program (DTP)of NCI NIH	Natural products Set IV	http://dtp.cancer.gov/orga- nization/dscb/obtaining/ available_plates.htm	_
INDOFINE Chemical	Natural products, flavonoids, coumarins, etc.	http://www.indofinechem- ical.com	INDOFINE Chemical Com- pany, 2019
Pharmeks	Screening compounds	http://www.pharmeks.com	-
Princeton BioMolecular Research	Macrocycles	http://www.princetonbio.	_
MicroSource Discovery Systems	Natural products collection (NatProd)	http://www.msdiscovery. com	_
Specs	Natural products	http://www.specs.net	-

2001), neural networks (Lo et al., 2018), logistic regression (Wu and Zhao, 2019), naive Bayesian classification (Flach and Lachiche, 2004), binary kernel discrimination (Wu and Zhao, 2019), partial least squares (Mazandu et al., 2011) and random forests (Lo et al., 2018), have contributed to recognizing patterns underlying the relationship between

compounds and calculated molecular descriptors or experimental measurements within a large chemo-genomic space (Weaver, 2004). In this space specific activities or classifications for a set of compounds are correlated with their features, thus enabling clustering of similarities among NPlike compounds in multidimensional space (Weaver, 2004). For instance, Fang et al. (2017) developed approaches to the development of novel targeted therapies by exploiting the poly-pharmacology of NP. In the new chemical entities (NCE), the initial resource value of substructure associations, drug-target interactions, and drug-substructure associations can be defined as $A_{CS}(i, j)$, $A_{DS}(i, j)$ and $A_{DT}(i, j)$, respectively. The initial resource matrix can be built via these three values for the substructure-drug (or NCE)-target network, and transformed into a final resource matrix by appropriate parameterization. Deep learning and multi-task learning emerge as important in silico methods (Figure 6C and D). For example, Francesco introduced Deep Docking (DD), a novel deep learning platform suitable for docking billions of molecular structures in a rapid vet accurate fashion (Gentile et al., 2020). Using a biomedical knowledge graph, a new potential medication, baricitinib, was identified by BenevolentAI for inhibiting SARS-CoV-2. It has shown promising effects on controlling infection and reducing inflammatory injury, and clinical trials have now started.

Future directions and perspectives

Challenges facing natural-product-based drug discovery for infectious diseases

There are many difficulties, related to the availability of efficient and rapid bioassay systems, in screening and identifying conventional NP-based drugs with auspicious antimicrobial activities (Akram et al., 2018). The production of NPs requires reliable supplies of the source material, which may vary by seasonal or through environmental change, or may be lost through extinction or legislation (Tanrikulu et al., 2009). The transition of a natural compound from a "screening hit" through a "drug lead" to a "marketed drug" is associated with demands for increasing amounts of the compound (Atanasov et al., 2015). The traditional approach for NP discovery once relied heavily upon luck (Jensen et al., 2014). The complex chemical structures of bioactive NPs hamper the development of methods for total synthesis or derivation that might be needed for the property optimization of lead candidates (Atanasov et al., 2015). Another major challenge for NP drug discovery is the incompatibility of NPs with HTS (Koehn and Carter, 2005). Adaption and changes in sample preparation and assay designs are necessary to use HTS for the detection of bioactivity in plant extracts and to identify potent pure compounds (Torres et al., 2017). A further challenge is to clarify the precise molecular mechanisms and signaling pathways involved in the bioactivity (Corson and Crews, 2007). Due to high costs, clinical trials of NPs are rarely supported by industry, while the pharmaceutical companies also worry about the patentability of NPs (Atanasov et al., 2015). These issues have all limited the development of NP medicines.

Antibiotic abuse, especially in developing countries, has resulted in the emergence of resistant microbes due to the evolutionary selection pressure imposed by antibiotics (Gwynn et al., 2010). Around 90%–95% of *Staphylococcus aureus* strains worldwide are resistant to penicillin, and 70%–80% of the same strains are methicillin-resistant in many Asian countries (Hemaiswarya et al., 2008). The number of effective therapeutic measures against life-threatening infectious pathogens has fallen dramatically because of the emerging MDR pathogens (Jensen et al., 2014). Therefore, it is extremely urgent for the discovery of novel antibiotics to keep pace with the growing threat of drug resistance (Silber et al., 2016).

The lack of effective therapies and vaccines for various viral infections, and the rapid emergence of new drug-resistant viruses, have produced a growing need for the development of effective new natural-product-based agents to treat viral diseases. New viruses have emerged throughout human history, causing tens of millions of deaths. At present, climate change and globalization have created favorable conditions for the spread of viruses, and in the future, new virus outbreaks may be even more frequent. Therefore, the development of effective antiviral drugs, especially broadspectrum antiviral drugs for conservative targets of different viruses, to fight potential outbreaks of new and re-emerging viruses in the future is a topic of general concern in the academic and industrial communities.

In viruses, the sequences and structures of proteins performing the same functions are often highly similar, such as the spike proteins of SARS-CoV and SARS-CoV-2. These proteins can be used as common targets in the development of antiviral drugs. Among them, representative drugs are Rukobia (fostemsavir, bms-663068), which went on the market in 2020, and the SARS-CoV-2 antibody developed by Eli Lilly and Regeneron Pharmaceuticals, which has been approved for emergency use in the COVID-19 outbreak. However, virus SFTSV is usually under high mutation pressure, so it is difficult to develop broad-spectrum antiviral inhibitors against this target.

At present, only nucleoside analogs targeting viral DNA/ RNA polymerase, such as ribavirin and favipiravir, and compounds targeting host proteins, can be used as broadspectrum antiviral drugs against a large class of viruses. Viruses from the same genus usually contain several highly conserved epitopes, such as TmS and viral protease, which can be further developed as common targets for broadspectrum antiviral drugs. For example, EK1, a virus inhibitor developed for coronavirus before the outbreak of COVID-19, also showed antiviral activity against SARS-CoV-2, indicating the feasibility of designing a broad-spectrum inhibitor for the viruses of the same genus in advance. Therefore, cocktail therapy composed of broad-spectrum viral drugs targeting DNA/RNA polymerase combined with two to three kinds of drugs targeting SP transmembrane subunits or viral proteases can be used as a reserve scheme for the treatment of new viruses from the same genus.

Meanwhile, unlike bacteria, viruses have no characteristics in common, such as cell walls. For example, coronavirus and orthomyxovirus have few common components. Therefore, common targets can only be found for certain families of viruses. The natural immune system of the human body plays an important role in antiviral immunity. Different viral infections lead to changes in the expression of various genes in the host, leading to the host presenting different physiological states. After the virus invades the body, it can induce type I and type III interferons (IFNs) to stimulate the body cells to produce hundreds of interferon stimulated genes (ISGs) to combat the virus. The proteins expressed by ISGs can not only directly inhibit the replication of the virus, but also regulate the expression of IFNs, which play indirect antiviral roles. A large number of studies have reported the antiviral effects and related mechanisms of a variety of ISGS proteins and ISGS family proteins. Different ISGs can inhibit the virus at different stages of infection, from invasion to release. Therefore, an herbal medicine targeting human ISG genes can play a broad-spectrum antiviral role by regulating the release of specific ISG genes.

Solutions and development in natural-product-based drug discovery for infectious disease

New NP screening technologies are currently being exploited to improve hit rates for antibacterial discovery. Lead optimization could be enhanced by the identification of new antibiotic classes with improved tractability, and by expanding the predictability of *in vitro* safety assays. Implementing multiple screening and target identification strategies is recommended for improving the likelihood of discovering new antibacterial compounds that address unmet needs (Jensen et al., 2014). The intrinsic complexity of NPbased drug discovery necessitates highly integrated interdisciplinary approaches (Atanasov et al., 2015).

NP isolation is laborious and time-consuming, and careful selections of environmental sources of microbes and experimental methods are required to improve the outcomes of isolation procedures (Jensen et al., 2014). The recent rapid increase in genetic information technology has led to novel screenings, and genetic techniques have facilitated the implementation of combinatorial biosynthetic technology and genome mining. The knowledge gained has allowed large numbers of previously unknown molecules to be identified (Akram et al., 2018). Omics analysis, including genomics, transcriptomics, proteomics, metabolomics, and metabolomics, results in the generation of complex multivariate datasets that require computational and chemometric tools for interpretation (Tanrikulu et al., 2009). Genomics has been

applied in the identification of NPs and biomarkers, proteomics has been applied to NP validation and biomarker identification, metabolomics and metabonomics have been used in identification studies, and big data has been applied to NP-based drug development and precision medicine as well as target identification and reverse pharmacology (Prachayasittikul et al., 2015).

Continued development of technologies such as genome sequencing, bioinformatics, microbial genomics, metabolic pathway engineering, and synthetic biology are poised to usher in a new age of NP discovery and drug development (Jensen et al., 2014). This is a highly informed approach, in which strains can be prioritized based on a bioinformatic assessment of their biosynthetic potential (Teijaro et al., 2018). The advent of high-throughput genome sequencing techniques has made accessible increasing numbers of microbial genomes which can be evaluated *in silico* for their capacity to produce secondary metabolites (Hoffmann et al., 2018). Advancements in microbial genomics have shown that, for instance, in Actinobacteria, approximately 90% of biosynthetic capacity has not been realized (Teijaro et al., 2018; Harvey et al., 2015).

The NPs from various sources, including terrestrial and marine microorganisms, fungi, invertebrates, and plants, each possess unique structural features. These features result from the production of NPs via enzymatic reactions catalyzed by proteins programmed by DNA sequences that differ greatly between the source organisms (Naman et al., 2017). Taxonomic distribution analysis of candidate signals can be especially useful to support the discovery of new NPs (Hoffmann et al., 2018). A study conducted by Pye et al. (2017) demonstrated that most NPs share structural similarity to previously published compounds. Combining the chemotaxonomic method with molecular phylogenetic data would help to select natural species from genera or families known to produce compounds or compound classes associated with certain bioactivities or therapeutic effects in a targeted manner (Ramos Barbosa et al., 2012). Phylogenetic and phytochemical studies have shown that there is a strong phylogenetic signal in the distribution of secondary metabolites in NPs (Saslis-Lagoudakis et al., 2011; Saslis-Lagoudakis et al., 2012). In particular, the exploration of crosscultural ethnomedical patterns within a phylogenetic framework is regarded as a very powerful tool for the identification of highly promising plant groups, when phylogenetically related plant species from very distant regions are found to be used for medical conditions with the same therapeutic aims (Saslis-Lagoudakis et al., 2011; Saslis-Lagoudakis et al., 2012; Atanasov et al., 2015).

Another strategy employed to overcome mechanisms of resistance is the use of a combination of drugs producing synergistic activity against microorganisms (Hemaiswarya et al., 2008). Antibiotics generally target the microbial cell wall, but using the approaches described in this review, new therapeutic targets are being identified. The auto-inducer molecules produced by quorum sensing systems are currently targeted by novel compounds to control antibiotic resistance and biofilm formation, quorum sensing, and quorum quenching, and are clearly important as a potential target for natural substances (Mulat et al., 2019). The future development of new antimicrobial agents will rest with those who employ synthetic and semi-synthetic methodology and a further understanding of microbial cell architecture and drug resistance mechanisms (Thomford et al., 2018).

Although this review presents a thorough evaluation of publicly available data, there may be some NP-derived compounds that have been overlooked. Emerging technologies, such as genomics and synthetic biology, are enabling new ways of discovering and utilizing NP-based products. We are entering an exciting era in which the ancient wisdom distilled into the world's traditional herbal medicines can be reinterpreted and exploited through the lens of modern science, demystifying conventional medicines with modern approaches (Li et al., 2017).

Relevant information about 268 NP-based compounds classified into ten groups against 19 selected infectious pathogens was collected for network analysis. The distribution of maximum Tanimoto scores indicated that compounds which inhibit parasites exhibited a low diversity, whereas the chemistries inhibiting bacteria, fungi, and viruses showed more structural diversity. There is a lack of innovation in the methods for discovery of NPs, and more efforts should be paid to exploring novel targets, deciphering mechanisms of action, and discovering bioactive compounds. A total of 331 species of medicinal plants with compounds exhibiting pharmacological activities against four specific infectious pathogens were selected to classify the family sources. NP libraries offer a diversity of chemical components for the discovery of drugs against infectious pathogens. This review summarizes the currently available data on NP-based antimicrobials against refractory infections, to provide information for further discovery and synthesis of drugs acting against infectious pathogens.

Compliance and ethics *The author(s) declare that they have no conflict of interest.*

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SUPPORTING INFORMATION

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