

A comprehensive drug repurposing study for Covid19 treatment: Novel putative DHODH inhibitors show association to Serotonin-Dopamine receptors

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Abstract

DHODH is a key enzyme required for de novo pyrimidine synthesis and it is suggested as a target for Covid19 treatment due to high pyrimidine demand by the virus replication in the infected host cells as well as its proven effect of blocking of cytokine release by the immune cells to prevent inflammation leading to acute respiratory distress. There are a number of clinical trials underway for Covid19 treatment using DHODH inhibitors, however, there are only a small number of known DHODH antagonists available for testing.

Here, we have applied a methodology to identify DHODH antagonist candidates, and compared using in silico target prediction tools. A large set of 7900 FDA-approved and clinical stage drugs obtained from DrugBank were docked against 20 different structures DHODH available in PDB. Drugs were eliminated according to their predicted affinities by Autodock Vina. 28 FDA-approved and 79 clinical trial ongoing drugs remained. The mode of interaction of these molecules were analyzed by repeating docking using Autodock 4 and DS Visualiser. Finally, the target region predictions of 28 FDA approved drugs were determined through PASS and SwissTargetPrediction tools.

Interestingly, the analysis of in silico target predictions revealed that serotonin-dopamine inhibitors could also be potential DHODH antagonists. Our candidates shared a common attribute, a possible interaction with serotonin-dopamine receptors as well as other oxidoreductases, like DHODH. Moreover, the BTK-inhibitor Acalabrutunib and serotonin-dopamine inhibitor drugs on our list have been found in the literature that have shown to be effective against Sars-Cov-2, while the path of activity is yet to be identified.

Identifying an effective drug that can suppress both inflammation and virus proliferation will play a crucial role in the treatment of Covid19 disease. Therefore, we suggest experimental investigation the 28 FDA-approved drugs on DHODH activity and Sars-Cov-2 virus proliferation. Those who are found experimentally effective can play an important role in Covid19 treatment. Moreover, we suggest investigating Covid19 case conditions in patients using schizophrenia and depression drugs.

Keywords: Covid-19, Sars-Cov-2, DHODH, Target Prediction, Molecular Docking

Introduction

In December 2019, it was announced that a fatal pneumonia disease occurred in Wuhan, China. Clinicians in China have found that the cause of this disease was a new coronavirus. This virus, called Sars-Cov-2, has been determined as a result of the genome sequencing analysis and was 79.5% similar genetically to multiple acute respiratory syndrome (SARS) virus^{1,2}. This virus, which spread from Wuhan to the whole world, has infected more than 6 million as of May 31, 2020, causing more than 375 thousand deaths. The Sars-Cov-2 virus, which the World Health Organization has declared as a pandemic as of February 11, 2020, causes acute lung injury, as the main cause of death.

More than 80 clinical trials are underway, such as drug development, vaccination studies, and serum/plasma treatment³. Considering patient serum amount is low as a recently spreading disease and the vaccine development is expected to take time, it is of great importance to test FDA-approved drugs in the treatment of Covid-19 disease, especially for the rapid control of the disease⁴. Current drug discovery studies focus on RNA polymerase inhibition⁵, ACE-2 and Spike Protein Blockers⁵⁻⁷, TMPRSS2 protease inhibitors⁸. These target regions are focused for preventing the virus from binding to the target protein, entry into the cell and replication. However, all these targets under study, to our knowledge, are not particularly focused for their effects against acute lung injury, which is the main cause of Covid-19 disease.

As mentioned above, 79.5% of the genomic sequences of Sars and Sars-Cov-2 viruses are highly similar regions, and their biochemical interactions and pathogenesis have been shown to utilize similar pathways⁹. The binding to the ACE-2 receptor of Sars-Cov-2 lung type-2 endothelial cells, inflammation triggers the cascade, results in respiratory failure due to acute lung respiratory damage (ARDS)^{9,10}. One of the most important factors of ARDS formation is due to the uncontrolled release of pro-inflammation cytokines (such as IFN- α , IFN- γ , IL-1B, IL-6, IL-12, IL-18, TNF- α). Following virus entry and infection of cells, host immune response and inflammation cascade begin via antigen-presenting cells (APC) and macrophages^{4,9}. This process takes place due to two functions of APC: (1) It provides antigen against foreign pathogen to CD4 + T cells (Th1) and (2) releases Interleukin-12 to stimulate Th1 cells. Stimulated Th1 cells stimulate CD8 + T-killer cells (Th2) and attack all cells with a foreign antigen against this pathogen¹¹. It also triggers B cells to produce antigen-specific antibodies

when Th1 cells are activated. Interferon-1 (IFN-1) has a protective effect especially in SARS and MERS infection. It has been detected in animal studies that IFN-1 signal transduction is suppressed in cells with infection of the SARS virus¹². Also, antigen creation is suppressed by the virus. It has been determined that non-structural (nsp) proteins located in the ORF regions of the virus genome play a role in these suppression mechanisms¹³.

Most common drug treatment methods aim treatment of the newly infected or about preventing the infection at the first place⁹. However, since patients admit to hospitals with symptoms of cough and fever, at this stage the inflammation has already started. The use of remdesivir and hydroxychloroquine at the time of the outbreak has been common for the treatment of Covid19. However, with increasing number of clinical studies, it has been revealed that these two drugs do not show statistically significant effect against this disease¹⁴.

Favalli et al. (2020) stated that similar pathophysiological findings were found between Covid-19 disease and Rheumatoid Arthritis diseases¹⁵. They suggested that anti-rheumatic drugs can also be used in Covid-19 disease¹⁵. Although rheumatoid arthritis (RA) is painful and quite debilitating, it is a chronic inflammatory disease characterized by gradual destruction of joints, deformity, disability and premature death in infants¹⁶.

Anti-inflammatory drugs, in particular, anti-cytokines has been used to prevent inflammatory formation¹⁷. One of these treatment methods has been the suppression of the DHODH (dihydroorotate dehydrogenase) enzyme¹⁸. Pyrimidines play roles in the formation of phosphodiester bonds with purines in double helix DNA, glycoprotein, phospholipids, RNA and DNA^{19,20}. There are 2 important ways in the synthesis of pyrimidines in most conserved organisms; salvage and *de novo* pathways¹⁹. Under normal conditions, along with differentiated cells, inactive lymphocytes prefer salvage pyrimidine synthesis^{21,24}. However, upon stimulation, they need *de novo* pyrimidine synthesis where the DHODH enzyme takes part. With the suppression of pyrimidine synthesis, activated lymphocytes undergo metabolic stress and the release of pro-inflammatory cytokines such as IL-17 and IFN- γ decreases leading to apoptosis²²⁻²⁴.

Dihydroorotate dehydrogenase (DHODH) is a flavoenzyme localized in the inner membrane of the mitochondria, involved in the 4th step of *de novo* pyrimidine synthesis^{21,25}. Using Ubiquione as a cofactor, it catalyses dihydroorotate to orotate. It is necessary for the cell as it is the only enzyme that can perform this process during the formation of uridine monophosphate (UMP) used in RNA synthesis²⁰.

DHODH enzyme is proven to play an active role in cancer and immunological disorders such as acute myeloid leukemia, rheumatoid arthritis, multiple sclerosis^{18,21,26-28}. In addition, DHODH inhibition has been found to have anti-viral effects against rotavirus, dengue virus, foot and mouth disease virus.²⁹⁻³³ Liu et al. (2020) showed that suppression of *de novo* pyrimidine synthesis in anti-viral treatment strategies, apart from RNA-dependent RNA polymerase inhibition, may be the target mechanism for many viral pathogens including coronavirus³⁴.

When the pathogenesis of Covid-19 disease on lung injury was examined, similar immune reactions were detected in autoimmune diseases such as rheumatoid arthritis¹⁵. Suppression of the DHODH enzyme in humans is a proven treatment method for immune disorders, especially cancer, rheumatoid arthritis (Leflunomide-Arava Company) and multiple sclerosis (Teriflunomide-Aubagio Company).

Beside the anti-inflammatory effect of the DHODH inhibition, the viruses are in need a high amount of nucleic acid to complete their life cycle in the host cell. Therefore, inhibition of nucleotide biosynthesis was previously considered a potential anti-viral strategy⁶. In accordance, DHODH inhibitors used in the treatment of autoimmune diseases have also been identified by studies that have broad spectrum anti-viral effects³¹.

The most detailed research for suppression of Sars-cov-2 virus replication through DHODH inhibition has been published as preprints by Xiong R. *et al.*, (2020)³⁵. According to the study announced in Wuhan, inhibition of the DHODH enzyme reported that the Sars-Cov-2 virus reduced the proliferation in the cell, and that it was an important strategy that could be used in Covid-19 disease due to the immune response suppression effect of DHODH inhibitors. The researchers identified 2 candidate molecules with a high binding score that could target the active site of the DHODH enzyme from 280,000 molecules via docking to protein data bank structure of DHODH, 6M2B. Both candidate molecules suppressed Sars-Cov-2 proliferation 17 times stronger ($IC_{50}=17 \mu M$) compared to FDA approved DHODH inhibitors ($IC_{50}=300 \mu M$). In addition, in the cell culture study, DHODH⁺ and DHODH-A549 cells, which was created with the CRISPR technique, were infected with the Sars-Cov-2 virus and the proliferation amount of the viruses was investigated. It was reported that there was no significant change in DHODH-A549 cell proliferation compared to DHODH⁺ cells after 72 hours. In DHODH-A549 cells, the virus was found to grow 1000 times slower. With the CRISPR study, it was determined that DHODH enzyme is important for the virus replication.

In vitro and clinical studies utilizing DHODH inhibitors has been increasing against Covid19 (Chictr2000030058, ClinicalTrials Code: NCT04361214, ClinicalTrials Code: NCT04379271). Yet, considering the fast pace of the pandemic, and low rate of drugs passing the clinical stage, more candidates are necessary for testing. The aim of our study is to determine potential candidates among approximately 7900 candidate compounds, including FDA approved drugs, to target the DHODH enzyme using molecular docking analysis on 20 different DHODH crystal structure. Those with the high binding energy in all DHODH enzymes tested were distinguished. Among the FDA-approved drugs further *in silico* target-prediction analysis was performed. The target similarities of the molecules determined by comparing FDA-approved DHODH inhibitor, leflunomide, are discussed. It is important to determine the effects of these drugs with high binding score on the DHODH enzyme obtained in the analysis on the Sars-Cov-2 virus *in vitro* and *in vivo* in the global effort to find an effective treatment against Covid-19.

Methods

Obtaining ligands

All drug structures of 7900 ligands with a 3D structure are downloaded from DrugBank³⁶ and converted from sdf format into individual PDBQT files using an in-house Python code and openbabel (version 3.0.0)³⁷. The hydrogens were added /removed in accordance to simulate pH 7.0 and gasteiger charges were added to PDBQT files. Only the largest and unique fragments were taken in each drug file.

Obtaining receptor files

From Protein Databank, (www.wwpdb.org)³⁸ among all structures of Human DHODH, only the asymmetric X-ray crystallography structures with a maximum resolution of 2.0 Å and published after 2010 are selected. Among these structures, some are also eliminated due to reading errors. The following PDB structures were remained: PDB ID's: 2wv8³⁹, 3kvj⁴⁰, 3kvl⁴⁰, 4igh⁴¹, 4jtu⁴², 4zl1⁴³, 4zmg⁴⁴, 5h2z⁴⁵, 5h73⁴⁶, 5hqe⁴⁷, 5k9c²⁷, 5k9d²⁷, 5mut⁴⁸, 5mvc⁴⁸, 5mvd⁴⁸, 6idj⁴⁹, 6j3b⁵⁰, 6j3c⁵⁰, 6jmd⁵¹, 6lzl⁵².

Docking analysis

The remaining 20 PDB files were then aligned to 6j3c PDB file as a reference using pairwise structure comparison tool on DALI Server⁵³. Only the first chain in each file (chain A) was used for the alignment (STable 1). The aligned protein structures were converted to PDBQT files,

using MGL Tools⁵⁴. All non-polar hydrogens were removed merging charges and polar hydrogens were added where necessary. The charges were recalculated using Kollman charges. All non-standard residues were removed. All water and other molecules except the main chain are removed.

Autodock Vina docking analysis

Since the all PDB structures were aligned, the active sites also were aligned. A single gridbox with a center at $x=34.569$, $y=-15.848$, $z=-18.938$ Å) and sizes of 18, 28 and 22 Å, at respective edges, encapsulated the active site and was used for all receptors.

Using an in-house Python code and Vina (Autodock), an automated docking was performed for all filtered ligands and receptor files⁵⁵. An exhaustiveness of 16 was used. All Vina docking analysis were performed on a computer with 12 core CPU and 32 GB ECC Ram. Each ligand that received an affinity equal or higher than -11.0 (kcal/mol) for any of the receptors was removed from future docking analysis. All dockings were started using 6j3c and continued with the rest, in alphabetical order according to PDB IDs.

Autodock 4.2 docking analysis

Same gridbox used for Vina with a center at $x=34.569$, $y=-15.848$, $z=-18.938$ Å) and sizes of 18, 28 and 22 Å, at respective edges, was used for docking using Autodock 4. Since Autodock 4.2 requires a step site, the default step size of 0.345 Å was used.

Identifying interacting aminoacids

Autodock docking analysis was displayed with MGLtools 1.5.6, and the protein and ligand were recorded in a single pdb file by selecting the best binding affinity from 10 different binding results. Receptor-ligand interactions were analyzed with the DS visualizer program of the obtained pdb.

***In silico* target predictions**

SMILE codes of the drug candiates were obtained from Pubchem and entered in <http://new.swisstargetprediction.ch/> and <http://www.pharmaexpert.ru/passonline> to search for probable targets in humans based on the similarity to known chemicals in the database^{56,57}.

Results and Discussion

Our study employed around 7900 structures from DrugBank³⁶ with 3d structures. In our effort to identify possible inhibitor candidates, we have performed molecular docking of these molecules to 20 DHODH crystal structures using a gridbox selected based on the known active site binding inhibitors. A distribution of all binding affinities towards to the first DHODH structures tested (6j3c) was obtained. Of this distribution, structures with affinity better than -11.0 kcal/mol, corresponding to top 12% percent, was selected for docking on the next DHODH structure (Figure1). Similarly, any structure that failed to achieve at least 11.0 kcal/mol binding affinity for any of the DHODH structures were filtered out for the rest of the study.

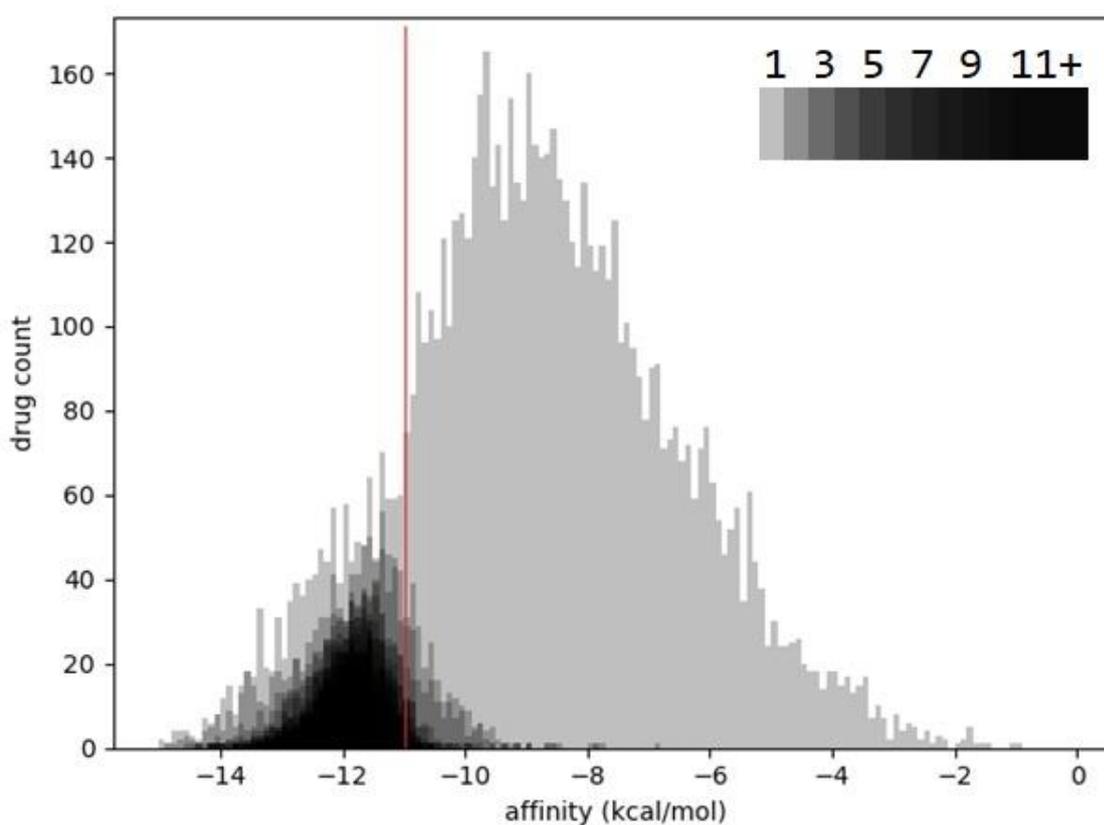


Figure 1. The distributions of drug affinities for each DHODH structure tested. The darker shades indicate the increased number of structures that recoded a drug at corresponding affinity. Red line indicates -11 kcal/mol, the threshold used to eliminate low affinity drug candidates.

The figure above, indicates the distribution of the drug affinities for all structures tested. The shades indicate the overlapping distributions, as a result, darker shades indicate where the most distributions are accumulated. As can be seen in the figure, the initial distribution obtained with

6j3c was centered around -9 kcal/mol. After the initial elimination the majority of the drugs were eliminated due to the threshold, and the remaining structures always showed a distribution with a mean of around -12 kcal/mol (Table1). A small portion of the drug candidates left over continued to spill above this threshold, resulting of their elimination as well. The continued docking and elimination process decreased the portion of the drugs eliminated in the upcoming analysis.

Table 1. Autodock vina and autodock 4.2 results of FDA approved drugs obtained by extensive molecular docking screening against DHODH structures.

DrugBank Code	Drug Name	PDB:6J3C Autodock Vina ΔG (kcal/mol)	PDB:6J3C Autodock 4.2 ΔG (kcal/mol)	Average of Autodock Vina ΔG Results (kcal/mol)	PDB:6J3C Autodock 4.2 Constant of Inhibition (Ki)
DB00398	Sorafenib	-12.7	-9.3	-12.365	152.27 nM
DB00450	Droperidol	-12	-7.83	-11.995	1.84 μM
DB00619	Imatinib	-11.9	-9.8	-12.53	65.63 nM
DB00734	Risperidone	-12.7	-10.16	-13.25	35.64 nM
DB01016	Glyburide	-12.4	-9.41	-12.27	126.51 nM
DB01067	Glipizide	-11.7	-9.53	-12.15	102.66 nM
DB01184	Domperidone	-12.2	-9.22	-12.155	174.27 nM
DB01238	Aripiprazole	-12.1	-8.81	-11.745	349.33 nM
DB01267	Paliperidone	-12.7	-9.84	-13.145	61.03 nM
DB06144	Sertindole	-11.6	-10.65	-12.17	15.71 nM
DB06684	Vilazodone	-12.3	-10.12	-12.195	38.36 nM
DB06817	Raltegravir	-11.9	-7.48	-12.09	3.31 μM
DB08883	Perampanel	-11.9	-9.87	-12.02	58.22 nM
DB08896	Regorafenib	-13.1	-9.17	-12.675	191.04 nM
DB08907	Canagliflozin	-11.9	-9.75	-11.79	71.00 nM
DB08930	Dolutegravir	-11.8	-8.9	-12.11	300.44 nM
DB09042	Tedizolid phosphate	-12.1	-9.38	-12.18	132.42 nM
DB09128	Brexpiprazole	-12.4	-12.1	-11.975	1.36 nM
DB11526	Masitinib	-12.6	-7.31	-13.145	4.37 μM
DB11703	Acalabrutinib	-12	-9.94	-12.15	52.01 nM
DB11732	Lasmiditan	-12.4	-7.2	-11.885	5.31 μM
DB11791	Capmatinib	-14.1	-9.68	-13.275	80.50 nM
DB11793	Niraparib	-11.2	-8.64	-11.745	462.16 nM
DB12836	Grapiprant	-12.2	-10.66	-12.28	15.41 nM
DB12867	Benperidol	-11.8	-8.67	-11.94	439.62 nM
DB12978	Pexidartinib	-12.4	-8.75	-12.35	388.56 nM
DB13931	Netarsudil	-12.7	-9.67	-12.65	81.36 nM
DB15305	Risdiplam	-11.8	-10	-11.815	46.83 nM

As a result of the molecular docking using Autodock Vina carried out on 20 different DHODH crystal structures and consecutive elimination of drugs with low affinity, 28 FDA approved and 79 clinically tested candidates were identified (Table 1, Table S2). These drugs are shown to be used effectively in a range of diseases such as rheumatoid arthritis, myeloid leukemia, schizophrenia, depression, HIV infection, spinal muscular atrophy and many types of cancer (Table 3). Among the drugs detected, Sorafenib^{58,59}, Regorafenib^{60,61}, Pexidartinib^{62,63}, Capmatinib^{64,65} are used as active drugs in many types of cancer as multiple kinase inhibitors. Raltegravir and Dolutegravir are used as anti-HIV drugs and clinical studies are continuing in covid-19 disease⁶⁶⁻⁶⁸. Glyburide, Glipizide and Canagliflozin are used in various type 2 diabetes diseases⁶⁹⁻⁷¹. Droperidol⁷², Risperidone⁷³, Domperidone⁷⁴, Aripiprazole^{75,76}, Paliperidone⁷⁷, Sertindole⁷⁸, Vilazodone⁷⁹, Brexpiprazole^{80,81}, Lasmiditan⁸² are used as serotonin, dopamine receptor antagonist in schizophrenia and depression diseases.

The FDA approved drugs were then docked again to 6j3c using Autodock 4.2 docking tool. This yielded affinity as well as an inhibition constant (Ki) for all 28 FDA approved drugs (Table1).

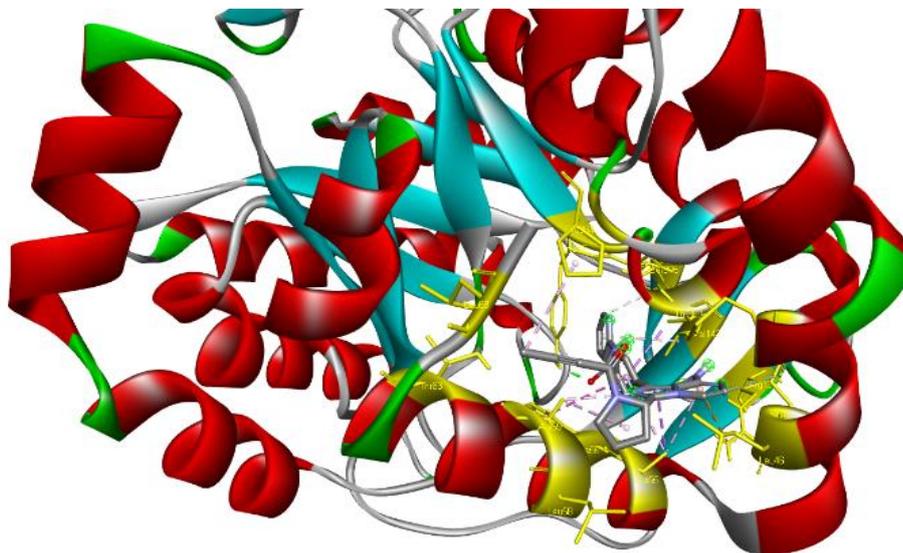


Figure 2. The five amino acids shown in with the highest number of interactions with the selected drugs are indicated with colored atom spheres, HIS56, VAL143, ALA59, ALA55 and TYR356.

The Autodock 4.2 docking results were then analyzed for their interaction with amino acids using DS Visualizer (Table 2) HIS56, VAL143 and ALA59 were the common amino acids that showed interaction with the majority of the drugs (20, 19 and 19 drugs, respectively). These were followed by ALA55, TYR356 (17 and 15 drugs, respectively). The amino acids showed

to be accumulated at the active site cavity, with ability to inhibit the enzyme at this position through competitive inhibition mechanism (Figure 2).

Table 2. The amino acids interacting the drugs detected through analysis of Autodock 4 docking results using DS Visualizer.

amino acids	acalabrutinib	aripiprazole	benperidol	brexipiprazole	canagliflozin	capmatinib	dolutegravir	domperidone	droperidol	glyburide	glypizide	grapiprant	imatinib	lasmiditan	masitinib	netarsudil	niraparib	paliperidone	perampanel	pexidartinib	regorafenib	risdiplam	risperidone	sertindole	sorafenib	tedizolid phosphate	vilazodone	total
THR285												X							X									2
ALA96										X									X									2
VAL333												X							X									2
ALA95								X			X							X	X									4
LYS255																			X									1
TYR356	X				X		X	X		X		X			X	X		X	X		X	X	X	X	X	X	X	15
ALA55	X	X				X		X		X		X	X	X	X	X	X	X		X		X		X	X		X	17
ARG136	X					X	X			X		X		X				X							X	X		9
PRO364	X		X	X										X			X			X	X	X	X	X				11
VAL143	X	X		X	X	X		X		X	X	X		X	X	X		X		X	X	X	X	X	X	X	X	19
ALA59	X		X	X		X		X	X	X	X	X		X	X	X		X		X	X	X	X	X	X	X	X	19
HIS56	X	X	X	X	X	X	X	X	X			X				X		X		X	X	X	X	X	X	X	X	20
THR360	X					X	X			X		X	X	X			X	X		X					X	X		12
SER305	X																											1
PHE62				X					X				X	X		X				X		X						7
TYR147				X	X																	X					X	4
MET111				X																X								2
PRO52				X			X			X		X						X		X								6
VAL134				X	X	X	X			X						X									X	X	X	9
LEU359		X	X	X	X		X			X	X						X			X	X	X	X				X	13
MET43		X				X	X	X					X	X		X	X			X		X		X	X	X	X	12
LEU46							X	X					X			X				X	X				X	X		8
LYS100													X												X	X		3
THR357					X					X			X								X	X						5
ASN145									X												X							2
ASN284					X																							1
LEU67														X														1
LEU58						X		X				X			X													4
TYR38						X									X													2
LEU42								X												X								2
PHE98																				X								1
PRO216																									X			1
PHE149																									X			1
THR63																							X					1

The possible targets were discovered with the SwissTargetPrediction and PASS prediction tools^{56,57}. In the SwissTargetPrediction, the interactions of 376,342 compounds with 3068 proteins consist of an experimentally proven data set. The query molecule is matched according to similarity among 376,342 compounds and the estimated target proteins are determined. 2D similarity is measured based on Tanimoto index and 3D similarity is measured based on Manhattan distance similarity quantity between ElectroScape 5D flat vector⁵⁶. This tool searches among known ligands based on 2D and 3D similarities to a given query to predict possible targets. In addition to SwissTargetPrediction the PASS (Prediction of Activity Spectra for Substances) web tool was also used for comparison. PASS web tool calculates using Bayesian approach by considering 2D structural similarities, only. PASS calculates probability for the query to be biologically active (Pa) or inactive (Pi) on target regions. This tool analyzes over 20,000 principal compounds and 180,000 biological related compounds using the MDDR database. In addition, PASS can analyze 3678 pharmacological activity⁵⁷.

Table 3. Drugbank codes, names, targets diseases of the discovered drugs and the diseases they are used in treatment of.

DrugBank Code	Drug Name	Target	Diseases
DB00398	Sorafenib	multiple intracellular (CRAF, BRAF and mutant BRAF) and cell surface kinases (KIT, FLT-3, VEGFR-2, VEGFR-3, and PDGFR-β)	Advanced Renal Cell Carcinoma Gastrointestinal Stromal Tumors Hemangiosarcoma Unresectable Hepatocellular Carcinoma Locally recurrent refractory to radioactive iodine treatment Thyroid carcinoma Metastatic refractory to radioactive iodine treatment Thyroid carcinoma
DB00450	Droperidol	Dopamine(2) receptor antagonism with minor antagonistic effects on alpha-1 adrenergic receptors	Agitation Chemotherapy-Induced Nausea and Vomiting (CINV) Delirium Nausea and vomiting
DB00619	Imatinib	inhibits the Bcr-Abl tyrosine kinase	treating chronic myelogenous leukemia (CML), gastrointestinal stromal tumors (GISTs) and a number of other malignancies.
DB00734	Risperidone	inhibition of dopaminergic D2 receptors and serotonergic 5-HT2A receptors	Acute Mania Irritability Mixed manic depressive episode

			Schizophrenia Agitated psychotic state
DB01016	Glyburide	the closure of ATP-sensitive potassium channels on beta cells	Gestational Diabetes Mellitus (GDM) Glycemic Control Type 2 Diabetes Mellitus
DB01067	Glipizide	A sulfonylurea medication used in Type 2 Diabetes to sensitize pancreatic beta cells and stimulate insulin release	Type 2 Diabetes Mellitus
DB01184	Domperidone	A specific blocker of dopamine receptors	Diabetic Gastroparesis Gastrointestinal Symptoms Upper gastrointestinal motility disorders
DB01238	Aripiprazole	agonism of dopaminic and 5-HT1A receptors and antagonism of alpha adrenergic and 5-HT2A receptors	Agitation Bipolar 1 Disorder Irritability Major Depressive Disorder (MDD) Mixed manic depressive episode Psychosis Psychotic Depression Schizophrenia Tourette's Disorder (TD) Acute Manic episode
DB01267	Paliperidone	central dopamine Type 2 (D2) and serotonin Type 2 (5HT2A) receptor antagonism. Paliperidone is also active as an antagonist at alpha 1 and alpha 2 adrenergic receptors and H1 histaminergic receptors	Delusional Parasitosis Schizoaffective Disorders Schizophrenia
DB06144	Sertindole	affinity for dopamine D2, serotonin 5-HT2A and 5-HT2C, and alpha1-adrenoreceptors	Schizophrenia
DB06684	Vilazodone	high affinity and selectivity for the 5-hydroxytryptamine (5-HT) transporter and 5-HT(1A) receptors	Major Depressive Disorder (MDD)
DB06817	Raltegravir	antiretroviral drug produced by Merck & Co., used to treat HIV infection	Human Immunodeficiency Virus Type 1 (HIV-1) Infection
DB08883	Perampanel	a noncompetitive AMPA glutamate receptor antagonist	Grand mal Generalized tonic-clonic seizure Partial-Onset Seizures
DB08896	Regorafenib	an orally-administered inhibitor of multiple kinases	Metastatic Gastrointestinal Stromal Tumor Locally advanced Gastrointestinal stromal tumor Refractory, metastatic Colorectal cancer Unresectable Gastrointestinal stromal tumor

DB08907	Canagliflozin	a sodium-glucose cotransporter 2 (SGLT2) inhibitor	Cardiovascular Events Cardiovascular Mortality End Stage Renal Disease (ESRD) Type 2 Diabetes Mellitus Elevated serum creatinine Hospitalization due to cardiac failure
DB08930	Dolutegravir	a HIV-1 integrase inhibitor	Human Immunodeficiency Virus Type 1 (HIV-1) Infection
DB09042	Tedizolid phosphate	generally effective against multidrug-resistant Gram-positive bacteria	Acute acute bacterial skin and skin structure infections
DB09128	Brexipiprazole	a novel D2 dopamine and serotonin 1A partial agonist	Major Depressive Disorder (MDD) Schizophrenia
DB11703	Acalabrutinib	Bruton Tyrosine Kinase (BTK) inhibitor	Chronic Lymphocytic Leukaemia (CLL) Mantle Cell Lymphoma (MCL) Small Lymphocytic Lymphoma
DB11732	Lasmiditan	selective agonism of the 5-HT1F receptor	Migraine Headache, With or Without Aura
DB11791	Capmatinib	a small molecule kinase inhibitor	Metastatic Non-Small Cell Lung Cancer
DB11793	Niraparib	orally active PARP inhibitor	Fallopian Tube Cancer Ovarian Epithelial Cancer Primary Peritoneal Cancer
DB12836	Grapiprant	prostaglandin receptor antagonists	The effects of grapiprant have been reported to be effective in the relief from arthritic pain in canine patients
DB12867	Benperidol	mechanism not clearly known	Dementia, Depression, Schizophrenia, Anxiety Disorders, and Psychosomatic Disorders, among others.
DB12978	Pexidartinib	a selective tyrosine kinase inhibitor (CSF1)	Symptomatic Tenosynovial Giant Cell Tumor
DB13931	Netarsudil	A Rho kinase inhibitor	Increased intraocular pressure
DB15305	Risdiplam	orally bioavailable mRNA splicing modifier	Spinal muscular atrophy (SMA)

In silico target prediction analysis of potential DHODH inhibitor candidates determined by comparing to FDA approved DHODH inhibitor, Leflunomide. Leflunomide is shown as probable inhibitor to other oxidoreductases, xanthine dehydrogenase, monoamine oxidase B,

arachidonate-5-lipogenase, cyclooxygenase-2 according to SwissTargetPrediction. Similarly, a number of drug candidates were also shown as probable inhibitors to the same oxidoreductases (Table 4). While only only acalabrutinib and raltegravir was found directly associated to DHODH, Aripiprazole, Benperidol, Brexpiprazole, Canagliflozin, Capmatinib, Domperidone, Droperidol, Grapiprant, Lasmiditan, Perampanel, Raltegravir, Risdiplam, Sertindole and Thedizolid phosphate have been found to be putative targets for other various oxidoreductases.

When compared using PASS, 4 out of 28 number of drugs were found directly associated to Dihydroorotase inhibition, as well as Leflunomide, suggesting a sound the experimental design. Capmatinib, Netarsudil, Regorafenib and Sorafenib are determined as Dihydroorotase inhibitor by PASS, contradicting SwissTargetPrediction results. However, an antiinflammatory activity was associated to a larger set including Dolutegravir, Lasmiditan, Mastinib, Netarsudil, Paliperidone, Perampanel, Pexidartinib, Risdiplam and Risperidone, most of which were not associated to DHODH inhibition directly (Table S3).

Table 4. Among the 28 drugs detected in our list, those with similarity to various oxidoreductase inhibitors are listed. The data were obtained from the SwissTargetPrediction tool. The score (Probability) between 0-1 was converted as a percentage. DHODH inhibitor, Leflunomide, is listed as reference.

Drug Name	Target	Common name	ChEMBL ID	Target Class	Probability (%)	Known actives (3D/2D)
Leflunomide	Dihydroorotate dehydrogenase (by homology)	DHODH	CHEMBL1966	Oxidoreductase	100	12/4
	Xanthine dehydrogenase	XDH	CHEMBL1929	Oxidoreductase	11.2	4/0
	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	11.2	40/0
	Arachidonate 5-lipoxygenase	ALOX5	CHEMBL215	Oxidoreductase	11.2	6/0
	Cyclooxygenase-2	PTGS2	CHEMBL230	Oxidoreductase	11.2	78/0
Acalabrutinib	Dihydroorotate dehydrogenase (by homology)	DHODH	CHEMBL1966	Oxidoreductase	11	18/0
	Cyclooxygenase-2	PTGS2	CHEMBL230	Oxidoreductase	11	343/0
	Inosine-5'-monophosphate dehydrogenase 2	IMPDH2	CHEMBL2002	Oxidoreductase	11	89/0

	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	11	319/0
Aripiprazole	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	10.6	62/15
	Egl nine homolog 1	EGLN1	CHEMBL5697	Oxidoreductase	10.6	20/0
	Arachidonate 5-lipoxygenase	ALOX5	CHEMBL215	Oxidoreductase	10.6	53/0
Benperidol	Egl nine homolog 1	EGLN1	CHEMBL5697	Oxidoreductase	11.5	51/0
Brexpiprazole	Arachidonate 5-lipoxygenase	ALOX5	CHEMBL215	Oxidoreductase	11.8	80/0
Canagliflozin	Glyceraldehyde-3-phosphate dehydrogenase liver	GAPDH	CHEMBL2284	Oxidoreductase	11.8	10/0
	Inosine-5'-monophosphate dehydrogenase 2	IMPDH2	CHEMBL2002	Oxidoreductase	11.8	34/0
Capmatinib	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	11.8	278/0
	Inosine-5'-monophosphate dehydrogenase 2	IMPDH2	CHEMBL2002	Oxidoreductase	11.8	173/0
Domperidone	Inosine-5'-monophosphate dehydrogenase 2	IMPDH2	CHEMBL2002	Oxidoreductase	10.6	99/0
	Cyclooxygenase-2	PTGS2	CHEMBL230	Oxidoreductase	10.6	502/0
Droperidol	Inosine-5'-monophosphate dehydrogenase 2	IMPDH2	CHEMBL2002	Oxidoreductase	11.5	155/0
Grapiprant	HMG-CoA reductase (by homology)	HMGCR	CHEMBL402	Oxidoreductase	11	161/0
	Arachidonate 5-lipoxygenase	ALOX5	CHEMBL215	Oxidoreductase	11	167/0
Lasmiditan	Cyclooxygenase-2	PTGS2	CHEMBL230	Oxidoreductase	11.3	18/0
Perampanel	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	11.3	116/0
	Dihydrofolate reductase	DHFR	CHEMBL202	Oxidoreductase	11.3	9/0
Raltegravir	Egl nine homolog 1	EGLN1	CHEMBL5697	Oxidoreductase	10.6	67/0

	4-hydroxyphenyl pyruvate dioxygenase	HPD	CHEMBL1861	Oxidoreductase	10.6	74/0
	Dihydroorotate dehydrogenase (by homology)	DHODH	CHEMBL1966	Oxidoreductase	0	12/0
	Monoamine oxidase A	MAOA	CHEMBL1951	Oxidoreductase	0	1/0
	Cyclooxygenase-2	PTGS2	CHEMBL230	Oxidoreductase	0	1/0
Risdiplam	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	10.6	9/0
Sertindole	Monoamine oxidase A	MAOA	CHEMBL1951	Oxidoreductase	11.8	56/0
	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	11.8	70/0
Thetizolid phosphate	Monoamine oxidase A	MAOA	CHEMBL1951	Oxidoreductase	15	0/11
	Monoamine oxidase B	MAOB	CHEMBL2039	Oxidoreductase	15	0/1

It is worth to note that, just like in Lenfunomide, most of these drug candidates were found as probable targets for not only oxidoreductases but also dopamine-serotonin receptors, attracting our attention.

As mentioned previously, 8 of the candidates, were found to be used as dopamin-serotonin receptor antagonists using SwissTargetPrediction. Oddly, in agreement with it, Leflunomide, as the known DHODH inhibitor but not being listed in our drug candidates, also showed high similarity to dopamine and norepinephrine transporter antagonists as well as the DHODH effectors. In addition, antagonists of other serotonin-receptors, 5HT-6, 5HT-2a, 5HT-2b, 5HT-2c, also show a degree of similarity with Leflunomide (Table 5).

The serotonin function is regulated by 7 members (15 subtypes) of the serotonin receptor in mammals⁸³. Immune cells express 5HT-1,5HT-2,5HT-3,5HT-4 and 5HT-7 class serotonin receptor, serotonin transporter (SERT), important enzyme for serotonin synthesis (TPH) and monoamine oxidase (MAO) for serotonin degradation⁸⁴.

Remarkably, among our candidates, Aripirazole, Domperidone, Lasmiditan, and Sertindole showed to be broad serotonin receptor antagonists as well as oxidoreductase inhibitors. Sertindole and Aripiprazole also have structural similarity with monoamine oxidase enzyme inhibitors. The fact that 5-hydroxytryptamine (5-HT) antagonists are also shown as probable

inhibitor for monoamine oxidase, which plays a role in serotonin degradation, and show high binding affinity to DHODH according to molecular docking analysis raises the question whether DHODH plays a role in the anti-inflammatory effect of 5-HT antagonists. In structural similarity analyzes, similar targets were found between leflunomide and 5-HT antagonist drugs, and 5-HT antagonist drugs were among the molecules with the highest binding affinity to the DHODH enzyme among the 7900 molecules. It has been demonstrated by other studies that 5-HT antagonists cause suppression of cytokine release in immune cells by causing serotonin blocking^{83,85-89}. However these studies did not consider or tested for DHODH inhibition that may be on the main pathway to the anti-inflammatory affect.

The aforementioned SwissTargetPrediction results also supported by the PASS analysis. Accordingly, besides our reference molecule, Leflunomide, Acalabrutinib, Aripiprazole, Benperidol, Brexpiprazole, Domperidone, Droperidol, Glipizide, Glyburide, Imatinib, Lasmiditan, Masitinib, Netarsudil, Nirapanib, Paliperidone, Risperidone, Sertindole and Vilazodone are predicted as antagonists for various members of 5-HTs. 10 of these drugs were also associated to mood disorder treatment.

Briefly, an apparent similarity between serotonin-dopamine receptor inhibitors and DHODH inhibitors was supported by both *in silico* analysis tools we have utilized. Despite, the anti-inflammatory effects of 5-HT antagonist drugs used in diseases such as schizophrenia and depression have not been fully elucidated. For this reason, it should be taken into consideration that possible DHODH inhibitory activities of the specified molecules may be responsible for the observed anti-inflammatory effect.

Table 5. List of discovered drugs showing similarity to serotonin-dopamine receptor antagonists. The data were obtained from the SwissTargetPrediction tool. The score (Probability) between 0-1 was converted as a percentage. DHODH inhibitor, Leflunomide, is listed as reference.

Drug Name	Target	Common name	ChEMBL ID	Target Class	Probability (%)	Known actives (3D/2D)
Leflunomide	Norepinephrine transporter	SLC6A2	CHEMBL222	Electrochemical transporter	100	20/1
	Dopamine transporter	SLC6A3	CHEMBL238	Electrochemical transporter	100	15/1
	Serotonin 6 (5-HT6) receptor	HTR6	CHEMBL3371	Family A G protein coupled receptor	11.2	13/0
	Androgen Receptor	AR	CHEMBL1871	Nuclear receptor	11.2	176/0

	Serotonin 2a (5-HT2a) receptor	HTR2A	CHEMBL224	Family A G proteincoupled receptor	11.2	33/0
	Serotonin 2c (5-HT2c) receptor	HTR2C	CHEMBL225	Family A G proteincoupled receptor	11.2	35/0
	Serotonin 2b (5-HT2b) receptor	HTR2B	CHEMBL1833	Family A G proteincoupled receptor	0	16/0
	Dopamine D2 receptor	DRD2	CHEMBL217	Family A G proteincoupled receptor	0	1/0
Aripiprazole	Serotonin 2b (5-HT2b) receptor	HTR2B	CHEMBL1833	Family A G proteincoupled receptor	100	175/29
	Serotonin 1b (5-HT1b) receptor	HTR1B	CHEMBL1898	Family A G proteincoupled receptor	100	492/35
	Serotonin 3a (5-HT3a) receptor	HTR3A	CHEMBL1899	Family A G proteincoupled receptor	100	74/1
	Serotonin 1d (5-HT1d) receptor	HTR1D	CHEMBL1983	Family A G proteincoupled receptor	100	471/14
	Serotonin 1a (5-HT1a) receptor	HTR1A	CHEMBL214	Family A G proteincoupled receptor	100	1215/127
	Serotonin 2a (5-HT2a) receptor	HTR2A	CHEMBL224	Family A G proteincoupled receptor	100	1507/160
	Dopamine D1 receptor	DRD1	CHEMBL2056	Family A G proteincoupled receptor	100	196/18
	Dopamine D2 receptor	DRD2	CHEMBL217	Family A G proteincoupled receptor	100	3320/423
	Dopamine D3 receptor	DRD3	CHEMBL234	Family A G proteincoupled receptor	100	1610/162
	Dopamine D4 receptor	DRD4	CHEMBL219	Family A G proteincoupled receptor	100	647/181
	Serotonin 2c (5-HT2c) receptor	HTR2C	CHEMBL225	Family A G proteincoupled receptor	100	484/47
	Serotonin transporter	SLC6A4	CHEMBL228	Electrochemical transporter	100	1188/102

	HERG	KCNH2	CHEMBL240	Voltage-gated ion channel	100	1035/39
	Serotonin 6 (5-HT6) receptor	HTR6	CHEMBL3371	Family A G proteincoupled receptor	100	371/43
	Serotonin 7 (5-HT7) receptor	HTR7	CHEMBL3155	Family A G proteincoupled receptor	100	602/55
	Serotonin 5a (5-H5a) receptor	HTR5a	CHEMBL3426	Family A G proteincoupled receptor	100	34/1
	Prostanoid EP2 receptor	PTGER2	CHEMBL1881	Family A G proteincoupled receptor	10.6	9/0
Domperidone	Serotonin 2b (5-HT2b) receptor	HTR2B	CHEMBL1833	Family A G proteincoupled receptor	100	60/4
	Serotonin 1a (5-HT1a) receptor	HTR1A	CHEMBL214	Family A G proteincoupled receptor	58.9	277/20
	Serotonin 2a (5-HT2a) receptor	HTR2A	CHEMBL224	Family A G proteincoupled receptor	100	413/35
	Serotonin 6 (5-HT6) receptor	HTR6	CHEMBL3371	Family A G proteincoupled receptor	10.6	204/13
	Serotonin 7 (5-HT7) receptor	HTR7	CHEMBL3155	Family A G proteincoupled receptor	10.6	128/45
	Serotonin 2c (5-HT2c) receptor	HTR2C	CHEMBL225	Family A G proteincoupled receptor	100	291/6
	Dopamine D2 receptor	DRD2	CHEMBL217	Family A G proteincoupled receptor	100	529/73
	Dopamine D3 receptor	DRD3	CHEMBL234	Family A G proteincoupled receptor	100	245/28
	Dopamine D4 receptor	DRD4	CHEMBL219	Family A G proteincoupled receptor	10.6	280/16
	Serotonin transporter	SLC6A4	CHEMBL228	Electrochemical transporter	100	128/25
	HERG	KCNH2	CHEMBL240	Voltage-gated ion channel	100	378/69

	Norepinephrine transporter	SLC6A2	CHEMBL222	Electrochemical transporter	100	10/31
Lasmiditan	Serotonin 1f (5-HT1f) receptor	HTR1F	CHEMBL1805	Family A G protein-coupled receptor	67.3	93/50
Sertindole	Serotonin 1b (5-HT1b) receptor	HTR1B	CHEMBL1898	Family A G protein-coupled receptor	100	535/241
	Serotonin 1d (5-HT1d) receptor	HTR1D	CHEMBL1983	Family A G protein-coupled receptor	11.8	488/218
	Serotonin 1a (5-HT1a) receptor	HTR1A	CHEMBL214	Family A G protein-coupled receptor	100	1452/128
	Serotonin 2a (5-HT2a) receptor	HTR2A	CHEMBL224	Family A G protein-coupled receptor	100	1689/119
	Dopamine D1 receptor	DRD1	CHEMBL2056	Family A G protein-coupled receptor	100	244/14
	Dopamine D2 receptor	DRD2	CHEMBL217	Family A G protein-coupled receptor	100	3894/335
	Dopamine D3 receptor	DRD3	CHEMBL234	Family A G protein-coupled receptor	100	1715/60
	Dopamine D4 receptor	DRD4	CHEMBL219	Family A G protein-coupled receptor	100	851/41
	Serotonin 2c (5-HT2c) receptor	HTR2C	CHEMBL225	Family A G protein-coupled receptor	100	649/41
	Serotonin transporter	SLC6A4	CHEMBL228	Electrochemical transporter	11.8	1598/275
	HERG	KCNH2	CHEMBL240	Voltage-gated ion channel	100	1071/49
	Serotonin 6 (5-HT6) receptor	HTR6	CHEMBL3371	Family A G protein-coupled receptor	100	600/59
	Serotonin 7 (5-HT7) receptor	HTR7	CHEMBL3155	Family A G protein-coupled receptor	11.8	842/22

Conclusion

On February 20, 2020, the clinical study of Leflunomide (FDA-approved DHODH inhibitor) was used in Phase-2 (240 patients) Covid-19 disease, however, results on 28 patients were recently shared as pre-print (Chictr2000030058). In the article published in Wuhan, China on March 12, 2020, it was announced that drugs that can suppress Sars-Cov-2 virus 10 times more effective than DHODH inhibitors such as Leflunomide and Teriflunomide were discovered. However, the structure of drugs S312 and S416 coded in this study has not been described. On April 28, the Phase-1 study started on 20 Covid-19 patients for the drug Leflunomide in the USA (ClinicalTrials Code: NCT04361214). Immun Therapeutics Company on April 22 proved that their medicines called IMU-838 are DHODH inhibitors. After their announcement that they will quickly try this drug in Covid-19 patients, in which they show that they have anti-viral and anti-inflammatory properties, Phase-2 and Phase-3 clinical studies were started in 600 patients in May 2020 (ClinicalTrials Code: NCT04379271).

The first clinical trial results of leflunomide (Chictr2000030058), an FDA-approved DHODH inhibitor, in Covid19 disease, Wang et al. (2020) published by pre-print⁹⁰. In an open-label controlled study conducted in Wuhan between 13 March and 17 April, leflunomide was found to be an effective drug in Covid19 disease. 12 patients were determined as SOC (standart of care) group and 15 patients as SOC + leflunomide group. After 14 days, leflunomide treatment showed that patients returned to a highly negative rate compared to the SOC group (80% versus %16.7 SOC). In consequence of the study, no adverse effects or deaths were detected.

With the help of *in silico* analyzes, drug repurposing studies are increasing day by day, targeting such as Sars-Cov-2 virus spike protein structure or RNA polymerases^{35,91-95}. However, considering that DHODH is a possible target for Covid19 treatment, not *in silico* analysis is available to our knowledge.

Here we have analyzed 7900 molecules with 3D structures for their putative binding at the active site of DHODH using docking analysis as a potential target candidate for Covid-19 treatment. Uniquely, we have provided a straight forward method to filter and identify the most probable antagonist candidates by repeating docking for several DHODH structures available. Any candidate that provided an affinity above the treshold is filtered until 28 FDA-approved and 79 clinically tested drugs were left. Next, the discovered drugs were analyzed using target prediction tools and the literature was scanned for their affect against Sars-Cov-2 virus. We have compared our candidates with FDA-approved DHODH inhibitor, Leflunomide. The only

other FDA-approved DHODH inhibitor, Teriflunomide, was not seen necessary for comparison due to high similarity to leflunomide at the first place.

Ellinger B. et al. (2020) screened 5632 candidate drugs on caco-2 cells against Sars-cov-2 virus. The multikinase inhibitors Sorafenib and Regorafenib in our list have been stated in the study published as that study and are among the effective molecules against Sars-cov-2 virus⁹⁵. In another study by Weston et al. (2020) published in another pre-print, 20 fda-approved drugs were analyzed against Sars-Cov-2 virus *in vivo* and *in vitro*. In this study, it is stated that imatinib plays a highly effective role against the virus with a selectivity index value of more than 9.5⁹³.

The effect of Acalabrutinib, known as BTK (Bruton Tyrosine Kinase) inhibitor, which is one of the drugs that we have identified as a possible DHODH inhibitor in our list, has been clinically researched and published by M. Roschewski et al. (2020) on June 5⁹⁶. As a hypothesis, the researchers predicted that acalabrutinib's BTK inhibitor activity can prevent ARDS in covid19 disease by blocking cytokine release in macrophages. Along with acalabrutinib treatment, a significant increase was observed in the oxygen respiration capacities of the patients and dramatic decrease was detected in IL-6 levels. Researchers suggest that acalabrutinib can be used effectively to suppress inflammation.

A study published in 24th of July by Riva L. *et al.* (2020), is quite remarkable when compared to our own data. Researchers have created a library called ReFRAME, which consists of 12,000 FDA-approved drugs and others in clinical trials. Molecules with an effective EC₅₀ value were selected, 21 were selected and 13 drugs were found to be effectively to suppress Sars-Cov-2 virus. When these experimentally identified drugs were compared to our findings, remarkable similarity can be seen. The fact that known target regions of these 13 selected molecules include serotonin and dopamine antagonists, coincides with the fact that 8 of 28 FDA approved drugs we have identified are too serotonin and dopamine antagonists. Especially, it was determined that Elopiprazole, a serotonin 1a-Dopamine D2 receptor antagonist, did not cause any changes in the number of cells by curb the infection in Vero6-Sars-Cov-2 infected cells at very low doses (EC₅₀: 1.6 μM). Elopiprazole is not included in Drugbank and, ergo, this study. Another effective molecule found in the study, Apilimod, was found effective in Phase II clinical studies against diseases such as rheumatoid arthritis, common variable immune deficiency (CVID). In addition, it has been reported by researchers that this drug has been found to suppress reproduction of EBOV, Lassa virus, Marburg viruses in human cells. Apilimod showed similarities with serotonin 2b-2c antagonists. It is noteworthy that the information obtained

about the apilimod and the given effects are similar to the expected effects as a result of DHODH inhibition, raising the question whether, an interaction between DHODH and the apilimod exists. When we analyzed this drug, which was eliminated only after seventh docking analysis (against 6j3c, 2wv8, 3kvj, 3kvl, 4igh, 4jtu and finally 4z11) showed affinity in a range of -11.0 and -12.9 kcal/mol according to Autodock Vina. In accordance our extensive potential DHODH inhibitor screening shows strong association between serotonin-dopamine antagonists and possible DHODH inhibitors compared to this study⁹⁷.

In summary, FDA approved drugs that may be more effective than Leflunomide were determined by molecular docking analysis. Target regions and structural similarities of these drugs with high binding energy are also discussed in detail. The similarities of Leflunomide drug structurally related to other oxidoreductase and dopamine, serotonin antagonists were analyzed.

Interestingly, it is quite remarkable that the drugs that show high binding energy against DHODH enzyme on our list have similarities to oxidoreductase inhibitors as well as the antagonist effects of dopamine and serotonin. We suggest that these drugs used for treatment of diseases such as schizophrenia and depression should be followed in Covid-19 patients. After determining the DHODH enzyme activation of 28 FDA approved drugs analyzed and determining cell culture Sars-Cov-2 virus activations, we recommend that the active drugs be tested clinically on Covid19 patients.

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