Genetic determinants of COVID-19 drug efficacy revealed by genome-wide CRISPR screens

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Abstract

Immunomodulatory agents dexamethasone and colchicine, antiviral drugs remdesivir, favipiravir and ribavirin, as well as antimalarial drugs chloroquine phosphate and hydroxychloroquine are currently used in the combat against COVID-19¹⁻¹⁶. However, whether some of these drugs have clinical efficacy for COVID-19 is under debate. Moreover, these drugs are applied in COVID-19 patients with little knowledge of genetic biomarkers, which will hurt patient outcome. To answer these questions, we designed a screen approach that could employ genome-wide sgRNA libraries to systematically uncover genes crucial for these drugs' action. Here we present our findings, including genes crucial for the import, export, metabolic activation and inactivation of remdesivir, as well as genes that regulate colchicine and dexamethasone's immunosuppressive effects. Our findings provide preliminary information for developing urgently needed genetic biomarkers for these drugs. Such biomarkers will help better interpret COVID-19 clinical trial data and point to how to stratify COVID-19 patients for proper treatment with these drugs.

Introduction

In response to the surging COVID-19 pandemic, several drugs have been reported to show efficacy in COVID-19 patients, but clear conclusions on some of them are still lacking. Among these drugs, dexamethasone and colchicine serve to suppress excessive immune response that damages tissues and organs in severe COVID-19 cases. Antiviral drugs such as remdesivir, favipiravir and ribavirin interfere with viral RNA polymerase and inhibit the replication of SARS-CoV-2. Antimalarial drugs chloroquine phosphate and hydroxychloroquine, whose use in COVID-19 is more controversial, suppress both immune response and viral replication. Currently little is known about the genetic biomarkers that modulate these drugs' efficacy. Such knowledge is urgently needed to better manage the COVID-19 pandemic.

Several groups of cellular proteins may affect drug efficacy. The list includes drug transporters, drug metabolizing enzymes, proteins that compete with the intended therapeutic target for drug binding, as well as proteins that mitigate or exacerbate drug-induced damage. Such proteins may differentially regulate the cellular stresses caused by a drug. If the stresses caused by COVID-19 drugs could result in massive cell death, it will be possible to capture genes that modify drug efficacy by analyzing cells that survived drug treatment (Fig. 1a).

However, most of the drugs used in COVID-19 treatment are not cytotoxic. They cannot efficiently kill commonly used cell lines even at very high concentrations (Fig. 1b), making it very difficult to conduct genome-wide screens. For antiviral drugs, there are additional difficulties that prevented large-scale screens. For example, if a sgRNA-mediated gene knockout in a host cell makes remdesivir more effective at suppressing viral replication, there is just no effective means to isolate that cell from other cells. Therefore, looking at drug's direct impact on viral replication *per se* in a pool-based large screen is hardly executable. As a result, no genome-wide screen studies on antiviral drugs have been reported and there is very little knowledge on genetic biomarkers for these groups of important drugs. In this report, we developed a screen platform that could effectively solve the above problems. Using such a platform, we performed genome-wide screening on COVID-19 drugs and identified

genes that critically regulate these drugs' activities. Screen results for these drugs are presented separately, with validation experiments shown at the end of this report.

Results

Screen system. The Eμ-Myc;Arf-/- cell line represents a good model system for interrogating gene-drug interactions¹⁷⁻²¹ and has led to renewed understanding of several important drugs^{19,21}. Importantly, with regard to drug screen, these cells are highly apoptosis-prone and readily undergo cell death at stress levels that cannot kill most other types of cells (Fig. 1b). The ability of Eμ-Myc;Arf-/- cells to amplify drug-induced perturbance into a cell death response is essential for the proposed screen (Fig. 1a), as it permits the extension of genome-wide screens onto many types of nontoxic drugs, including seven COVID-19 drugs in this study.

Using a retroviral backbone, we introduced a genome-wide sgRNA library ²² into cas9-expressing Eµ-Myc;Arf-/- cells at 200-fold coverage. For the vast majority of genes, there were 10 different sgRNAs per gene in the screen. Around 25 million of sgRNA-expressing cells were subjected to two rounds of treatment with a COVID-19 drug. Drug doses were adjusted so that approximately 10% to 20% of cells were alive at the lowest viability point (Extended Data Fig. 1). By examining which sgRNAs were highly enriched or depleted in surviving cells, we identified genes that modify the efficacy of these COVID-19 drugs.

Colchicine. Colchicine is an immunomodulatory agent used to suppress inflammation in gout, and more recently shown to greatly reduce cardiovascular events and death in chronic coronary disease²³. Multiple studies showed colchicine significantly lowered death rate in COVID-19 patients ¹⁻³. Mechanistically, colchicine destabilizes microtubule and damages mitotic spindles, thereby ceasing the proliferation of immune cells and alleviating COVID-19 symptoms. In our genome-wide screen, we identified a number of genes whose sgRNAs were significantly enriched after colchicine treatment (Fig. 2a). Showing the validity of this

screen system, ranked at No.10 was KIF2C (Fig. 2a'), a microtubule depolymerizing protein previously known to regulate the toxicity of microtubule destabilizers²⁴. The top hit gene in our screen was TMEM161B, an uncharacterized protein localized primarily at the plasma membrane. In addition, both FLCN and FNIP were identified as top hits in our screen, suggesting the FLCN-FNIP protein complex may regulate colchicine-induced cell death. Other top hits include RQCD1, PTBP1, KMT2D, AP1G1, POU2F1 and BRD7 (Fig. 2a'), most of which have not been previously implicated for microtubule drugs.

Prior to this study, we have completed genome-wide screens on taxol, vincristine and vinorelbine, three anticancer drugs that also target microtubule. To understand the performance of this screen system and whether certain genes are specific to colchicine, we referenced the other three screen results (Fig. 2b). We found that TMEM161B and POU2F1 affect cell viability in response to microtubule destabilizers colchicine, vincristine and vinorelbine, but not to taxol, which is a microtubule stabilizing drug. Other colchicine screen hit genes yielded similar phenotypes in all four microtubule drug screens, suggesting they are common crucial determinants of cell fate in response to microtubule disruption. Of note, these genes did not show up in our screens of other types of anticancer drugs (Fig. 2b). This analysis showed that for drugs of similar mechanisms of action, this screen system can consistently identify the same group of genes, highlighting Eu-Myc; Arf-/- cell as a high performance screen platform. Importantly, the genes identified in our screen for the first time pointed to a group of potential biomarkers for colchicine's use. If a COVID-19 patient carries genetic deficiency on one of these genes, colchicine may not sufficiently suppress the immune response to alleviate symptoms. This screen also shed light on the cytotoxic mechanisms of microtubule poisons, an important group of anticancer drugs whose death mechanisms are very unclear.

Dexamethasone. Dexamethasone is a glucocorticoid that suppresses lymphocyte proliferation. It is used to treat many types of autoimmune diseases and hematological malignancies. Multiple studies demonstrated its efficacy in lowering

death rate of COVID-19 patients⁴⁻⁶. As a cell line of B lymphocyte lineage, Eµ-Myc;Arf-/- cell serves as an ideal platform for interrogating dexamethasone's immunosuppressive mechanism. In dexamethasone screen, sgRNAs targeting transcription factor NR3C1, the glucocorticoid receptor for dexamethasone, were dramatically enriched (Fig. 2c). Such a finding is consistent with previous knowledge that dexamethasone binding of NR3C1 initiates a transcription program that is highly detrimental to lymphocytes²⁵. Our screen also recovered PTGES3, a NR3C1 chaperone protein required for glucocorticoid response²⁶, and TSC22D3, which is an immunosuppressive gene induced by glucocorticoids²⁷ (Fig. 2c). These results confirmed that the NR3C1-centered transcriptional program is central to dexamethasone's immune suppressive effect. Considering this, we further asked whether certain cellular proteins with transcriptional inhibitory activities might antagonize such a transcription program and protect lymphocytes from dexamethasone-induced cell death. sgRNAs against such genes will be significantly depleted after dexamethasone selection. We identified one gene that fits such category. In the dexamethasone screen, inhibitor of DNA Binding 3 (ID3) was one of the most severely depleted genes. ID3 is capable of inhibiting transcriptional activity and is important for the B cell lineage. It is possible that ID3 may serve to limit the dexamethasone-NR3C1 transcription response, thereby protecting B cells from dexamethasone. Taken together, NR3C1, PTGES3, TSC22D3 and ID3 may represent candidate genetic biomarkers for dexamethasone's usage in various diseases including COVID-19.

Ribavirin. Ribavirin is a nucleoside analog that exhibits broad-spectrum antiviral activities. A recent study showed that a triple-drug combination of ribavirin, lopinavir-ritonavir and interferon β1, or the ribavirin lopinavir-ritonavir combination significantly accelerated viral clearance compared with lopinavir-ritonavir alone⁷. This suggests ribavirin may exert anti-SARS-CoV-2 activity in patients.

To suppress virus replication, ribavirin needs to be gradually converted to NTP analog to gain the ability to inhibit RNA-dependent RNA Polymerase (RdRp).

Importantly, the same antiviral NTP analog will also interfere with vital cellular proteins including eIF4E and IMPDH²⁸ and become detrimental to mammalian cells. Consequently, a cell death-based screen, enabled by Eµ-Myc;Arf-/- cells (Fig. 1b), will help identity enzymes that are important for the NTP conversion of ribavirin, which will also shed light on the antiviral effects of ribavirin (Fig 1a). Indeed, the top hit in our ribavirin screen was adenosine kinase (Adk) (Fig. 3a), an enzyme that catalyzes the formation of ADP from adenosine. This is consistent with a previous report that ADK is a key determinant for the anti-HCV activity of ribavirin²⁹. As a proof of principle result, such an observation shows the validity of our screen rationale. Our screen uncovered two additional enzymes involved in nucleotide metabolism (Fig. 3a). Adenylosuccinate Synthase (ADSS) catalyzes the first step of the purine salvage pathway to form AMP. NUDT2 is a phosphatase that hydrolyzes dinucleotides to generate ATP or GTP. sgRNAs against these two genes were significantly enriched in cells that survived ribavirin treatment. This suggests that similar to ADK, both ADSS and NUDT2 may be critical to the NTP conversion and antiviral activity of ribavirin in cells. It may be interesting to analyze the SNP, copy number and expression level of these genes in responders and non-responders of ribavirin to explore potential biomarkers for COVID-19 ribavirin therapy.

Remdesivir. Currently, remdesivir is the only antiviral drug approved by FDA for COVID-19 treatment. Earlier *in vitro* studies and subsequent compassionate use of remdesivir indicated its potential efficacy in treating COVID-19³⁰⁻³². A randomized clinical trial in patients with severe disease failed to show superior outcome of remdesivir over control treatment⁸, whereas other studies suggested remdesivir could speed up recovery in COVID-19 patients ^{9,10}. In the latter studies, remdesivir closely missed the significance threshold in terms of reduction of death rate (7.1 % with remdesivir vs 11.9% with placebo). It is therefore interesting to explore whether certain genetic backgrounds render remdesivir ineffective and whether excluding such genetic non-responders from the treatment group might help push it over the

significance threshold.

Once imported into cells, remdesivir is gradually degraded to AMP analog, which is then converted to ATP analog to interfere with viral RdRp³³⁻³⁵. The same ATP analog may also compete with endogenous ATP to inhibit other cellular proteins, which may inflict damage. Importantly, such damage is enough to cause cell death in Eµ-Myc;Arf-/- cells (Fig. 1b), which allowed for a screen to discover genes crucial for remdesivir importation and NTP conversion. The two top hits in our remdesivir screen fell into such categories (Fig. 3b). We identified SLC29A3, a membrane transporter that facilitates the cellular uptake of chemical compounds. Among the more than 350 SLC proteins in the genome, only SLC29A3 was significantly enriched in our screen, suggesting its uniqueness for the importation of remdesivir. Our screen also identified HINT1 (Histidine Triad Nucleotide Binding Protein 1), which is an enzyme that breaks down phosphoramidate bonds on purine nucleotides. Remdesivir was designed such that once inside cells, it is converted to AMP analog in several steps³³ (Fig. 3c). Importantly, one of which involves breakage of such a phosphoramidate bond. Our screen results suggested this step, catalyzed by a cellular enzyme HINT1, could be rate-limiting in generating active remdesivir metabolites inside cells. Taken together, our study indicated SLC29A3 and HINT1 could be two cellular proteins crucial for the cellular concentration of remdesivir and its conversion into active metabolites, pointing to the possibility that genetic and expression differences in these two genes may affect COVID-19 patients' response to remdesivir treatment.

Favipiravir. Favipiravir is an antiviral prodrug that partially resembles a purine base. Favipiravir exhibited potent antiviral activity in SARS-CoV-2 animal models³⁶ and multiple clinical trials showed that favipiravir significantly shortened viral clearance time and sped up clinical recovery^{11,37-39}. In our screen, the enriched sgRNAs in cells that survived favipiravir treatment were mostly targeting cell death genes (Fig. 3d). It is possible that multiple proteins may play redundant roles in the importation and/or activation of favipiravir, therefore the enriched sgRNA sets could not identify specific drug importers or metabolizing enzymes for favipiravir. However,

the sgRNAs depleted from cells that survived favipiravir treatment indicated drug inactivation mechanisms, which are discussed below.

Drug export and inactivation mechanisms. In cells that survived drug selection, we were also able to analyze which sgRNAs were strongly depleted after drug selection. Such sgRNAs point to genes whose deficiency enhances drug efficacy and/or toxicity. We first examined the ABC family of drug exporters ⁴⁰. Such proteins pump chemicals out of cells, therefore their deficiency will increase cellular drug concentration and result in higher levels of death. We examined in our COVID-19 drug screens, sgRNAs against which ABC genes were the most prominently depleted (Fig. 4a and a', Extended Data Table 1). The results suggested that ABCC1 is a candidate exporter for remdesivir, and ABCC1, ABCB1 for colchicine.

As discussed before, the efficacy of antiviral drugs strongly depends on the cellular concentration of their respective NTP analogs. For nucleoside analogs, the first phosphorylation step to generate NMP analog is the rate-limiting step for activation. Consequently, nucleotidases that break down NMP analog could significantly reduce the antiviral efficacy of such drugs. Indeed, recent animal studies showed that for remdesivir, significant amounts of the AMP analog were dephosphorylated into inactive adenosine analog (depicted in Fig. 3c) ^{34,41}, which could greatly reduce the antiviral efficacy of remdesivir. Since this type of dephosphorylation is carried out by 5'-Nucleotidases⁴², we examined all 5'-Nucleotidase genes in our screen (Extended Data Table 2) and found that NT5C3B and NT5C2 are potentially responsible for such a dephosphorylation reaction for remdesivir. A similar analysis also suggested that NT5C2 is responsible for the dephosphorylation of NMP analog for favipiravir (Fig. 4b and b').

Notably, in the treatment of leukemia, NT5C2 activity is the most frequent cause of thiopurine treatment failure and disease relapse ⁴³. Through dephosphorylation of GMP analog formed by thiopurine, this enzyme inactivates thiopurine. Based on our screen results, we asked whether NT5C2 could similarly inactivate remdesivir and favipiravir. Indeed, increased cellular NT5C2 activity significantly reduced cytotoxicity

of both remdesivir and favipiravir in multiple cell lines (Fig. 4c and 4d), but had little effects on ribavirin. This suggested that analogous to the case of thiopurine, NT5C2 critically reduces the NMP analog of remdesivir and favipiravir, consequently reducing the level of active antiviral NTP analogs. Therefore, NT5C2 may also represent a crucial point for determining the antiviral efficacy of remdesivir and favipiravir.

Validation of sgRNA screen results. Our screens, leveraging NTP analog-induced cell death, discovered genes that affect the import, export and metabolism of antiviral drugs used in COVID-19 treatment. Prompted by the observation with NT5C2, we further validated top hits for three antiviral drugs in our study (Fig. 5a). Consistent with the screen data, sgRNAs targeting SLC29A3 and HINT1 significantly protected cells from remdesivir, whereas sgRNAs targeting ABCC1, NT5C2 and NT5C3B caused sensitization. Similar validation results were also seen for genes discovered in favipiravir and ribavirin screens, highlighting their important roles in regulating the efficacy of these drugs.

Due to lack of timely access, we could not further confirm our finding with Sars-CoV-2 models. As an alternative, we used Zika virus, whose replication⁴⁴ could be suppressed by remdesivir and ribavirin (Fig. 5b). We knocked out SLC29A3, HINT1, ABCC1 and NT5C2 in Huh7 cells and found that deficiencies of these genes indeed significantly affected remdesivir's antiviral efficacy (Fig. 5c). For example, HINT1 or SLC29A3 deficiency reduced remdesivir's antiviral efficacy by 30 to 100 folds. This demonstrated the essential roles of SLC29A3 and HINT1 in the importation and activation of remdesivir. Conversely, knockout of ABCC1 or NT5C2 caused significant enhancement of remdesivir's antiviral efficacy, confirming their importance as the exporter and inactivator of remdesivir.

In the case of ribavirin, knockout of genes identified in this study, including ADK, ADSS and NUDT2 also decreased ribavirin's antiviral efficacy (Fig. 5d). Taken together, these results confirmed the virological relevance of our screen strategy and provided the first set of potential genetic biomarkers for remdesivir.

Lastly, we also cloned sgRNAs to interrogate some of the top hits in colchicine and dexamethasone screens. We confirmed that knockout of such genes indeed conferred resistance or hypersensitivity to corresponding drugs (Fig. 5e), indicating they may modulate the immunosuppressive effects of these drugs. For example, TMEM161D, RQCD1, PTBP1, FLCN and FNIP knockout significantly protected cells from colchicine-induced death. Our results for the first time linked these genes to microtubule poisons. We also confirmed that ID3 knockout sensitized cells to dexamethasone. Taken together, screens conducted in this study will help advance the understanding of the action of COVID-19 drugs, potentially pointing to genetic biomarkers for disease management.

Discussion

Dexamethasone, colchicine, remdesivir, favipiravir and ribavirin have been shown to speed up SARS-coV-2 viral clearance or reduce COVID-19 death rate by various studies. The role of chloroquine and hydroxychloroquine in COVID-19 is more controversial 13-16,45-47, with larger studies of hospitalized patients arguing against their use 14-16. We also conducted CRISPR screens on chloroquine phosphate and hydroxychloroquine and identified nearly identical lists of resistance genes (Extended Data Fig. 2). Our results suggest that a group of genes involved in autophagy, as well as lipid and cholesterol synthesis significantly modulate the action of chloroquine drugs. Such a gene list may be useful for understanding chloroquine drugs' clinical use.

Due to the urgency of COVID-19 situation and lack of relevant knowledge, COVID-19 drugs have been applied with little consideration of genetic biomarkers. Given the rapid disease progression of COVID-19 and the presence of multiple treatment options, it is urgent to discover genetic modulators for each COVID-19 drug, such that patients may be matched to more effective drug choice. Here we present the first report on genome-wide CRISPR screens on these COVID-19 drugs, with data on both resistance and sensitization mechanisms.

For antiviral drugs, low toxicity and lack of executable screen endpoints presented two significant hurdles for genome-wide screens. Our approach, using NTP analog-induced cell death as a surrogate screen endpoint, overcame these obstacles and helped uncover the transport and metabolism mechanisms crucial for the antiviral efficacy of these drugs. From a method standpoint, this screen strategy is the first that is capable of employing genome-wide libraries to identify genetic biomarkers for antiviral drugs. Such a screen platform is broadly applicable for other types of nucleoside analogs used to treat existed and emerging viral diseases. Extended Data Figure 3 summarized our findings and showed the relative uniqueness of the drug transporters and metabolizing enzymes discovered for these COVID-19 drugs.

The clinical efficacies of several COVID-19 drugs including remdesivir remain under debate⁴⁸. Incorporation of genetic biomarkers of drugs may lead to better

assessment of their efficacy, as demonstrated by previous clinical trials of PARP inhibitors⁴⁹. Importantly, for nucleoside-based anticancer drugs that are structurally similar to antiviral drugs, numerous clinical studies have shown that SNPs and expression status of their relevant transporters and metabolizing enzymes dictated clinical outcome⁵⁰⁻⁵⁴. Therefore, the transportation, activating and detoxifying mechanisms uncovered in this screen may point to clinical biomarkers of COVID-19 drugs. If such knowledge could be established in the future, it may help better understand the efficacy of individual drugs in the context of COVID-19. For example, a SNP on NT5C2, present in 15% of the population, significantly affected the cellular activity of thiopurine and its incorporation into DNA⁵⁵. Our results would similarly suggest that patients with such genetic variation may respond differently to remdesivir and favipiravir, but not to ribavirin (Fig. 4c and 4d). In addition, medically relevant SNPs for many of the genes identified in our screen have been reported in human population. For example, deleterious SNPs of SLC29A3, which we discovered as the transporter for remdesivir, underly a diverse range of disease phenotypes including short stature, osteoporosis, diabetes, hyperpigmentation, chronic inflammation and hearing loss⁵⁶⁻⁵⁹. Given our finding that SLC29A3 deficiency greatly reduced remdesivir's antiviral effects (Fig. 5c), such a genetic background may call for further consideration on the use of remdesivir, yet the efficacy of favipiravir and ribavirin may be less affected. From a precision medicine perspective, our findings may provide rationales for pairing COVID-19 patients with proper drug choice, potentially improving the overall cure rate. Such knowledge may also be useful for the management of other diseases that utilize these drugs, such as influenza, Dengue and Zika virus infections.

Lastly, rational design of drug combination is still lacking for COVID-19 treatment. Our study may also shed light on such drug interaction. For example, given that NT5C2 causes the inactivation of both remdesivir and favipiravir, combinatorial use of both drugs may potentially saturate NT5C2 activity, which might lead to better antiviral outcome for these drugs.

There are several limitations to our study. As a loss of function screen, our study

cannot identify certain kinds of genes. For example, if several proteins function redundantly in a drug-metabolizing step, our screen will not be able to uncover such proteins either in the enriched or depleted sgRNA dataset. Genome-wide overexpression screens may help complement this study. Moreover, there are many discussions that drug efficacy in COVID-19 patients may be significantly influenced by the timing of treatment. Our screen cannot address that question either. Due to the urgency of the situation, we also did not have time to further delve into the mechanisms of some of the screen hits. We present our screen results in the current form to facilitate relevant research, in the hope that among the list of screen hits, drug biomarkers will be established, such that better understanding of COVID-19 treatment will be achieved to help combat the pandemic.

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Author Contributions:

H.J. conceived, supervised the study and wrote the manuscript. W.J. performed the screens and validation. A.Y. established the screen platform and participated in validation experiments in this study. J.M. analyzed screen data, prepared figures and participated in the earlier steps of the screen platform, including transferring the genome-wide sgRNA library into a retroviral backbone. A.Y., J.M. and W.J. completed anticancer drugs screens referenced in this study. The Zika virus validation experiments were performed by D. L. under the supervision of G. L. L. X., M. L., C. W., Z. H., S. C., J. Z., S.C., Q. J., Y.C. and L.S. participated in drug screens or provided technical helps to establish the screen protocol. Z. Z. and Y. Z. advised on and provided helps to the screen platform.

Competing interests:

The authors declare no competing interests.

Data and material availability:

All data is available in the main text or the supplementary materials. Plasmids generated in this study will be deposited to the plasmid sharing platform at BIO-Research Innovation Center Suzhou (www.brics.ac.cn/plasmid/?columnId=15) for distribution.

Figures

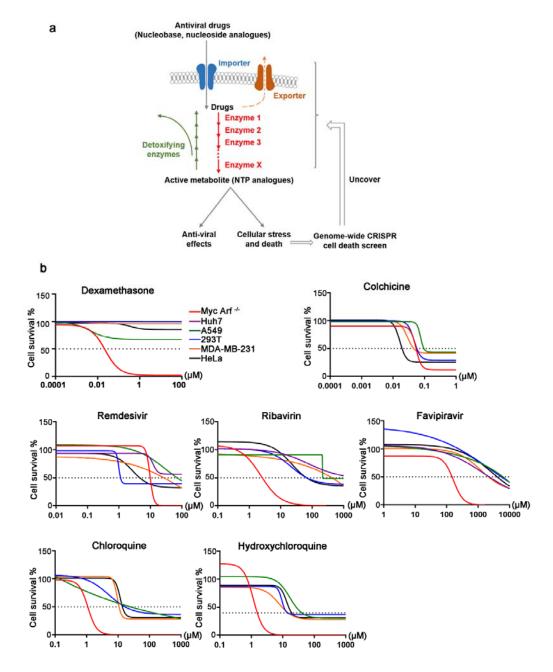


Figure 1. A genome-wide screen approach to identifying genes that modify the efficacy of COIVD-19 drugs. a) A diagram depicting the rationale of the screen. For example, antiviral drugs suppress viral replication by generating NTP analogs, which inhibit viral RNA polymerase. Such NTP analogs also interfere with cellular process that utilize normal NTPs, causing cellular stress that are sufficient to induce death in E μ -Myc;Arf-/- cells. By introducing genome-wide sgRNA library into such cells and analyzing which gene knockout enabled survival after antiviral drug treatment, it is possible to uncover genes that mediate the transport of these drugs, as well as genes

crucial for generating NTP analogs. Both classes of genes will also modify these drugs' antiviral efficacy. **b**) COVID-19 drugs efficiently killed Eµ-Myc;Arf-/- cells, but not other commonly used cell lines.

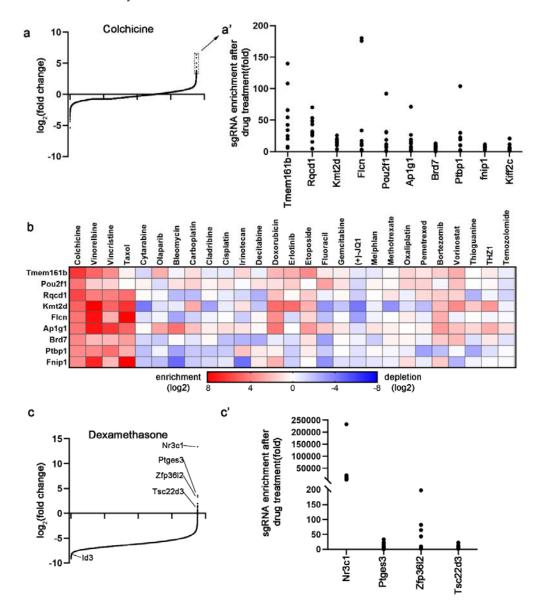


Figure 2. Genes enriched in colchicine and dexamethasone screens. a and c) Figures showing the overall distribution patterns of genes in the drug screen. Genes are sorted based on the fold change of sgRNAs after drug selection. Those most enriched genes are located at the far right. Top enriched genes are boxed out and detailed in a' and c'), which show the extent of enrichment of individual sgRNAs after drug selection. In A' and C'), each dot represents a sgRNA targeting the indicated gene. p values for all genes depicted in a' and c') are lower than 2.6×10⁻⁷. In c), p

value for Id3 depletion is 0.0007. **b)** How genes enriched in colchicine screen behaved in other drug screens conducted in our lab.

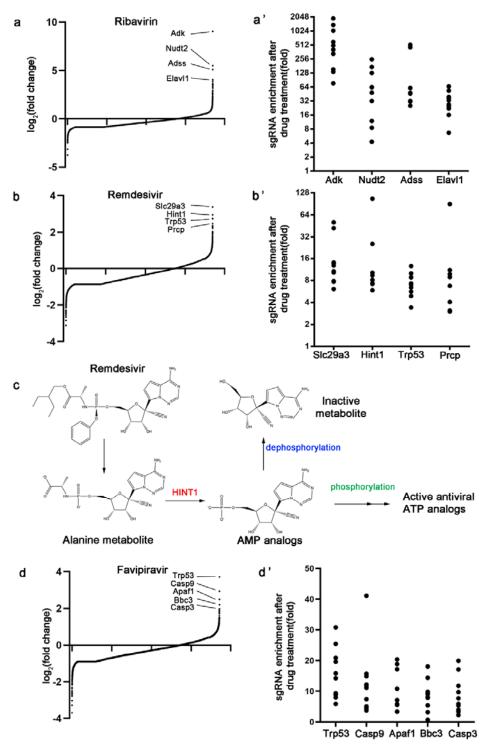


Figure 3. Genes enriched in antiviral drug screens. a, b and d) Figures showing the overall distribution patterns of genes in the drug screen. a', b' and d') Figures

showing the extent of enrichment of individual sgRNAs after drug selection. p values for all genes depicted in these panels are lower than 2.6×10⁻⁷. **c).** A diagram showing the metabolic process of remdesivir inside cells. One of the activation steps is carried out by HINT1.

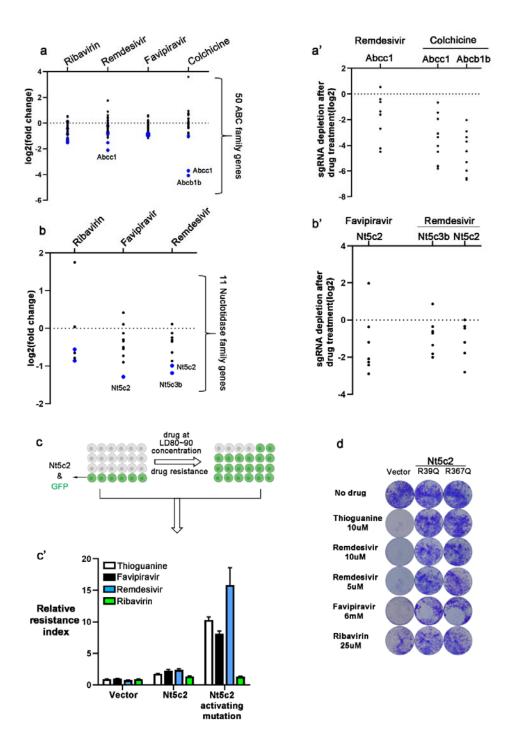


Figure 4. Analysis of the depleted gene sets with regards to ABC transporters and nucleotidases. a) and b) How genes belonging to the families of ABC transporters and nucleotidases behaved in the screens. Each dot represents a gene. Blue dots indicate genes whose depletion is statistically significant (p<0.05). a') and b') How individual sgRNA targeting the listed genes behaved in the drug screen. c) An

experiment setting to assess whether NT5C2 activity protects cells from drug-induced toxicity. Eµ-Myc;Arf-/- cells partially infected by NT5C2 and GFP-expressing virus were subjected to drug treatment at concentrations that killed 80~90% (LD80-90) of parental cells. Drug inactivation by NT5C2 enabled better survival of NT5C2&GFP expressing cells, leading to increase of GFP positive rate in surviving cells. c') Relative resistance index calculated from GFP percentages before and after drug treatment. Shown are Mean±SD from a triplicate experiment. Enhanced NT5C2 activity potently protected cells from remdesivir and favipiravir. Thioguanine was used as a positive control. d) Enhanced NT5C2 activity protected 293T cells from thioguanine, remdesivir and favipiravir.

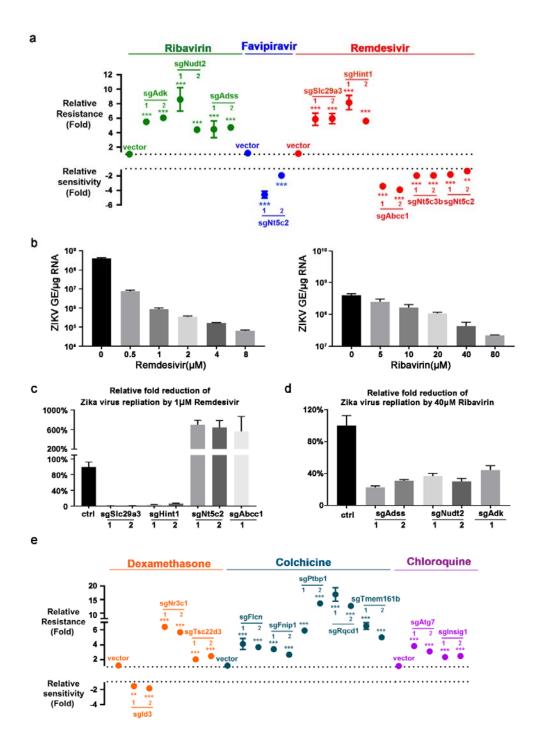
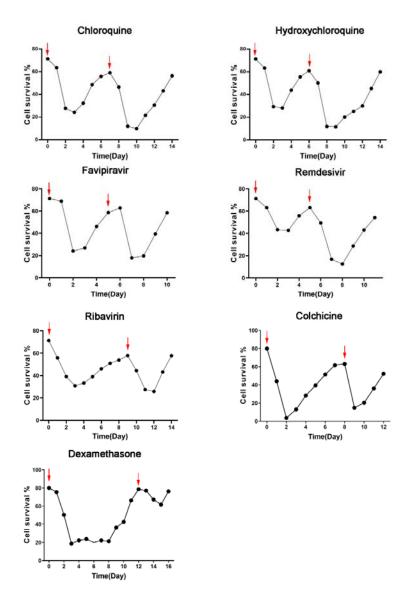


Figure 5. Validation of CRISPR screen results for COVID-19 drugs. a and e). Effects of indicated screen hit genes on drug sensitivity. sgRNAs targeting the indicated genes were expressed in Eμ-Myc;Arf-/- cells. An experimental setting similar to Fig.4c was used to assess whether a gene knockout caused resistance or sensitization to COVID-19 drugs. Shown are Mean±SEM from three independent

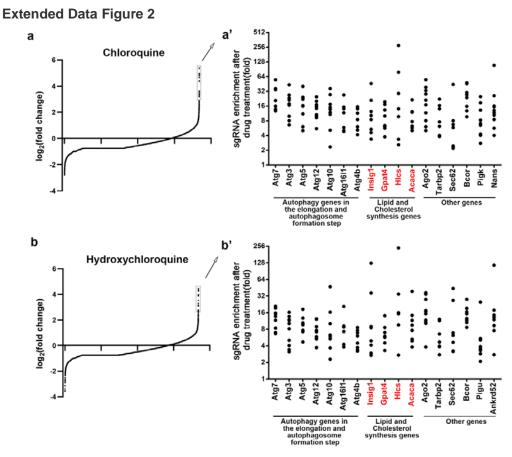
experiments. **p<0.05. ***p<0.01. **b)** Remdesivir and ribavirin suppressed Zika virus replication in Huh7 cells. Antiviral efficacies of remdesivir **(c)** and ribavirin **(d)** were impacted by genes identified in CRISPR screens. Shown are Mean±SD from a triplet experiment.

Extended Data Figure 1



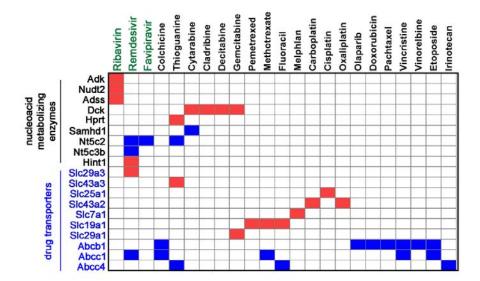
Extended Data Figure 1. Survival curves of cells during COVID-19 drug screens.

Eμ-Myc;Arf-/- cells were subjected to two rounds of treatment with the indicated drugs. Overall cellular viabilities were analyzed every day during the duration of the experiment. Blue arrows indicated the date when drugs were applied. Drugs were used at the following concentrations: remdesivir (16.4 μ M), ribavirin (20 μ M), favipiravir (400 μ M), chloroquine diphosphate (10 μ M), hydroxychloroquine sulfate (11.2 μ M), colchicine (25nM) and dexamethasone (4.7 μ M).



Extended Data Figure 2. Genes enriched in chloroquine and hydroxychloroquine screens. a and b) Figures showing the overall distribution patterns of genes in the drug screen. Top enriched genes are boxed out and detailed in a' and b'), which show the extent of enrichment of individual sgRNAs after drug selection. In a' and b'), each dot represents a sgRNA targeting the indicated gene. p values for all genes depicted in a' and b') are lower than 0.0002.

Extended Data Figure 3



Extended Data Figure 3. Summary and comparison of drug metabolizing enzymes and transporters identified in genome-wide CRISPR screens using our platform. Red boxes indicate enrichment of sgRNA against a gene in corresponding drug screen, whereas blue boxes indicate depletion.

Extended Dataset Table 1. sgRNA depletion data of ABC family genes

Ribavirin

Favipiravir

	neg p-value	neg fc	Тамрпа	neg p-value	neg fc
Abcc8	0.003077	-1.5156	Abca8b	0.23569	-1.15
Abcc3	0.034212	-1.4072	Abcb1a	0.031251	-0.9581
Abcc4	0.010486	-1.2617	Abcb4	0.070602	-0.924
Abcg1	0.16707	-1.1251	Abcb10	0.34441	-0.8991
Abcb6	0.14157	-0.93373	Abcb7	0.33918	-0.8991
Abca8b	0.30698	-0.9216	Abcc12	0.081291	-0.8991
Abca3	0.1954	-0.8579	Abcd4	0.37898	-0.8991
Abcb10	0.044315	-0.8579	Abcf1	0.1883	-0.8991
Abcb5	0.097301	-0.8579	Abcg1	0.12712	-0.8789
Abcb7	0.37865	-0.8579	Abcc8	0.022155	-0.8708
Abcc12	0.14951	-0.8579	Abcc3	0.059974	-0.8520
Abcc6	0.19855	-0.8579	Abcg4	0.01943	-0.8392
Abcd3	0.69236	-0.8579	Abca14	0.042137	-0.8376
Abcf1	0.52664	-0.8579	Abcf3	0.067539	-0.836
Abcc9	0.061263	-0.85457	Abcb6	0.14777	-0.8329
Abcg2	0.389	-0.71598	Abcc4	0.062047	-0.7817
Abcb1a	0.51934	-0.70515	Abcf2	0.29154	-0.7286
Abcc10	0.25211	-0.70248	Abcg3	0.41842	-0.663
Abcf3	0.15479	-0.69912	Abcb5	0.83967	-0.6569

Abcg3	0.62438	-0.67067	Abcc5	0.12365	-0.59984
Abcb11	0.62685	-0.67018	Abcc2	0.16517	-0.59918
Abcg8	0.36036	-0.64776	Abca9	0.74571	-0.59878
Abcd4	0.90214	-0.6191	Abcg5	0.39006	-0.55112
Abca16	0.23058	-0.56844	Abcc1	0.56369	-0.45893
Abcb1b	0.13487	-0.44343	Abcb8	0.64987	-0.43625
Abcg5	0.53517	-0.42675	Abce1	0.60958	-0.42984
Abca5	0.04259	-0.42594	Abca2	0.73498	-0.37159
Abca7	0.55151	-0.41768	Abcd2	0.71381	-0.36206
Abca6	0.2073	-0.38677	Abca8a	0.69848	-0.35601
Abcg4	0.47187	-0.38492	Abcg2	0.52123	-0.32787
Abca12	0.72019	-0.35919	Abcc6	0.86709	-0.27431
Abcb4	0.47863	-0.31863	Abcb11	0.84497	-0.2499
Abcf2	0.66141	-0.30544	Abca12	0.51611	-0.24117
Abca13	0.67504	-0.2721	Abcc9	0.71725	-0.21776
Abcd2	0.80491	-0.22473	Abca5	0.61769	-0.20874
Abcb8	0.68668	-0.18075	Abcg8	0.59331	-0.1691
Abcb9	0.78317	-0.17785	Abcd3	0.66523	-0.12106
Abca17	0.9393	-0.12117	Abca17	0.97923	-0.04641
Abca14	0.13763	-0.05418	Abca3	0.48619	-0.04173
Abca8a	0.79077	-0.03012	Abca6	0.48357	-0.02157
Abcc5	0.51849	-0.01037	Abca1	0.89728	0.052016

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Abca2	0.89116	0.005475
Abcd1	0.43639	0.019189
Abca15	0.99201	0.096685
Abca4	0.042978	0.12195
Abce1	0.71592	0.14598
Abca9	0.4097	0.15395
Abcc1	0.42793	0.18255
Abca1	0.87411	0.48318
Abcc2	0.66796	0.54094

Abca16	0.42702	0.063678
Abca7	0.56369	0.096084
Abcc10	0.084379	0.099543
Abca13	0.29636	0.12797
Abcb1b	0.78531	0.16157
Abca15	0.62245	0.21226
Abcd1	0.31849	0.22587
Abca4	0.49731	0.52231
Abcb9	0.35953	0.60835

Remdesivir

Colchicine

	neg p-value	neg fc		neg p-
Abcc1	0.020161	-2.0999	Abcb1b	2.5
Abcb1b	0.005281	-1.517	Abcc1	0.0
Abca8b	0.28956	-1.1338	Abcf3	0.0
Abcf3	0.17009	-1.1338	Abcb11	0
Abcc10	0.075594	-0.99255	Abca5	0
Abcb11	0.14203	-0.92101	Abca13	0
Abca5	0.097856	-0.86529	Abcc10	0
Abcb10	0.25776	-0.86529	Abca3	0
Abcb7	0.11927	-0.86529	Abcg8	0
Abcc6	0.27651	-0.86529	Abca6	0
Abcd3	0.68353	-0.86529	Abcd3	0

	neg p-value	neg fc
Abcb1b	2.58E-05	-4.0689
Abcc1	0.000232	-3.6995
Abcf3	0.094647	-1.1244
Abcb11	0.17268	-1.1216
Abca5	0.13853	-1.1065
Abca13	0.20858	-1.1044
Abcc10	0.05106	-1.0596
Abca3	0.13519	-1.0244
Abcg8	0.02603	-1.0122
Abca6	0.17949	-0.97968
Abcd3	0.37746	-0.97968
	·	

Abcd4	0.18106	-0.86529	Abcb5	0.61921	-0.97968
Abce1	0.38406	-0.86529	Abcc5	0.13853	-0.95108
Abcg4	0.003616	-0.83107	Abcc3	0.072588	-0.94427
Abcf1	0.81044	-0.81641	Abca8b	0.26477	-0.84231
Abcf2	0.47099	-0.78725	Abcg5	0.74008	-0.80633
Abcc3	0.065716	-0.78371	Abcg4	0.19529	-0.78522
Abcb1a	0.020161	-0.75848	Abca4	0.11447	-0.75759
Abca6	0.3429	-0.73881	Abcb4	0.075114	-0.74139
Abcc12	0.019469	-0.71439	Abcb8	0.058129	-0.73697
Abcg8	0.15236	-0.69372	Abcb10	0.091837	-0.73697
Abcb8	0.51458	-0.67643	Abcb7	0.23772	-0.73697
Abcc5	0.088522	-0.67444	Abca15	0.46594	-0.73697
Abca13	0.35866	-0.65848	Abcf1	0.53923	-0.73697
Abcg5	0.1584	-0.65766	Abcd4	0.69873	-0.73697
Abcg3	0.54099	-0.6203	Abcc8	0.37855	-0.38212
Abca3	0.23551	-0.56685	Abcg2	0.52186	-0.34495
Abcc4	0.091141	-0.4709	Abce1	0.35628	-0.33989
Abca15	0.82608	-0.44177	Abcb1a	0.066604	-0.30143
Abcb6	0.46295	-0.3611	Abcb9	0.63493	-0.24717
Abcc8	0.22414	-0.32037	Abcg3	0.63095	-0.23253
Abcg1	0.71025	-0.31868	Abca14	0.5661	-0.22138
Abca1	0.33156	-0.26337	Abcg1	0.46196	-0.2081

Abca2	0.30572	-0.24608	Abcd2	0.72039	-0.18562
Abcb4	0.01098	-0.21034	Abca12	0.3112	-0.07306
Abcb5	0.77433	-0.19624	Abcc9	0.28025	-0.06381
Abcc9	0.51618	-0.16856	Abca17	0.75542	-0.0436
Abcd2	0.11389	-0.1437	Abcc4	0.89082	0.007186
Abcg2	0.54454	-0.09267	Abcc6	0.5406	0.020976
Abca8a	0.24664	-0.05013	Abcd1	0.84116	0.034183
Abcb9	0.58398	0.046644	Abcc2	0.61052	0.067179
Abca16	0.99032	0.15467	Abca16	0.92919	0.15988
Abcc2	0.41512	0.18291	Abca8a	0.91745	0.32193
Abca17	0.79322	0.22082	Abcb6	0.57296	0.3775
Abca14	0.16074	0.24236	Abca1	0.85417	0.62195
Abca7	0.6593	0.37878	Abcc12	0.5637	0.66643
Abca9	0.7641	0.4662	Abca7	0.10436	0.72461
Abca12	0.3192	0.62697	Abca9	0.75756	0.7792
Abca4	0.055959	0.89422	Abca2	0.99453	0.94851
Abcd1	0.97373	1.7485	Abcf2	2.6E-07	3.5927

Extended Dataset Table 2. sgRNA depletion data of NT5 family genes

Ribavirin

neg|p-value neg|fc Nt5c 0.0048494 -0.8579 Nt5c1b 0.59428 -0.8579 Nt5c3 0.46134 -0.8579 Nt5c3b 0.18998 -0.78704 Nt5dc2 0.36552 -0.65088 Nt5c1a 0.097767 -0.56254 Nt5dc1 0.46624 -0.55633 Nt5dc3 0.39761 0.038616 Nt5m 0.85146 0.03982 Nt5e 0.95324 0.043502 Nt5c2 0.71592 1.7507

Remdesivir

	neg p-value					
Nt5c3b	0.0084289	-1.1827				
Nt5c2	0.032509	-0.98832				
Nt5c	0.26641	-0.86529				
Nt5c3	0.35866	-0.86529				
Nt5c1b	0.64403	-0.63843				
Nt5dc1	0.32554	-0.34892				
Nt5c1a	0.80472	-0.31396				
Nt5dc2	0.10876	-0.31347				
Nt5dc3	0.54549	-0.24357				
Nt5e	0.77701	-0.12608				
Nt5m	0.95614	0.11014				

Favipiravir

	neg p-value	neg fc
Nt5c2	0.0002462	-1.284
Nt5c	0.38607	-0.89911
Nt5dc2	0.26132	-0.72846
Nt5c1b	0.58077	-0.54048
Nt5c1a	0.14884	-0.50111
Nt5c3b	0.34501	-0.3387

Nt5c3	0.2497	-0.30023
Nt5m	0.84726	-0.13039
Nt5dc1	0.88353	0.10464
Nt5e	0.6698	0.41373
Nt5dc3	0.59289	0.41744

Extended Dataset Table 3 sequence of sgRNA used in

validation experiments

mouse	sglnsig1_2	TCAACCTGCTGCAGATCCAG
mouse	sgFlcn_1	GGAGCTCGCAGAAGTGGCAG
mouse	sgFlcn_2	GGAGCTCACTGATGATGCCG
mouse	sgFnip1_1	AATGCTGTCTGAGTACCTGA
mouse	sgFnip1_2	GCGGAGCGGATTAGGCGCGC
mouse	sgPtbp1_1	GGTTAACTACTATACATCGG
mouse	sgPtbp1_2	GGTTTCTTGAAGCGGTGCAG
mouse	sgRqcd1_1	CCAAATGAATGCCATAGCAT
mouse	sgRqcd1_2	GAAAGAACTCACCAAGATCA
mouse	sgTmem161b_1	GAGTCGAAGAGTATCAAATG
mouse	sgTmem161b_2	TTGCAAAAGACAAAACCAGG
mouse	sgld3_1	GGTGCGCGGCTGCTACGAGG
mouse	sgld3_2	GCTCTGCCAGGACCACCTGA
mouse	sgNr3c1_1	GGGGCAGTGCCACACTGCTG
mouse	sgNr3c1_2	AAAGCCATTGTCAAAAGGGA
mouse	sgTsc22d3_1	ACCTGAAGCCCCAGAAACCC
mouse	sgTsc22d3_2	AGAGGAGGTGGAGGTCCTAA

human	sgSLC29A3-1	TCGCCTACCTGTTGACAAGC
human	sgSLC29A3-2	GGATGACCGTCAGTGAGGCC
human	sgHINT1-1	CCAAAGATCGTGTCGCCACC
human	sgHINT1-2	CCTGAATAAGGGTTATCGAA

human	sgABCC1-1	CAAGTTCGTGAATGACACGA
human	sgNT5C2-1	TGTGAACCGAAGTTTAGCAA
human	sgNT5C2-2	ATATACATACCTGGTATATC
human	sgADSS-1	CCTTTCTAGGACTAGAAGGC
human	sgADSS-2	ATATCTGACAGAGCTCATAT
human	sgNUDT2-1	ATGAGCACCAAGCCTACCGC
human	sgNUDT2-2	TAACTAGGCCATGTGGAACC
human	sgADK-1	TGGAGGCGCCGCAAGCGCTG

References

- Scarsi, M. *et al.* Association between treatment with colchicine and improved survival in a single-centre cohort of adult hospitalised patients with COVID-19 pneumonia and acute respiratory distress syndrome. *Ann Rheum Dis*, doi:10.1136/annrheumdis-2020-217712 (2020).
- Deftereos, S. G. *et al.* Effect of Colchicine vs Standard Care on Cardiac and Inflammatory Biomarkers and Clinical Outcomes in Patients Hospitalized With Coronavirus Disease 2019: The GRECCO-19 Randomized Clinical Trial. *JAMA Netw Open* **3**, e2013136 (2020).
- Brunetti, L. *et al.* Colchicine to Weather the Cytokine Storm in Hospitalized Patients with COVID-19. *Journal of clinical medicine* **9**, doi:10.3390/jcm9092961 (2020).
- 4 Horby, P. *et al.* Dexamethasone in Hospitalized Patients with Covid-19 Preliminary Report. *N Engl J Med* (2020).
- Tomazini, B. M. *et al.* Effect of Dexamethasone on Days Alive and Ventilator-Free in Patients With Moderate or Severe Acute Respiratory Distress Syndrome and COVID-19: The CoDEX Randomized Clinical Trial. *JAMA*, doi:10.1001/jama.2020.17021 (2020).
- Sterne, J. A. C. *et al.* Association Between Administration of Systemic Corticosteroids and Mortality Among Critically III Patients With COVID-19: A Meta-analysis. *JAMA*, doi:10.1001/jama.2020.17023 (2020).
- Hung, I. F.-N. *et al.* Triple combination of interferon beta-1b, lopinavir-ritonavir, and ribavirin in the treatment of patients admitted to hospital with COVID-19: an open-label, randomised, phase 2 trial. *Lancet* **395**, 1695-1704,

- doi:10.1016/S0140-6736(20)31042-4 (2020).
- Wang, Y. *et al.* Remdesivir in adults with severe COVID-19: a randomised, double-blind, placebo-controlled, multicentre trial. *Lancet (London, England)* **395**, 1569-1578, doi:10.1016/S0140-6736(20)31022-9 (2020).
- 9 Beigel, J. H. *et al.* Remdesivir for the Treatment of Covid-19 Preliminary Report. *The New England journal of medicine*, NEJMoa2007764, doi:10.1056/NEJMoa2007764 (2020).
- Beigel, J. H. *et al.* Remdesivir for the Treatment of Covid-19 Final Report. *N Engl J Med*, doi:10.1056/NEJMoa2007764 (2020).
- 11 Fujifilm to seek approval for Avigan in Covid-19 after positive data,

 https://www.pharmaceutical-technology.com/news/fujifilm-avigan-covid-data/ (23

 September 2020).
- Arshad, S. *et al.* Treatment with Hydroxychloroquine, Azithromycin, and Combination in Patients Hospitalized with COVID-19. *Int J Infect Dis* (2020).
- d'Arminio Monforte, A., Tavelli, A., Bai, F., Marchetti, G. & Cozzi-Lepri, A. Effectiveness of hydroxychloroquine in COVID-19 disease: A done and dusted deal?

 Int J Infect Dis 99, 75-76 (2020).
- Geleris, J. *et al.* Observational Study of Hydroxychloroquine in Hospitalized Patients with Covid-19. *The New England journal of medicine*, NEJMoa2012410, doi:10.1056/NEJMoa2012410 (2020).
- Rosenberg, E. S. *et al.* Association of Treatment With Hydroxychloroquine or Azithromycin With In-Hospital Mortality in Patients With COVID-19 in New York State.

- JAMA, e208630, doi:10.1001/jama.2020.8630 (2020).
- Horby, P. *et al.* Effect of Hydroxychloroquine in Hospitalized Patients with Covid-19. *N Engl J Med*, doi:10.1056/NEJMoa2022926 (2020).
- Jiang, H., Pritchard, J. R., Williams, R. T., Lauffenburger, D. A. & Hemann, M. T. A mammalian functional-genetic approach to characterizing cancer therapeutics. *Nat Chem Biol* **7**, 92-100, doi:10.1038/nchembio.503 (2011).
- Pritchard, J. R. *et al.* Defining principles of combination drug mechanisms of action. *Proc Natl Acad Sci U S A* **110**, E170-179, doi:10.1073/pnas.1210419110 (2013).
- Bruno, P. M. *et al.* A subset of platinum-containing chemotherapeutic agents kills cells by inducing ribosome biogenesis stress. *Nat Med* **23**, 461-471, doi:10.1038/nm.4291 (2017).
- Ding, H. et al. Systematic Analysis of Drug Vulnerabilities Conferred by Tumor Suppressor Loss. *Cell Reports* 27, 3331-3344.e3336, doi:10.1016/j.celrep.2019.05.043 (2019).
- Bruno, P. M. *et al.* The primary mechanism of cytotoxicity of the chemotherapeutic agent CX-5461 is topoisomerase II poisoning. *Proceedings of the National Academy of Sciences* **117**, 4053-4060 (2020).
- Wang, T. *et al.* Gene Essentiality Profiling Reveals Gene Networks and Synthetic

 Lethal Interactions with Oncogenic Ras. *Cell* **168**, 890-903.e815,

 doi:10.1016/j.cell.2017.01.013 (2017).
- Nidorf, S. M. *et al.* Colchicine in Patients with Chronic Coronary Disease. *N Engl J Med*, doi:10.1056/NEJMoa2021372 (2020).

- Jost, M. et al. Combined CRISPRi/a-Based Chemical Genetic Screens Reveal that Rigosertib Is a Microtubule-Destabilizing Agent. Mol Cell 68, 210-223 e216 (2017).
- Wandler, A. M. *et al.* Loss of glucocorticoid receptor expression mediates in vivo dexamethasone resistance in T-cell acute lymphoblastic leukemia. *Leukemia* **34**, 2025-2037 (2020).
- Grad, I. et al. The Hsp90 cochaperone p23 is essential for perinatal survival.

 Molecular and cellular biology 26, 8976-8983, doi:10.1128/mcb.00734-06 (2006).
- Yang, H. *et al.* Stress-glucocorticoid-TSC22D3 axis compromises therapy-induced antitumor immunity. *Nature medicine* **25**, 1428-1441, doi:10.1038/s41591-019-0566-4 (2019).
- 28 Casaos, J. *et al.* The Use of Ribavirin as an Anticancer Therapeutic: Will It Go Viral? *Mol Cancer Ther* 18, 1185-1194, doi:10.1158/1535-7163.Mct-18-0666 (2019).
- 29 Mori, K. *et al.* Adenosine kinase is a key determinant for the anti-HCV activity of ribavirin. *Hepatology* **58**, 1236-1244, doi:10.1002/hep.26421 (2013).
- Wang, M. *et al.* Remdesivir and chloroquine effectively inhibit the recently emerged novel coronavirus (2019-nCoV) in vitro. *Cell research* **30**, 269-271, doi:10.1038/s41422-020-0282-0 (2020).
- Antinori, S. *et al.* Compassionate remdesivir treatment of severe Covid-19 pneumonia in intensive care unit (ICU) and Non-ICU patients: Clinical outcome and differences in post-treatment hospitalisation status. *Pharmacological research* **158**, 104899-104899, doi:10.1016/j.phrs.2020.104899 (2020).
- 32 Grein, J. et al. Compassionate Use of Remdesivir for Patients with Severe Covid-19.

- The New England journal of medicine, NEJMoa2007016, doi:10.1056/NEJMoa2007016 (2020).
- Warren, T. K. *et al.* Therapeutic efficacy of the small molecule GS-5734 against Ebola virus in rhesus monkeys. *Nature* **531**, 381-385, doi:10.1038/nature17180 (2016).
- 34 Williamson, B. N. *et al.* Clinical benefit of remdesivir in rhesus macaques infected with SARS-CoV-2. *bioRxiv*, 2020.2004.2015.043166, doi:10.1101/2020.04.15.043166 (2020).
- 35 Gao, Y. *et al.* Structure of the RNA-dependent RNA polymerase from COVID-19 virus. *Science* **368**, 779-782, doi:10.1126/science.abb7498 (2020).
- 36 Kaptein, S. J. F. *et al.* Favipiravir at high doses has potent antiviral activity in SARS-CoV-2-infected hamsters, whereas hydroxychloroquine lacks activity. *Proc Natl Acad Sci U S A*, doi:10.1073/pnas.2014441117 (2020).
- Cai, Q. et al. Experimental Treatment with Favipiravir for COVID-19: An Open-Label Control Study. Engineering (Beijing, China), 10.1016/j.eng.2020.1003.1007, doi:10.1016/j.eng.2020.03.007 (2020).
- 38 Chen, C. *et al.* Favipiravir versus Arbidol for COVID-19: A Randomized Clinical Trial. *medRxiv*, 2020.2003.2017.20037432, doi:10.1101/2020.03.17.20037432 (2020).
- Lou, Y., Liu, L. & Qiu, Y. Clinical Outcomes and Plasma Concentrations of Baloxavir Marboxil and Favipiravir in COVID-19 Patients: an Exploratory Randomized, Controlled Trial. 2020.2004.2029.20085761, doi:10.1101/2020.04.29.20085761 %J medRxiv (2020).
- 40 Liu, X. ABC Family Transporters. Adv Exp Med Biol 1141, 13-100,

- doi:10.1007/978-981-13-7647-4_2 (2019).
- Hu, W. J. *et al.* Pharmacokinetics and tissue distribution of remdesivir and its metabolites nucleotide monophosphate, nucleotide triphosphate, and nucleoside in mice. *Acta Pharmacol Sin*, doi:10.1038/s41401-020-00537-9 (2020).
- Składanowski, A. C. The role of soluble 5'-nucleotidases in the conversion of nucleotide analogs: metabolic and therapeutic aspects. *Curr Med Chem* **20**, 4249-4259, doi:10.2174/0929867311320340005 (2013).
- Tzoneva, G. *et al.* Activating mutations in the NT5C2 nucleotidase gene drive chemotherapy resistance in relapsed ALL. *Nature medicine* **19**, 368-371, doi:10.1038/nm.3078 (2013).
- Zhao, F. *et al.* Negligible contribution of M2634V substitution to ZIKV pathogenesis in AG6 mice revealed by a bacterial promoter activity reduced infectious clone. *Scientific reports* **8**, 10491, doi:10.1038/s41598-018-28890-0 (2018).
- Hoffmann, M. *et al.* Chloroquine does not inhibit infection of human lung cells with SARS-CoV-2. *Nature*, doi:10.1038/s41586-020-2575-3 (2020).
- Maisonnasse, P. *et al.* Hydroxychloroquine use against SARS-CoV-2 infection in non-human primates. *Nature*, doi:10.1038/s41586-020-2558-4 (2020).
- Arshad, S. *et al.* Treatment with hydroxychloroquine, azithromycin, and combination in patients hospitalized with COVID-19. *Int J Infect Dis* **97**, 396-403, doi:10.1016/j.ijid.2020.06.099 (2020).
- 48 Pan, H. *et al.* Repurposed antiviral drugs for COVID-19; interim WHO SOLIDARITY trial results. 2020.2010.2015.20209817, doi:10.1101/2020.10.15.20209817 %J

medRxiv (2020).

- Ledermann, J. *et al.* Olaparib maintenance therapy in patients with platinum-sensitive relapsed serous ovarian cancer: a preplanned retrospective analysis of outcomes by BRCA status in a randomised phase 2 trial. *The Lancet. Oncology* **15**, 852-861, doi:10.1016/s1470-2045(14)70228-1 (2014).
- Odin, E., Sondén, A., Gustavsson, B., Carlsson, G. & Wettergren, Y. Expression of Folate Pathway Genes in Stage III Colorectal Cancer Correlates with Recurrence Status Following Adjuvant Bolus 5-FU-Based Chemotherapy. *Molecular medicine*(Cambridge, Mass.) 21, 597-604, doi:10.2119/molmed.2014.00192 (2015).
- Okamura, Y. *et al.* Human equilibrative nucleoside transporter-1 expression is a predictor in patients with resected pancreatic cancer treated with adjuvant S-1 chemotherapy. *Cancer Sci* 111, 548-560, doi:10.1111/cas.14258 (2020).
- Cascorbi, I. Role of pharmacogenetics of ATP-binding cassette transporters in the pharmacokinetics of drugs. *Pharmacol Ther* **112**, 457-473, doi:10.1016/j.pharmthera.2006.04.009 (2006).
- Kulsoom, B., Shamsi, T. S. & Afsar, N. A. Gene expression of hENT1, dCK, CDA, dCMPD and topoisomerase IIα as an indicator of chemotherapy response in AML treated with cytarabine and daunorubicin. *Cancer Manag Res* 10, 5573-5589, doi:10.2147/CMAR.S181299 (2018).
- Dieck, C. L. & Ferrando, A. Genetics and mechanisms of NT5C2-driven chemotherapy resistance in relapsed ALL. *Blood* 133, 2263-2268, doi:10.1182/blood-2019-01-852392 (2019).

- Tulstrup, M. *et al.* NT5C2 germline variants alter thiopurine metabolism and are associated with acquired NT5C2 relapse mutations in childhood acute lymphoblastic leukaemia. *Leukemia* **32**, 2527-2535, doi:10.1038/s41375-018-0245-3 (2018).
- Cliffe, S. T. *et al.* SLC29A3 gene is mutated in pigmented hypertrichosis with insulin-dependent diabetes mellitus syndrome and interacts with the insulin signaling pathway. *Hum Mol Genet* **18**, 2257-2265, doi:10.1093/hmg/ddp161 (2009).
- Campeau, P. M. *et al.* Whole-exome sequencing identifies mutations in the nucleoside transporter gene SLC29A3 in dysosteosclerosis, a form of osteopetrosis. *Hum Mol Genet* **21**, 4904-4909, doi:10.1093/hmg/dds326 (2012).
- Hussain, K. *et al.* Diabetes mellitus, exocrine pancreatic deficiency, hypertrichosis, hyperpigmentation, and chronic inflammation: confirmation of a syndrome. *Pediatr Diabetes* **10**, 193-197, doi:10.1111/j.1399-5448.2008.00470.x (2009).
- Jonard, L. *et al.* Progressive hearing loss associated with a unique cervical node due to a homozygous SLC29A3 mutation: a very mild phenotype. *Eur J Med Genet* **55**, 56-58, doi:10.1016/j.ejmg.2011.06.009 (2012).

Materials and Methods:

Cell lines and drugs:

Eμ-Myc;Arf ^{-/-} cells were cultured in B-cell medium (45% DMEM, 45% IMDM and 10% FBS, supplemented with L-glutamate, and 5 μM β-mercaptoenthanol). Phoenix, HeLa, MDA-MB-231, A549, 293T and Huh7 were cultured in Dulbecco's modified Eagle's medium supplemented with glutamate and 10% (v/v) FBS. Remdesivir, ribavirin, favipiravir, chloroquine phosphate and hydroxychloroquine were provided by or purchased from MedChemExpress. Colchicine was purchased from Targetmol. Dexamethasone was purchased from Selleck.

Plasmids and generation of stable cell lines:

Murine or human NT5C2 cDNA and their activating mutant forms were synthesized by Genescripts. Cas9 was cloned to a retroviral vector. About 1% of Eμ-Myc;Arf ^{-/-} was infected by such Cas9-GFP retrovirus, from which we established a Cas9-expressing Eμ-Myc;Arf ^{-/-} stable cell line. Sequences of sgRNAs used in validation experiments are listed in Extended Data Table 3.

Transfer of sgRNA libraries to a retroviral backbone:

 $E\mu$ -Myc;Arf $^{-/-}$ cells cannot be efficiently infected with lentivirus. Therefore, we bulk transferred sgRNA sequences from the lentiviral library (Addgene: #1000000096) into a pBabe-SV40-puro backbone. Such a backbone can efficiently express sgRNAs and cause gene editing in Cas9-expressing $E\mu$ -Myc;Arf $^{-/-}$ cells. Next-generation sequencing confirmed the quality of the subclone library, with around 98% of sgRNAs still present.

First, the U6-sgRNA segment from LentiGuide-Puro(Addgene #52963) was PCR amplified and transferred to pBabe-SV40-puro to generate a pBabe-U6sgRNA-SV40-puro vector with Nhel and Xhol cloning sites.

Next, a mouse KO sgRNA pooled library (Addgene: #1000000096, 10sgRNA per gene) was amplified at 1000X fold coverage, and the sgRNAs were digested with Nhel/Xhol from the library. The insert was gel purified twice and concentrated using

the GeneJET Gel Extraction Kit(thermo scientific #K0692). pBabe-U6sgRNA-SV40-puro vector was digested with Nhel/Xhol. After ligating the vector and sgRNA inserts, 20 tubes of 5-alpha Electrocompetent E. coli (Takara,#9027) were transformed using electroporation and the subcloned library was obtained by combining 4 maxipreps. The library complexity was confirmed by streaking diluted bacteria onto plates and counting colonies. The total number of colonies was > 100x the size of the library.

To generate a mCherry-expressing sgRNA vector used in Fig.5a, the plasmid pBabe-U6sgRNA-SV40-puro was digested to remove puro fragment and the mCherry fragment was cloned into the plasmid to create pBabe-U6sgRNA-SV40-mCherry.

Q5 Hot Start High-Fidelity DNA Polymerase (NEB #M0531L) was used for all cloning PCR reactions.

Generation of sgRNA-expressing retrovirus and drug screen:

Retrovirus packaging was achieved in phoenix cells using standard protocols. Around 100 million of phoenix cells were transfected with retroviral sgRNA library (10sgRNA per gene). Virus-containing medium was collected from these cells 2 days after transfection, and added to 230 million of Eμ-Myc;Arf ^{-/-} cells for infection. Infected cells gained puromycin resistance from the retroviral backbone, which enabled us to determine that around 23 million of Eμ-Myc;Arf ^{-/-} cells were infected. At such an infection rate, most cells should only contain one sgRNA. 23 million of infected cells also roughly correlate to 100X coverage of the sgRNA library.

After expansion of these infected cells, they were divided and 25 million of such cells were each subjected to treatment with a drug. Around 48hrs after treatment, cell viability dropped to 10~20%. Cells were centrifugated and resuspended in normal culture medium to allow for recovery. 3~5 days after this, cells reached viability between 60% to 70%. They were again treated with drugs and similar protocols was followed to remove drugs after this second round of treatment. After recovery from the second round of drug selection, 25 million of alive cells were collected. 25 million of no-treatment cells were also collected as a control.

Genomic DNA of the collected cells was extracted with Magen HiPure Tissue DNA Midi Kits (D3122-03), and sgRNA sequences were PCR amplified using the Q5® Hot Start High-Fidelity 2X master Mix. From each sample, 40µg of genomic DNA was used for PCR reaction. 8 tubes of 50µl PCR reaction was set up for each sample. PCR was stopped at 23 cycles. Water control was used to ensure there was no aerosol contamination.

The 400µl PCR products were purified with thermo scientific GeneJET Gel Extraction Kit (K0692) and dissolved in 30µl dH2O. Such products were separated with 2.5% agarose gel to obtain the 250bp PCR products. After purification with thermo scientific GeneJET Gel Extraction Kit (K0692), the composition of sgRNAs was analyzed by NGS at Novagene (Beijing). After NGS, Analysis of raw sequencing reads was performed using MAGeCKFlute analysi⁶⁰. Fig. 2a and similar figures were generated from MAGeCKFlute analysis.

To generate the data for the type of analysis in 2A', we performed the following analysis. If a gene knockout induced by sgRNA1.1 protected cells from a drug, this sgRNA1.1 will be enriched in the drug-selected population. Before drug selection, sgRNA1.1 made up 0.0025‰ of all sgRNAs from non-treated samples. After drug selection, that percentage increased to 1‰. From such data, we can conclude that sgRNA1.1 was enriched by 400 fold after drug selection. Similar method was used if a sgRNA was relatively depleted after drug selection.

GFP or mCherry-based survival competition assay to assess relative resistance index conferred by cDNA expression or sgRNA knockout. Such experiments were used to produce data for Fig.4C' and Fig.5A and 5E. The overall experimental design was previously described in Jiang *et al*¹⁷. Briefly, NT5C2 cDNA or its activating mutant along with GFP was stably expressed in a small portion of Eμ-Myc;Arf -/- cells. Remdesivir was applied to such cells at concentration that would kill 80~90% of uninfected parental cells. Since NT5C2 inactivates remdesivir and protected cells from remdesivir-induced cell death, after drug treatment, most of the surviving cells

will be GFP positive that express NT5C2. If the GFP percentage rose from 20% in untreated cells to 80% in remdesivir-treated cells, it can be calculated that the **relative resistance index** is (0.8-0.8*0.2)/(0.2-0.8*0.2)=16, meaning that at such drug concentration, NT5C2-expressing cells are 16 fold as likely to survive remdesivir compared with control cells. The rationale behind such calculation and the biological meaning of relative resistance index is detailed in Jiang *et al* ¹⁷.

Certain events, such as ABCC1 knockout sensitizes cells to the killing by remdesivir. Therefore, sgABCC1 expressing, mCherry positive cells will die more after remdesivir treatment. In such scenario, we may observe that the mCherry positive rate drops from 50% in untreated cells to 20% in remdesivir-selected samples. From such numbers, it can be calculated that the **relative resistance index** is (0.2-0.2*0.5)/(0.5-0.5*0.2)=0.25, meaning ABCC1 knockout cells are **4 fold** as likely to die from remdesivir-induced killing. Accordingly, we will designate a **relative sensitization index** of **-4** to sgABCC1.

Virus production. Zika virus stock was prepared as previously described⁴⁴. Briefly, *in vitro* transcribed viral RNA was electroporated into Vero E6 cells, 72h-120h supernatants was harvested and titrated by plaque forming assay after removal of dead cells and cell debris.

Infection of Huh7 derived cell line. Cells were seeded in 48-well plates for 24 hours before incubation with Zika virus at multiplicity of infection (MOI) equal to 1. 4 hours later, the unbound viruses were replaced by fresh medium containing remdesivir or ribavirin. After 24 hours, cells were collected and lysed by TRIZOL(Tiangen) to extract RNA. Such RNA samples were subjected to reverse transcription (Toyobo) and ZIKV genome copies were analyzed by quantitative PCR(Toyobo) with primers targeting ZIKV NS3. The sequences for the qPCR primers are GTTTGGCTGGCCTATCAGGT (forward) and GTGTCTGGTCCACACCTCTG (reverse).

References not cited in main text:

60. B. Wang et al., Integrative analysis of pooled CRISPR genetic screens using

MAGeCKFlute. Nat Protoc 14, 756-780 (2019).