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Original Article

Artificial intelligence approach fighting COVID-19 with repurposing drugs

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Abstract

Background: The ongoing COVID-19 pandemic has caused more than 193,825 deaths during the past few months. A quick-to-be-identified cure for the disease will be a therapeutic medicine that has prior use experiences in patients in order to resolve the current pandemic situation before it could become worsening. Artificial intelligence (AI) technology is hereby applied to identify the marketed drugs with potential for treating COVID-19.

Material and methods: An AI platform was established to identify potential old drugs with anti-coronavirus activities by using two different learning databases; one consisted of the compounds reported or proven active against SARS-CoV, SARS-CoV-2, human immunodeficiency virus, influenza virus, and the other one containing the known 3C-like protease inhibitors. All AI predicted drugs were then tested for activities against a feline coronavirus in vitro cell-based assay. These assay results were feedbacks to the AI system for relearning and thus to generate a modified AI model to search for old drugs again.

Results: After a few runs of AI learning and prediction processes, the AI system identified 80 marketed drugs with potential. Among them, 8 drugs (bedaquiline, brequinar, celecoxib, clofazimine, convivaptan, gencatibase, tolcapone, and vismodegib) showed in vitro activities against the proliferation of a feline infectious peritonitis (FIP) virus in Fcwf-4 cells. In addition, 5 other drugs (boceprevir, chloroquine, homoharringtonine, tilorone, and salinomycin) were also found active during the exercises of AI approaches.

Conclusion: Having taken advantages of AI, we identified old drugs with activities against FIP coronavirus. Further studies are underway to demonstrate their activities against SARS-CoV-2 in vitro and in vivo at clinically achievable concentrations and doses. With prior

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The outbreak of a new coronavirus SARS-CoV-2, previously known as the 2019-nCoV, has caused 2.81 million confirmed cases of the infection. There are more than 193,825 patients died of the coronavirus disease 2019 (COVID-19) affecting 210 countries and territories and the death toll is still climbing up sharp. SARS-CoV-2 was first reported found in Wuhan, China and, soon after, a human-to-human transmission is confirmed followed by a global COVID-19 pandemic as WHO declared on the 11th of March. Majority of the patients of COVID-19 present with fever and respiratory symptoms and death in some serious complications [1]. There are a number of clinical studies ongoing for treating COVID-19 with potential old drugs and investigational new drugs. In order to quickly resolve the current pandemic with more medicinal cures, repurpose of old drug is one of the strategies highly appreciated.

We have hereby practiced with an artificial intelligence (AI) system for searching marketed drugs potentially with antiviral activities against coronaviruses. An AI system could be an efficient tool to quickly screen large number of compounds with assigned learning datasets and an intention to find drugs for specific purposes such as for treating COVID-19. The established AI system analyzed the most important descriptors, generated relevant prediction models, performed screening for the marketed drugs to quickly identify those with potential activities inhibiting SARS-CoV-2. An in vitro cell model for feline coronavirus replication was set up to evaluate the AI-identified drugs for antiviral activity verification.

Feline coronavirus is a z-coronavirus and the virus causing enteritis in domestic and wild cats. Approximately 5–15% of infected cats develop feline infectious peritonitis (FIP) which is fatal to cats [2]. The infection by FIP virus in cats presented similar features to the severe acute respiratory syndrome (SARS) infection such as pulmonary lesions in humans [3]. It was documented that both the nucleoside analog GS-441524 and 3C-like protease inhibitor GC376 exhibited antiviral activities against FIP virus in vitro and were effective for treating FIP in cats [4,5]. Remdesivir (GS-5734) is the prodrug of GS-441524 that inhibits viral RNA-dependent RNA polymerase and prevents viral replication in vitro including the middle east respiratory syndrome (MERS) virus, Ebola virus, Lassa fever virus, Junin virus and respiratory syncytial virus [6]. Remdesivir is now an investigational new drug that shows promising outcomes in compassionate uses.

| Table 1 Learning datasets for AI training. |
|------------------------------------------|
| **Dataset for AI Model 1**              |
| 1. SARS-CoV active drugs - Promazine, Niclosamide. |
| 2. Influenza drugs - Favipiravir, Oseltamivir, Peramivir, Zanamivir. |
| 3. HIV drugs - Atazanavir, Darunavir, Elvitegravir, Fosamprenavir, Indinavir, Lopinavir, Remdesivir, Ritonavir, Saquinavir, Tipranavir. |
| 4. Others reported - Abacavir, Cinanserin, Deoxyrhapontin, Ebselen, Enzaplavitovir, Maribavir, Shikonin, Montelukast, Polydatin, Presatovir, Paltegravir, Shikonin, Sophoradin, Tideglusib. |

| **Dataset for AI Model 2**              |
| 1. Blanchard et al. (5 compounds) [15]. |
| 2. Chen et al. (12 compounds) [16]. |
| 3. Chen et al. (8 compounds) [17]. |
| 4. Chen et al. (8 compounds) [18]. |
| 5. Ghosh et al. (10 compounds) [19]. |
| 6. Ghosh et al. (7 compounds) [20]. |
| 7. Jacobs et al. (19 compounds) [21]. |
| 8. Jain et al. (8 compounds) [22]. |
| 9. Kao et al. (1 compounds) [23]. |
| 10. Kuo et al. (9 compounds) [24]. |
| 11. Lu et al. (21 compounds) [25]. |
| 12. Mukherjee et al. (2 compounds) [26]. |
| 13. Thanigaimalai et al. (5 compounds) [27]. |
| 14. Thanigaimalai et al. (7 compounds) [28]. |
| 15. Tsai et al. (27 compounds) [29]. |
| 16. Turlington et al. (25 compounds) [30]. |
| 17. Wen et al. (4 compounds) [31]. |
| 18. Wu et al. (11 compounds) [32]. |
| 19. Yang et al. (9 compounds) [33]. |
| 20. Zhang et al. (4 compounds) [34]. |
| 21. Zhang et al. (8 compounds) [35]. |

**Dataset for Modified AI Model**

1. Results from AI model 1.
2. Results from AI model 2.
3. Other drugs active against FIP virus shown in Table 4.

We use experiences in patients, these old drugs if proven active against SARS-CoV-2 can readily be applied for fighting COVID-19 pandemic.

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present study is therefore relevant and shall provide a useful screening tool to identify marketed drugs with a broad-spectrum antiviral activity.

**Materials and methods**

Two different learning datasets were generated in which one is consisted of the compounds reported or proven active against SARS-CoV, SARS-CoV-2, human immunodeficiency virus (HIV), and influenza virus and the other contains the known 3C-like protease inhibitors. An AI-system was established and, based on the learning datasets to predict drugs potentially active against coronavirus out of the marketed drugs. The predicted drugs were then tested for activities against a feline coronavirus in a cell-based FIP virus replication assay for a verification. During the AI practices, these assay results served as feedbacks to the AI system for relearning and thus to generate a modified AI model to search for old drugs.

**Generation of datasets and AI models**

Three types of molecular descriptors, extended connectivity fingerprint (ECFP), functional-class fingerprints (FCFPs), and octanol–water partition coefficient (ALogP_count), were performed to AI learning. The extended connectivity fingerprints (ECFPs) are generated in a molecule-directed manner by systematically recording the neighborhood of each non-hydrogen atom into multiple circular layers up to a given diameter of that molecule. The functional-class fingerprints (FCFPs) detail circular fingerprints via the pharmacophore identification of atoms, which reports topological pharmacophore fingerprints. The ALogP_count is an array of 120 numbers that correspond to the 120 Ghose and Crippen atom types. The 613 descriptors in total were used for the AI practices and prediction of the promising drugs. All these descriptors were generated by Discovery Studio v18.1/Calculates ligand properties program (BIOVIA Inc., San Diego, CA, USA), including ALogP_count (101 descriptors), ECFP_4 (256 descriptors), and...
FCFP_4 (256 descriptors) as shown in the supplemental information. The system used a Deep Neural Network (DNN) [9] algorithm to identify the most important descriptors and gave different weightings to generate AI prediction models. We collected information of those compounds and drugs that had been reported with promising activities for treating COVID-19, such as anti-influenza drugs [10], drugs and compounds inhibiting SARS-CoV [11], HIV drugs [12], and some showing activities against SARS-CoV-2 as recently reported [13,14]. The other AI prediction model is generated from a learning dataset of 210 drugs and compounds reported with inhibitory activities on the 3C-like protease of SARS-CoV listed in Table 1 [15–35]. The database adopted to search for marketed drugs was the DrugBank at website: https://www.drugbank.ca.

Virus and cells

The serotype II FIP virus of Taiwan isolate NTU156 strain, was a kind gift from National Taiwan University [36]. The virus was propagated and titrated in the feline catus whole fetus-4 (Fcwf-4) cells (ATCC®CRL-2787). Fcwf-4 cells were amplified and maintained in Dulbecco’s modified Eagle’s medium (DMEM) containing 10% fetal bovine serum (FBS) with 1% penicillin/streptomycin at 37°C and 5% CO2 for 1 h and the supernatant were discarded. Each concentration of testing compounds (0.4, 2, 10, and 50 μM) in duplicate wells (Row a and b) and cytotoxicity of brequinar at 0.4, 2, 10, 50 μM was also investigated (Row c). FIPV-infected cells treated with a positive control, anti-FIP virus compound GC-376 (Row d) and vehicle only (Row e). Uninfected cells without drugs in parallel treatments were shown (Row f).

Measurement of antiviral effects in FIP virus-infected Fcwf-4 cells

In brief, 3 × 10⁴ Fcwf-4 cells per well were seeded in 96-well plates containing DMEM supplemented with 10% FBS and incubated at 37°C with 5% CO₂. Sixteen hr post inoculation, the cells were infected with NTU156 at 300 TCID₅₀ per well and incubated at 37°C for 1 h and the supernatant was discarded followed by addition of 0.4, 2, 10, and 50 μM of testing compounds in DMEM medium containing 2% FBS. The culture plates were incubated for additional 48 h at 37°C with 5% CO₂ followed by fixing with 10% formalin and stained with 0.1% crystal violet 20% ethanol (w/v). The cytotoxic effect (CPE) caused by FIP virus were measured visually for intensity of crystal violet staining. Back titration of FIP virus were performed parallely and the TCID₅₀ values were calculated according a Reed and Muench method [37].

Results

Results from AI model 1 approach

AI prediction model 1 was exercised based on the learning dataset listed in Table 1, containing some of the drugs for influenza, from previous reports showing activities against SARS-CoV, HIV drugs, and drugs and compounds of potential inhibiting SARS-CoV-2 as recently reported as shown in Fig. 1. The AI model identified 22 drugs to be having the potential of antiviral effects. Followed by evaluations on these 22 drugs with the in vitro cell-based FIP virus replication assay, 2 promising drugs clofazimine and gemcitabine were verified with activity against the FIP coronavirus (Table 2) (see Fig. 2). Q2

Results from AI model 2 approach

An independent approach from the AI model 1, AI prediction model 2 was practiced based on the learning dataset of 210 inhibitors of the 3C-like protease of SARS-CoV listed in Table 1. The AI model predicted 38 drugs with potential inhibiting the 3C-like protease of SARS-CoV and they were subject to tests in the FIP virus replication assay. Two out of the 38
AI-identified drugs, celecoxib and tolcapone, showed antiviral activities against the FIP coronavirus (Table 3).

**Results from the modified AI model**

The results from the two AI models 1 and 2 were shown in Tables 2 and 3, respectively, and were served together as a feedback dataset to the AI system for relearning and therefore a modified AI prediction system was established and used for the follow-up prediction exercise. During our integrated efforts for repurposing drugs to tackle COVID-19, there were additional drugs found active in our in vitro assay against the FIP virus (Table 4), such as boceprevir [38], chloroquine [39], homoharringtonine [40], tilorone [41], and salinomycin [42] as they were also reported previously with potential to inhibit SARS-CoV or SARS-CoV-2. These 5 drugs were included as a sub-dataset for the relearning and thus to generate the modified AI prediction system which was used to screen and identify those 20 drugs out of the approved 2684 drugs for tests in the FIP virus replication assay (Table 5). With the AI evolution, the system showed a higher fidelity and 4 drugs (bedaquiline, brequinar, convivapam, and vismodegib) out of the 20 AI-identified were verified with antiviral activities against the FIP coronavirus.

**Discussion**

Predicted by the AI system, 8 drugs were identified with antiviral activities against FIP virus proliferation in the Fcwf-4 cells. Together with the 5 additional drugs identified during the current AI approach and drug repurposing practices, there are 13 drugs with great potentials for further development toward treating COVID-19. Chloroquine showed inhibition activity against FIP virus at 10 μM, which is in agreement with the recent report about its broad-spectrum antiviral effects and it is one of the drugs used to treat COVID-19 patients in China and authorized for emergency use in United States [14]. Remdesivir is a

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**Table 2 Assay results of the drugs predicted by AI Model 1.**

| AI predicted drugs | Concentration (μM) | Cytotoxicity | Viral Inhibition |
|--------------------|-------------------|--------------|-----------------|
| 5-Azacytidine      | >10               | NA°          |                 |
| Agomelatine        | >50               | >50          |                 |
| Carprofen          | >50               | >50          |                 |
| Clofazimine        | >50               | >10          |                 |
| Cobistat           | >50               | >50          |                 |
| Cyclofenil         | >10               | NA           |                 |
| Cytarabine         | >2                | NA           |                 |
| Dexamethasone      | >50               | >50          |                 |
| Dolasetron         | >50               | >50          |                 |
| Flurbiprofen       | >50               | >50          |                 |
| Gemcitabine        | >50               | ≥2           |                 |
| Histamine          | >50               | >50          |                 |
| Ipriflavone        | >50               | >50          |                 |
| Loratadine         | >10               | NA           |                 |
| Melatonin          | >50               | >50          |                 |
| Pexidartinib       | >2                | NA           |                 |
| Raxolotib          | >50               | >50          |                 |
| Sennoside A        | >50               | >50          |                 |
| Torsemide          | >50               | >50          |                 |
| Triclabendazole    | >2                | NA           |                 |
| Tropisetron        | >50               | >50          |                 |
| GC376b             | >50               | 0.4–2        |                 |

° NA, not available to detect due to cytotoxicity.

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**Table 3 Assay results of the drugs predicted by AI Model 2.**

| AI predict drugs | Concentration (μM) | Cytotoxicity | Viral Inhibition |
|------------------|-------------------|--------------|-----------------|
| ABT-199          | >2                | NA           |                 |
| Baricitinib       | >50               | >50          |                 |
| Bifonazole        | >10               | NA           |                 |
| Celecoxib         | ≥50               | 50           |                 |
| Chlorothiazide    | >50               | >50          |                 |
| Clofazimine       | >50               | >50          |                 |
| Dapsonate         | >50               | >50          |                 |
| Dihydroergotamine | >50               | >50          |                 |
| Dolasetron        | >50               | >50          |                 |
| Eletriptan HBr    | >50               | >50          |                 |
| Eptifibatide      | >50               | >50          |                 |
| Etodolac          | >50               | >50          |                 |
| Fenbufen          | >50               | >50          |                 |
| Flufenamic acid   | >50               | >50          |                 |
| Fomepizole        | >50               | >50          |                 |
| Hesperidin        | >50               | >50          |                 |
| Hydrochlorothiazide| >50             | >50          |                 |
| Indapamide        | >50               | >50          |                 |
| Naratriptan       | >50               | >50          |                 |
| Nimesulide        | >50               | >50          |                 |
| Nitrazoxanide     | >10               | NA           |                 |
| Nitroprusside     | >50               | >50          |                 |
| Quinapril         | >50               | >50          |                 |
| Rizatriptan       | >50               | >50          |                 |
| Ruxolotib         | >10               | NA           |                 |
| Serotonin         | >50               | >50          |                 |
| Sorafenib         | >50               | >50          |                 |
| Sumatriptan       | >50               | >50          |                 |
| Tadalafil         | >50               | >50          |                 |
| Thalidomide       | >50               | >50          |                 |
| Tofacitinib       | >50               | >50          |                 |
| Tolcapone         | >50               | 50           |                 |
| Trichloromethazine| >50               | >50          |                 |
| Triclabendazole   | >50               | >50          |                 |
| Trifluoracetil    | >50               | >50          |                 |
| Tropisetron       | >50               | >50          |                 |
| Vemurafenib       | >10               | NA           |                 |
| Zinc Pyrithione   | >0.4              | NA           |                 |

° NA, not available to detect due to cytotoxicity.

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**Table 4 Other drugs showing antiviral activities against FIP virus.**

| Drugs            | Concentration (μM) | Cytotoxicity | Viral Inhibition |
|------------------|-------------------|--------------|-----------------|
| Boceprevir       | >50               | 50           |                 |
| Chloroquine      | >10               | 10           |                 |
| Homoharringtonine| >2                | 2            |                 |
| Salinomycin      | >10               | 10           |                 |
| Tilorone          | >10               | 10           |                 |

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AI identified drugs show activities against the proliferation of SARS-CoV-2 RNA-dependent RNA polymerase inhibitor and a promising drug candidate for treating COVID-19 [4]; now used in COVID-19 patients in clinical trials in a number of countries including Taiwan. GS-441524 is the active metabolite of remdesivir and showed activity against FIP virus at around 2 μM similar to the previously reported for GS-441524 being developed as a veterinary medicine treating feline coronavirus infections [4]. The results therefore provide solid evidence to support the validity of using cell-based FIP virus infection in vitro model as an antiviral activity assay in the present study. To have the promising drugs further being tested in a SARS-CoV-2 in vitro assay and in vivo animal models is planned to be ongoing.

Among these AI-identified active drugs, four drugs (brequinar, clofazimine, gemcitabine and conivaptan) are of potential in their clinical uses and hereby elaborated further. Brequinar is a synthetic quinolinecarboxylic acid analogue and a dihydroorotate dehydrogenase (DHODH) inhibitor with experiences of patients uses for antitumor activities [43]. It was also used as an immunosuppressant in combination with other drugs such as cyclosporine for synergism [44]. It is of high water solubility and given by IV infusion. The biodistribution studies in rats of brequinar showed that liver and small intestines were the most exposed organs. The elimination rate of brequinar from all tissues were approximately equal to that from the blood circulation indicating no tissue accumulation of the intact drug molecules [45]. The present finding demonstrated that brequinar has an inhibiting activity of 2–10 μM. The systemically exposed plasma concentration range of brequinar in patients are around 73–261 μM for doses of 135–300 mg/kg in 5-day infusion period suggesting that brequinar may be effective against coronaviruses in appropriate dose regimens given alone or in combination with others [44].

Clofazimine is a water-soluble dye developed for the treatment of tuberculosis and used in combination for leprosy. It has a long circulation half-life of 25 days and the serum concentrations remain relatively low with evident tissue accumulation to extremely high levels approximately as 1.5 mg/g in jejunum. After 20 weeks treatment, the peak concentrations in lungs were 106 μg/g and 9020 μg/g in the lungs and spleen, respectively, in mice [46]. Clofazimine exhibited a dose-related relationship of tissue and serum concentrations (1.56–2.23 μM) from an oral dose of 100–300 mg in patients. A long onset of at least two weeks was observed in patients. Taking the advantages of the very slow clearance from body long half-life, and high drug concentration in the lungs [47], clofazimine with minimum adverse effects could be further investigated for treating COVID-19.

Conivaptan is a dual V1A/V2 vasopressin antagonist and given intravenously for the treatment of euvolemic and hypervolemic hyponatremia. The peak concentration C0.5hr of the loading dose at 40 mg/day was about 1.4 μM and the maintenance dose at 40 mg/day for 4-day infusion was around 0.4 μM in patients [48]. Although these concentrations were lower than the inhibitory concentrations at approximately 50 μM, a clinical regimen may be manageable to seek possible efficacy and tolerability in COVID-19 patients with proper study design.

Gemcitabine is a antimetabolite and first-line anticancer drug for pancreatic cancer. It also exhibits activities against breast cancer, ovarian cancer, non-small cell lung cancer, and bladder cancer. Gemcitabine is currently used alone or combined in cancer patients reached 600–1200 mg/m² gemcitabine of 60–120 min i.v. infusion patients in treatment cycles with achievable plasma concentrations of 25–139 μM [49]. Side effects of gemcitabine include pneumonitis, pulmonary fibrosis, pulmonary edema, and adult respiratory distress syndrome. These pulmonary events could lead to fatal respiratory failure in some cases despite discontinuation of gemcitabine therapy 2 weeks ahead. A treatment regimen of low dose and frequency may be explored for single use against COVID-19 in animal models and for combined uses with the other drugs [50]. The present study revealed that the inhibitory activity of gemcitabine on coronavirus FIPV was approximately at 2 μM. A proper combined use with the other drugs in a lower dose of gemcitabine may overcome the pulmonary adverse effects while active inhibiting SARS-CoV-2 for cure of COVID-19 in further preclinical and clinical studies.

### Table 5 Assay results of drugs predicted by the modified AI model.

| AI predicted drugs | Concentration (μM) | Cytotoxicity | Viral Inhibition |
|--------------------|-------------------|-------------|-----------------|
| Bedaquiline        | >50               | 50          | NA              |
| Bifonazole         | >10               | NA*         | 2               |
| Brequinar          | >50               | 2           | NA              |
| Clofazimine        | >2                | NA          | 50              |
| Conivaptan         | >50               | 2           | NA              |
| Duvelisib          | >50               | >50         | NA              |
| Eaconazole         | >10               | NA          | >50             |
| Fenticonazole      | >2                | NA          | >50             |
| Fumaric acid       | >50               | NA          | >50             |
| Lapatinib          | >10               | NA          | >50             |
| Miconazole         | >2                | NA          | >50             |
| Miconazole (nitrate)| >10               | NA          | >50             |
| Fumonisin          | >50               | NA          | >50             |
| Pranlukast         | >50               | NA          | >50             |
| Sertaconazole      | >10               | NA          | >50             |
| Sucranolose        | >10               | NA          | >50             |
| Sucranolose (nitrate)| >10             | NA          | >50             |
| Tacrolimus         | >50               | >50         | NA              |
| Telmisartan        | >50               | >50         | NA              |
| Tipifarnib         | >10               | NA          | >50             |
| Vismodegib         | >50               | 50          | NA              |

* NA, not available to detect due to cytotoxicity.

### Conclusion

AI approach is an efficient way to help quickly identify potential drugs as antiviral therapeutics. Some of the AI-identified drugs show activities against the proliferation of feline coronavirus in vitro. At the time of writing, there are still thousands of people dying of COVID-19 pandemic every day and we hope any of these AI-identified old drugs could soon make it to an effective cure for COVID-19.

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Declaration of Competing Interest
The authors declare that they have no competing interests.

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Appendix A. Supplementary data
Supplementary data to this article can be found online at https://doi.org/10.1016/j.bj.2020.05.001.

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