

Functional multi-omics reveals genetic and pharmacologic regulation of surface CD38 in multiple myeloma

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1 **Abstract**

2 CD38 is a surface ectoenzyme expressed at high levels on myeloma plasma cells and is the target
3 for the monoclonal antibodies (mAbs) daratumumab and isatuximab. Pre-treatment CD38 density
4 on tumor cells is an important determinant of mAb efficacy. Several small molecules have been
5 found to increase tumor surface CD38, with the goal of boosting mAb efficacy in a co-treatment
6 strategy. Numerous other CD38-targeting therapeutics are currently in preclinical or clinical
7 development. Here we sought to extend our currently limited insight into CD38 surface expression
8 by using a multi-omics approach. Genome-wide CRISPR-interference screens integrated with
9 patient-centered epigenetic analysis confirmed known regulators of *CD38*, such as RARA, while
10 revealing XBP1 and SPI1 as other key transcription factors governing surface CD38 levels. *CD38*
11 knockdown followed by cell surface proteomics demonstrated no significant remodeling of the
12 myeloma “surfaceome” after genetically-induced loss of this antigen. Integrated transcriptome and
13 surface proteome data confirmed high specificity of all-trans retinoic acid in upregulating CD38 in
14 contrast to broader effects of azacytidine and panobinostat. Finally, unbiased phosphoproteomics
15 identified inhibition of MAP kinase pathway signaling in tumor cells after daratumumab treatment.
16 Our work provides a resource to design strategies to enhance efficacy of CD38-targeting
17 immunotherapies in myeloma.

18

19 **Keywords:** myeloma, CD38, daratumumab, CRISPR interference, proteomics

20

1 **Introduction**

2 Harnessing the immune system to treat myeloma has rapidly become the most exciting therapeutic
3 frontier in this disease. The first such immunotherapy agent to achieve United States Food and Drug
4 Administration (FDA) approval was the monoclonal antibody (mAb) daratumumab¹. Daratumumab
5 targets CD38, a cell surface ectoenzyme highly expressed on myeloma plasma cells. Daratumumab
6 is currently FDA-approved for use as either monotherapy or combination therapy in the
7 relapsed/refractory setting, or front-line therapy in combination with other small molecule agents¹.
8 A second mAb targeting CD38, isatuximab, was also recently approved for relapsed/refractory
9 myeloma; at least 15 additional CD38-targeting agents are in development². Extensive and
10 encouraging clinical data has already been obtained with daratumumab, though resistance appears
11 to inevitably occur^{3,4}. Biologically, this process appears to be quite complex, with determinants of
12 resistance ranging from alteration of surface antigens on tumor cells⁴⁻⁶ to dysfunction of the tumor
13 immune microenvironment^{7,8}. While it remains unclear whether CD38 downregulation on tumor
14 cells after mAb treatment is a marker of resistance^{5,9}, or, instead, successful therapy¹⁰, compelling
15 preclinical and clinical data suggests that CD38 surface antigen density prior to treatment strongly
16 correlates with mAb efficacy^{5,11}.

17 This latter observation has led to numerous efforts to identify small molecules which can
18 increase tumor surface antigen density of CD38, representing potential co-treatments with CD38-
19 targeting mAbs. The first such example of a CD38-boosting small molecule was all-trans retinoic
20 acid (ATRA)¹². Subsequent studies identified the pan-histone deacetylase (HDAC) inhibitor
21 panobinostat¹³, the thalidomide analog lenalidomide¹⁴, the Janus kinase (JAK) inhibitor
22 ruxolitinib¹⁵, and the DNA methyltransferase inhibitor azacytidine¹⁶ as agents that could lead to
23 myeloma surface CD38 increase. A clinical trial combining ATRA with daratumumab has led to
24 encouraging outcomes in patients previously refractory to daratumumab¹⁷.

1 While these published strategies suggest ways to improve CD38 mAb outcomes, they also
2 leave many questions unanswered. Most notably, we do not yet have a broad global sense of the
3 transcriptional or post-transcriptional networks that most strongly impact CD38 expression. Bi- and
4 tri-specific antibodies¹⁸ and chimeric antigen receptor (CAR) T cells¹⁹ targeting CD38 are also in
5 clinical development. As seen for similar modalities against other targets²⁰, efficacy of these novel
6 agents, in addition to mAbs, is likely to also be impacted by CD38 antigen density on tumor cells.
7 Furthermore, prior studies showed that CD38 downregulation after daratumumab treatment was
8 accompanied by increases in the complement inhibitors CD55 and CD59 (ref.⁵). Are there other
9 features of myeloma surface remodeling driven by CD38 downregulation? And, for the small
10 molecules noted above, it is unknown how they more generally impact the makeup of the myeloma
11 cell surface proteome beyond CD38. The tumor cell surface not only harbors opportunities for
12 immunotherapeutic targeting but also is the interface for communication between tumor and
13 microenvironment, potentially leading to other alterations in myeloma biology after changes in
14 surface proteomic profile. To address these questions, here we have taken advantage of CRISPR
15 interference-based functional genomic screens, cell surface proteomics, epigenetic analyses, and
16 phosphoproteomics to provide a multi-omic perspective on CD38 regulation and tumor cell
17 consequences of targeting CD38 in myeloma.

18

19 **Methods**

20 *CRISPR interference screening and hit validation*

21 Genome-wide CRISPRi screening was performed as described previously²¹. Briefly, RPMI-8226
22 cells stably expressing dCas9-KRAB were transduced with a genome-wide library comprised 5
23 sgRNA/protein coding gene. After 14 days cells were stained for surface CD38 and flow-sorted to
24 enrich for populations of cells expressing low or high cell surface levels of CD38. Cell populations

1 were then processed for next-generation sequencing as previously described²² and sequenced on a
2 HiSeq-4000 (Illumina). Reads were analyzed by using the MAGeCK pipeline as previously
3 described²³. Further validation was performed by knockdown with individual sgRNA's extracted
4 from the genome-wide library with conformation by flow cytometry or Western blotting. Antibody-
5 dependent cytotoxicity assays were performed using NK92-CD16 cells as described previously¹⁶.
6 Additional details in Supplementary Methods.

7

8 *Epigenetic analysis and machine learning for CD38 transcriptional regulation*
9 Publicly available ATAC-seq data from primary myeloma samples (ref.²⁴) was analyzed with the
10 Homer tool findPeaks. Motif binding in the identified ATAC peak regions was called with PROMO
11 tool²⁵. Newly-diagnosed patient tumor RNA-seq data in the Multiple Myeloma Research
12 Foundation CoMMpass trial (MMRF; research.themmr.org) was used to correlate expression of
13 predicted transcription factors with *CD38* expression. To build a predictive model for *CD38*
14 expression as a function of transcription factor expression, we developed an XGBoost (Extreme
15 Gradient Boosting) model with randomized search with cross validation to find optimal parameters.
16 80% of CoMMpass data was used for training and the remainder for model testing. Additional
17 details in Supplementary Methods.

18

19 *Cell surface proteomics and phosphoproteomics*
20 Cell surface proteomics was performed using an adapted version of the *N*-glycoprotein Cell Surface
21 Capture²⁶ method, as we have described previously²⁷. Unbiased phosphoproteomics was performed
22 using immobilized metal affinity (IMAC) chromatography using methods described previously²⁸.
23 All samples were analyzed on a Thermo Q-Exactive Plus mass spectrometer with data processing in
24 MaxQuant²⁹. Additional details in Supplementary Methods.

1

2 **Results**

3 *A CRISPR interference-based screen reveals regulators of CD38 surface expression*

4 We first sought to use an unbiased approach to identify regulators of surface CD38 in myeloma

5 tumor cells. We specifically employed genome-wide screening with CRISPR interference

6 (CRISPRi), which leads to much higher specificity of knockdown than shRNA while avoiding

7 potential toxicity of double-strand breakage with CRISPR deletion³⁰. We recently used this

8 approach to characterize regulators of surface B-cell Maturation Antigen (BCMA) in myeloma²¹.

9 Here, we employed an RPMI-8226 cell line with the dCas9-KRAB machinery, required for

10 CRISPRi, as described previously²¹. We confirmed that this RPMI-8226 cell line robustly expressed

11 CD38 (**Supp. Fig. 1A**).

12 The genome-wide screen was performed as shown in **Fig. 1A**. Briefly, RPMI-8226 cells

13 were transduced with a pooled genome-wide sgRNA library. After 14 days the cells were then

14 stained with fluorescently-labeled anti-CD38 antibody and flow sorted into low- and high-CD38

15 populations. Frequencies of cells expressing each sgRNA was quantified using next generation

16 sequencing. As an important positive control, increasing confidence in the screen results, we first

17 noted that knockdown of *CD38* itself strongly decreased surface CD38 expression (**Fig. 1B**). On the

18 other hand, several dozen genes, when repressed, did indeed lead to increased surface CD38 (right

19 side of volcano plot in **Fig. 1B; Supp. Table 1**). As another positive control, one of these top hits

20 included *RARA*, whose degradation is catalyzed by ATRA treatment to drive CD38 increase¹².

21 To find pathways that may be useful for pharmacologic targeting, we first applied Gene

22 Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) analysis to the list of

23 genes that, when inhibited, significantly increased CD38 (**Fig. 1C**). We were intrigued to find that

24 many of the strongest effects appeared to be driven by transcriptional or other epigenetic factors.

1 These specifically included pathways such as “DNA replication”, “mRNA processing”, “DNA-
2 templated transcription”, and “spliceosome”.

3 We considered whether any hits associated with these pathways may be “druggable”, with
4 the goal of expanding our repertoire of small molecules that enhance surface CD38 in myeloma.
5 *SS18*, a component of the BAF (BRG1/BRM associated factor) chromatin remodeling complex,
6 scored highly as a hit. However, treatment with the proposed BAF inhibitor caffeic acid phenol
7 ester (CAPE)³¹ did not lead to consistent increases in surface CD38 (**Supp. Fig. 1B**). Similarly, the
8 lysine demethylase *KDM4A* was a prominent hit, but treatment with the inhibitory metabolite (*R*)-2-
9 hydroxyglutarate³² also had no effect (**Supp. Fig. 1B**).

10 The strongest hits for genes whose knockdown increased surface CD38 were two
11 transcription factors, *HEXIM1* and *TLE3*. Validation studies using individual sgRNA knockdown
12 confirmed increased surface CD38 (**Fig. 1D**, **Fig. 2A** and **Supp. Fig. 2A**), as well as functional
13 impact in NK cell antibody-dependent cellular cytotoxicity (ADCC) assays with daratumumab (**Fig.**
14 **2B**). However, these proteins are known to be widespread negative regulators of transcription^{33,34},
15 suggesting little scope for specific therapeutic targeting at the *CD38* locus.

16 We were surprised that other targets proposed to increase CD38 expression after
17 pharmacologic inhibition, such as HDACs³, or catalyzed degradation, such as IKZF1/3 (ref.²⁵), did
18 not appear as prominent hits (**Fig. 1B**). However, this result may reflect a limitation of functional
19 genomic screens. A pharmacologic agent may inhibit multiple members of a protein class to drive a
20 phenotype, whereas, with single gene knockdown, functional redundancy may prevent this
21 phenotype from appearing³⁵ (i.e., multiple HDACs may need to be ablated at once, or both IKZF1 and
22 IKZF3 simultaneously, to drive increased CD38). We speculate this is the case with DNA
23 methyltransferases (DNMTs). We previously showed that treatment with the DNMT inhibitor
24 azacytidine, which promotes degradation of all cellular DNMTs³⁶, could robustly increase surface

1 CD38 (ref.¹⁶). Here, however, we found that knockdown of any individual DNMT only led to minor
2 CD38 increase (**Fig. 1B**).

3 Given these findings, we therefore shifted our focus to genes that, when knocked down, led
4 to CD38 decrease (left side of volcano plot in **Fig. 1B**). We reasoned this approach could still reveal
5 important biological inputs that regulate the surface expression of CD38. Examining specific genes,
6 we found that the transcription factor *SPI1* was the strongest hit besides *CD38* that, when knocked
7 down, repressed surface CD38 expression. We also noted that *NFKB1* and *NFKB2* knockdown
8 appeared to drive CD38 decrease. This finding was intriguing given the known importance of NF-
9 κ B signaling in myeloma proliferation and survival³⁷. KEGG and GO analysis of genes whose
10 knockdown significantly decreased CD38 showed enrichment for MAP kinase pathway and protein
11 phosphorylation more broadly (**Supp. Fig. 1C**), suggesting key roles for intracellular signaling in
12 regulating surface CD38.

13 Validation experiments with individual sgRNAs confirmed that *SPI1* knockdown strongly
14 decreased CD38 surface expression by flow cytometry, with a lesser decrease in surface CD38 with
15 *NFKB2* knockdown (**Fig. 1D**, **Fig. 2C**, **Supp. Fig. 2C-D**). These alterations also led to functional
16 impacts. RPMI-8226 cells with knockdown of these genes showed significantly decreased NK cell
17 lysis in ADCC assays (**Fig. 2D**). We further probed this dynamic *in vivo*, finding that RPMI-8226
18 cells with *SPI1* knockdown were relatively resistant to daratumumab in a murine model (**Fig. 2E**).
19 We note that we attempted to expand these results to additional cell lines. However, our four other
20 myeloma cell lines harboring the CRISPRi machinery²¹ all express extremely low levels of *SPI1*
21 (**Supp. Fig. 2B**) and attempted knockdown in two of them (AMO1, KMS12PE) did not elicit any
22 phenotype (not shown). Therefore, this finding suggests that *SPI1* may play an important role in
23 regulating CD38 expression in some myeloma tumors, but it is less likely to be a universal
24 regulator.

1

2 *Epigenetic analysis suggests XBP1 as a key determinant of CD38 in primary myeloma tumors*

3 Our CRISPRi results suggest that epigenetic and/or transcriptional regulation is a critical
4 driver of surface CD38 levels. However, we do not know whether these specific hits in a myeloma
5 cell line will extend to primary myeloma tumors. We therefore took a complementary approach to
6 find potential transcriptional regulators of *CD38*. Using ATAC-seq data from 24 primary myeloma
7 tumor samples²⁴, we extracted open chromatin motifs near the *CD38* promoter (**Supp. Fig. 3A**) to
8 identify a list of 46 transcription factors with potential binding sites at this locus (**Supp. Table 2**).
9 We then correlated expression (via Pearson *R*) of these transcription factors with *CD38* expression
10 across 664 primary patient tumors at diagnosis in the Multiple Myeloma Research Foundation
11 CoMMpass database (research.themmr.org; release IA13). In this analysis, we found the
12 transcription factor most negatively correlated with *CD38* expression was *RARA* (**Fig. 3A**),
13 consistent with our CRISPRi screen data, and underscoring the promise of ATRA as a co-treatment
14 to increase *CD38*. Intriguingly, the transcription factor with strongest positive correlation was *XBP1*
15 (**Fig. 3A-B**), a central driver of plasma cell identity³⁸. *XBP1* also showed strong positive
16 correlations with *CD38* in two other patient tumor gene expression datasets^{39,40} (**Supp. Table 2**).

17 To further extend this analysis, we sought to build a predictive model which could estimate
18 *CD38* transcript level as a function of transcription factor expression. We used an XGBoost method
19 applied to CoMMpass mRNA-seq data to find weights of transcription factor expression that most-
20 influence *CD38* levels in patient tumors. We first tested this analysis on 80% of patient data as a
21 training set with 20% left out as test set. We found this model, solely based on transcription factor
22 expression, could predict about half of the variance (coefficient of variation $R^2 = 0.49$ using 5-fold
23 cross validation) in test set *CD38* levels (**Fig. 3C**). Using model weights and Shapley Additive
24 Explanations (SHAP) analysis (see Supplementary Methods) to determine transcription factors that

1 have the greatest impact, either positive or negative, on CD38 expression, we found that *XBP1*
2 played the strongest role overall. Other strong hits from both of our analyses included *IRF2*, *ATF1*,
3 and *STAT1*. *SPI1* also appeared in the top 10 most relevant transcriptional regulators (**Fig. 3D**;
4 **Supp. Fig. 3B**), consistent with our CRISPRi results, suggesting that *SPI1* may play a key role in
5 regulating *CD38* in a subset of tumors.

6 We further evaluated *XBP1* given its prominent role in these two complementary
7 bioinformatic analyses. In a prior dataset of shRNA knockdown of *XBP1* in myeloma plasma
8 cells⁴¹, *CD38* mRNA was decreased ~3-fold after *XBP1* silencing (**Supp. Fig. 3C**). This finding
9 was consistent with results in our CRISPRi screen, where *XBP1* knockdown led to significant
10 surface *CD38* decrease (**Fig. 1B**). We further validated this relationship by using a doxycycline-
11 inducible sgRNA construct for CRISPRi vs. *XBP1*, finding a clear dose-response between degree of
12 *XBP1* knockdown and loss of surface *CD38* by flow cytometry (**Supp. Fig. 3D-G**). Supporting
13 relevance of this link, a recent myeloma tumor whole genome sequencing study found that deletion
14 of *XBP1* was one of the strongest determinants of clinical response to daratumumab⁴². We
15 attempted to perform promoter activation assays to directly link *XBP1* binding to *CD38* expression,
16 but were unable to successfully generate a reporter construct that reflected all 8 loci where *XBP1*
17 may bind at *CD38* regulatory elements (**Supp. Fig. 3H**). Taken together, these results nominate
18 *XBP1* as a particularly strong determinant of surface *CD38* in myeloma plasma cells, though future
19 investigation will be required to validate a direct or indirect relationship to *CD38* transcription.

20

21 *No consistent large-scale remodeling of the myeloma surface proteome after CD38 downregulation*

22 We next evaluated *CD38* surface regulation from the perspective of monoclonal antibody
23 therapy. In clinical samples, *CD38* loss after daratumumab was accompanied by increases in *CD55*
24 and *CD59*, which may inhibit complement-dependent cytotoxicity and contribute to daratumumab

1 resistance⁵. In preclinical studies, macrophage trogocytosis has been proposed as a mechanism
2 contributing to CD38 loss after mAb treatment, that also leads to alterations in other surface
3 antigens including CD138/SDC1 (ref.⁷). However, we hypothesized that given its enzymatic
4 activity and role as a cellular differentiation marker⁴³, loss of CD38 on its own may influence
5 surface expression of other myeloma antigens. Such alterations may reveal new biology or
6 (immuno)therapeutic vulnerabilities of CD38 mAb-treated disease.

7 To test this hypothesis, we employed a method we recently developed termed “antigen
8 escape profiling”²⁷. We used CRISPRi to transcriptionally repress *CD38* in RPMI-8226, AMO-1,
9 and KMS12-PE myeloma cell lines, using this genetic approach to partially mimic loss of surface
10 antigen seen after mAb therapy (**Fig. 4A**). We then performed Cell Surface Capture proteomics^{26,27}
11 to uncover surface proteome alterations in a relatively unbiased fashion. Across cell lines, analyzed
12 in biological triplicate with *CD38* knockdown vs. non-targeting sgRNA, we quantified 897 proteins
13 annotated as membrane-spanning in Uniprot (minimum of two peptides per protein) (**Supp. Table**
14 **3**). As a positive control, in all lines we found that the strongest signature was decrease of CD38
15 itself (**Supp. Fig. 4A-C**). However, when aggregating proteomic data, we found no significant
16 alterations in any surface antigens beyond CD38 itself (**Fig. 4B**). Integration with RNA-seq data
17 revealed only THY-1/CD90 as upregulated >3-fold at both the mRNA and surface proteomic level
18 after *CD38* knockdown (**Fig. 4C**). Intriguingly, CD90 is known as a marker of “stemness” in early
19 hematopoietic lineage cells that is lost when CD38 expression is increased⁴⁴. CoMMpass analysis
20 also confirmed increased *THY1* expression in tumors with lower *CD38* (**Supp. Fig. 4D**). However,
21 further validation as to whether CD90 is truly altered after CD38 mAb will require pre- and post-
22 treatment clinical specimens, beyond the scope of our work here. Overall, we conclude that loss of
23 CD38 in isolation leads to limited remodeling of the myeloma surface proteome.

24

1 *Integrated surface proteomic and transcriptional analysis suggests ATRA is highly specific in CD38*
2 *upregulation*

3 Data from our group¹⁶ and others¹²⁻¹⁵ have suggested that several small molecules can
4 increase myeloma surface CD38. However, the broader impacts of these agents on membrane
5 antigens beyond CD38 have not been directly compared. We performed integrated cell surface
6 proteomics and transcriptional analysis of RPMI-8226 cells treated with 10 nM all-trans retinoic
7 acid (ATRA), 2 μ M azacitidine (Aza), and 10 nM Panobinostat (Pano), all treated for 72 hr, in
8 comparison to DMSO (**Supp. Table 4**). These doses are chosen as they have been previously
9 published to significantly increase myeloma surface CD38 by flow cytometry^{12,13,16}. In this
10 integrated analysis we found much broader impacts of azacytidine and panobinostat than ATRA on
11 the “surfaceome” of plasma cells, beyond increasing CD38 (**Fig. 5A**). These results suggest that, at
12 doses driving CD38 upregulation, for Aza or pano altering CD38 is just a small component of their
13 impact on myeloma tumor cells, whereas ATRA is much more specific in driving CD38
14 upregulation.

15 Toward understanding how CD38 is modulated after drug treatment, in our previous work¹⁶
16 we noted that the mechanism of *CD38* increase after Aza treatment was unclear. We thus further
17 investigated the global transcriptional response (i.e. not limited to membrane) after Aza (**Supp.**
18 **Table 4**). Prior studies have suggested that Aza anti-tumor effect is largely mediated by reactivation
19 of endogenous retroviruses stimulating a tumor-autonomous interferon response^{45,46}. Consistent
20 with this work, we found a pronounced increase in interferon-responsive genes after Aza, but not
21 ATRA, including *IRF1*, *IFITM1*, *IFITM2*, and *IFITM3* (**Fig. 5B**). KEGG analysis also confirmed
22 this effect (**Fig. 5C**). Given evidence across multiple systems that interferon upregulates *CD38*
23 expression^{47,48}, our transcriptional profiling thus also supports an interferon-based mechanism
24 driving surface CD38 increase in plasma cells after Aza treatment.

1
2 *Plasma cell proliferative signaling pathways are inhibited by mAb binding to CD38*

3 In our final set of experiments related to targeting surface CD38, we were intrigued as to
4 whether binding of a therapeutic mAb leads to specific cellular phenotypes within myeloma plasma
5 cells. For example, isatuximab is known to directly lead to apoptosis of plasma cells⁴⁹, and
6 daratumumab can do so after crosslinking⁵⁰. However, the mechanism underlying this transduction
7 of extracellular mAb binding to intracellular phenotype remains unclear. In addition, our CRISPRi
8 screen data (**Fig. 1B** and **Supp. Fig. 1C**) suggests that surface CD38 expression may be strongly
9 impacted by intracellular phospho-signaling pathways.

10 Therefore, we used unbiased phosphoproteomics by mass spectrometry to probe intracellular
11 signaling effects driven by CD38 mAb binding. In RPMI-8226 cells we compared 20 μ M
12 daratumumab treatment vs. IgG1 isotype control. We chose a time point of 20 minutes of treatment
13 given known rapid alterations in signaling pathways in similar phosphoproteomic experiments⁵¹. In
14 total, across triplicate samples we quantified 5430 phosphopeptides (**Supp. Table 5**; **Supp. Fig.**
15 **5A**). Analyzing phosphopeptide changes by Kinase Substrate Enrichment Analysis (KSEA)⁵², we
16 were intrigued to find downregulation of phosphorylation motifs consistent with both cyclin-
17 dependent kinases as well as several kinases of the MAP kinase pathway (**Fig. 6A**). Downregulation
18 of phosphorylation on several central nodes in the MAP kinase as well as AKT pathway was also
19 apparent via KEGG analysis (**Supp. Fig. 5B**). Across a time course we further confirmed effects on
20 MAP kinase pathway (reported by phosphorylation of MAPK (ERK1/2), a key node in this
21 response) and AKT signaling after daratumumab treatment via Western blotting in RPMI-8226 and
22 MM.1S cell lines, respectively (**Fig. 6B**). While the absolute value of changes in MAPK signaling
23 are modest, both by phosphoproteomics and Western blot, these results indicate that daratumumab

1 binding to CD38 can at least partially inhibit this central proliferative pathway within myeloma
2 tumor cells, and thus may form a component of daratumumab's anti-tumor effect.

3

4 **Discussion**

5 Our studies here present a “multi-omics” view of therapeutically targeting CD38 in multiple
6 myeloma. Our integrated functional genomics and epigenetic analysis point to the central role of
7 transcriptional regulators in governing CD38 surface expression. Using surface proteomics, we
8 further identify that loss of CD38 in isolation is unlikely to drive large changes in the “surfaceome”,
9 while known pharmacologic strategies to increase CD38 have largely divergent impacts on other
10 surface antigens. Finally, unbiased phosphoproteomics reveals that binding of anti-CD38 mAb can
11 impair intracellular proliferative signaling within plasma cells.

12 Our CRISPRi screen illustrated the central role of numerous transcription factors, such as
13 SPI1, HEXIM1, and TLE3, in regulating surface CD38. This functional genomic study suggests that
14 regulation of surface CD38 largely occurs at the transcriptional, as opposed to protein trafficking,
15 level. This finding was in sharp contrast to our prior CRISPRi results with BCMA, where we found
16 that post-transcriptional mechanisms, such as proteolytic cleavage by γ -secretase and protein
17 trafficking via the SEC61 translocon, played the strongest roles in determining surface BCMA
18 levels²¹.

19 Another recent study used genome-wide CRISPR deletion screening to find genes that,
20 when knocked out, could abrogate IL-6-mediated downregulation of surface CD38 (ref.¹⁵). The
21 strongest hits in this prior study included the transcription factors *STAT1* and *STAT3*, demonstrating
22 a role for JAK-STAT signaling in regulating tumor CD38 expression within the bone marrow
23 microenvironment¹⁵. In support of this notion, our integrated epigenetic and machine learning
24 analyses, extracted from bone marrow-derived patient tumor samples, also support a critical role for

1 STAT1 in governing surface CD38. However, in our CRISPRi screen in an *in vitro* monoculture
2 system, neither *STAT1* or *STAT3* affected CD38 surface expression (**Fig. 1B**). This result suggests
3 that JAK-STAT signaling may not play a major role in CD38 regulation in the absence of
4 exogenous tumor stimulation. This finding illustrates the complementary nature of our genome-
5 wide screen to that previously published under the context of IL-6 stimulation¹⁵.

6 Toward the goal of finding key regulators of CD38 that were not previously known, our
7 epigenetic and machine learning approaches suggest that *XBP1* is a critical regulator of plasma cell
8 *CD38*. To our knowledge, there are not currently any known pharmacologic mechanisms to
9 potentiate *XBP1* activity. Given the important role of *XBP1* splicing in myeloma plasma cells⁵³,
10 future work will investigate the role spliced vs. unspliced *XBP1* in specifically regulating CD38, as
11 this strategy may provide new avenues for CD38 manipulation. Future work will also investigate
12 the role of *XBP1* deletion in determining clinical response to daratumumab⁴².

13 Given that plasma cells demonstrate frequent loss of CD38 after daratumumab treatment⁵, a
14 pressing question is whether CD38-low, daratumumab-resistant cells have novel
15 immunotherapeutic vulnerabilities. However, our recently-described strategy of “antigen escape
16 profiling”²⁷ – CRISPRi knockdown followed by unbiased cell surface proteomics (**Fig. 4**) –
17 suggests that other surface antigens on plasma cells do not exhibit consistent changes due to CD38
18 downregulation alone. This finding supports the notion that alterations in surface proteins found
19 after mAb treatment on patient tumors, such as increases in CD55 and CD59 (ref.⁵), are caused by
20 other therapy-induced selective pressure within the tumor microenvironment, not CD38 loss.

21 While *in vitro* assays have suggested a strong relationship between CD38 antigen density
22 and either NK-cell⁵⁴ or macrophage-mediated antibody-dependent cell killing⁶, these experiments
23 cannot readily take into account the critical role of the immune microenvironment in determining
24 daratumumab response or resistance⁸. Furthermore, even with the higher specificity of ATRA, there

1 is the potential to alter CD38 expression on other hematopoietic cells, which may impact clinical
2 responses to daratumumab¹⁷. Notably, current clinical data is most consistent with pre-treatment
3 tumor CD38 antigen density positively correlating with daratumumab depth of response^{5,11}.
4 Analysis of transcriptional data in CoMMpass demonstrates that while increased *CD38* expression
5 was associated with improved outcomes in all patients, only those treated with daratumumab
6 showed a divergence in survival curves coinciding with the median time of initial daratumumab
7 treatment (**Supp. Fig. 6**). Pharmacologic manipulation of CD38 density on tumor cells may
8 ultimately be most fruitful pre-treatment rather than in the context of daratumumab resistance.
9 Similar strategies may also be most beneficial for other CD38-targeting immunotherapeutics².

10 Also directly related to mAb therapeutic effects, our unbiased phosphoproteomic results
11 suggest that daratumumab binding to CD38 can directly decrease signaling along the MAP kinase
12 and PI3K-AKT pathways. It remains to be investigated whether this inhibition of central
13 proliferative signaling pathways plays a role in the anti-tumor effect of daratumumab in patients.

14 In terms of limitations of our work, the most prominent is that the many of our studies are
15 derived from large-scale “omics” experiments in myeloma cell lines. There may be biological
16 differences between our findings *in vitro* and primary tumors growing within the bone marrow
17 microenvironment.

18 Taken together, our multi-omic studies comprise a resource that reveals new insight into the
19 genetic, epigenetic, and pharmacologic regulation of surface CD38 in myeloma plasma cells. We
20 anticipate these findings will have utility in deriving new strategies to enhance CD38-targeting
21 therapies in myeloma, including mAbs in current clinical practice as well as emerging antibody and
22 cellular therapies². The technologies described here also comprise a blueprint to comprehensively
23 assess determinants of surface antigen regulation, and impacts of associated therapeutic
24 manipulation, that could be applied across targets in hematologic malignancies.

1

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10 performed experiments. P.C., B.P.E., N.P., Y-H.T.L., B.J.B., and P.R. analyzed data. D.W., P.P., V.
11 Steri, and B.H. performed murine studies. H.G. analyzed patient epigenetic data. V. Sarin
12 performed machine learning analysis. P.C. and A.P.W. wrote the manuscript with input from all
13 authors.

14

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16 time of completing this project she was fully employed by the University of California, San
17 Francisco. P.R. is currently an employee and shareholder of Senti Biosciences, though during the
18 time of completing this project she was fully employed by the University of California, San
19 Francisco. A.P.W. is an equity holder and scientific advisory board member of Indapta
20 Therapeutics, LLC and Protocol Intelligence, LLC. M.K. has filed a patent application related to
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24

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28
29

1 **FIGURE LEGENDS**

2

3 **Figure 1. CRISPR interference (CRISPRi) screening reveals genetic determinants of surface**
4 **CD38 regulation.** **A.** Schematic of CRISPRi screen design. **B.** Results of CRISPRi screen
5 demonstrating genes that, when knocked down, regulate surface CD38 in RPMI-8226 cells. X-axis
6 indicates phenotype (epsilon) from MAGeCK⁵⁵ statistical analysis. Dashed line indicates cutoff for
7 significant change at False Discovery Rate (FDR) < 0.05. Genes of interest are specifically labeled.
8 4,000 negative control non-targeting sgRNAs are in grey. **C.** Gene Ontology (GO) Biological
9 Process and Kyoto Encyclopedia of Genes and Genomes (KEGG) analysis of all genes that when
10 knocked down lead to significant CD38 upregulation. **D.** Follow-up flow cytometry validation of
11 CRISPRi screen hits using two individual sgRNAs per gene demonstrates *TLE3* knockdown drives
12 increased CD38, while *SPI1* knockdown leads to CD38 decrease.

13

14 **Figure 2. Validation of CRISPRi screen hits as functionally impacting daratumumab efficacy.**

15 **A.** Knockdown of HEXIM1 and TLE3 with two independent sgRNA's/gene (AMO1 myeloma
16 cells, $n = 3$) followed by flow cytometry shows significant surface CD38 increase with TLE3_i2
17 sgRNA and trend toward increased CD38 with HEXIM1_i1 sgRNA. Scri = non-targeting control
18 sgRNA. **B.** Results from ADCC assays with AMO1 cells stably expressing the noted sgRNA's and
19 incubated with the indicated concentration of daratumumab or isotype control antibody (1:20
20 myeloma:NK ratio, 20 hours, $n = 2$). The percent lysis by ADCC was calculated using the following
21 formula : % Lysis = (signal in presence of daratumumab – signal in presence of IgG1 control
22 antibody) x100 / signal in presence of IgG1 control antibody. At 10 μ M daratumumab, both
23 HEXIM1 and TLE3 knockdown led to significant increase in ADCC. **C.** Similar to A., sgRNA
24 knockdown of *NFKB1*, *NFKB2*, and *SPI1* with fold-change in CD38 by flow cytometry (RPMI-

1 8226 cells, $n = 3$). **D.** Similar to B., knockdown with the most effective sgRNA for each gene show
2 significant decreases in NK-cell ADCC at 10 μ M daratumumab in the RMPI-8226 cells ($n = 3$). **E.**
3 *In vivo* validation of *SPII* knockdown driving daratumumab resistance. NOD *scid* gamma mice
4 were I.V. implanted with CRISPRi RPMI-8226 cells stably expressing both luciferase and noted
5 sgRNA, then treated with 200 μ g daratumumab on the noted schedule. Bioluminescence imaging
6 measurement of tumor burden demonstrates significantly increased fold-change in tumor burden
7 (normalized to pre-daratumumab intensity) with either CD38 or SPI1 knockdown compared to
8 scramble sgRNA. For **A-E**, $*p < 0.05$, $**p < 0.01$ by two-tailed *t*-test.

9

10 **Figure 3. Patient-centered epigenetic analysis and machine learning predicts most potent**
11 **transcriptional regulators of CD38. A.** 46 transcription factors predicted to bind to the CD38
12 locus were derived from motif analysis of published ATAC-seq data (see Supplementary Figure 3).
13 Gene expression of each transcription factor (TF) was correlated with *CD38* expression in the
14 Multiple Myeloma Research Foundation (MMRF) CoMMpass database (release IA13), with RNA-
15 seq data from CD138+ enriched tumor cells at diagnosis ($n = 664$ patients). Top predicted positive
16 and negative regulators are shown based on Pearson correlation (*R*). **B.** CoMMpass RNA-seq data
17 illustrates strong positive correlation between *XBP1* and *CD38* expression. **C.** XGBoost machine
18 learning model was used to extract features of transcription factor gene expression that best-model
19 *CD38* expression in CoMMpass tumors (shown in log2 TPM (Transcripts per Million)). 80% of
20 data was used as a test set with 20% left out as a training set. Coefficient of variation (R^2) for
21 predictive model = 0.49 after five-fold cross validation. **D.** Shapley Additive Explanations (SHAP)
22 analysis indicates transcription factors whose expression most strongly impacts *CD38* expression
23 levels in CoMMpass tumors.

24

1 **Figure 4. Minimal alterations of the myeloma cell surface proteome after CD38 loss. A.**
2 Schematic of “antigen escape profiling” approach to reveal new cell-surface therapeutic
3 vulnerabilities in the context of CD38 downregulation. **B.** Cell surface capture proteomics
4 comparing *CD38* knockdown vs. non-targeting sgRNA control, with aggregated data across three
5 cell lines (CRISPRi-expressing RPMI-8226, AMO1, and KMS12-PE; $n = 3$ replicates per cell line
6 per sgRNA) reveals minimal changes in the cell surface proteome beyond *CD38* knockdown at
7 significance cutoff of $p < 0.05$ and \log_2 fold-change $>|1.5|$. **C.** Integrated analysis of cell surface
8 proteomics and mRNA-seq ($n = 2$ per cell line per guide) across three cell lines reveals the only
9 consistent change at both protein and transcript level after *CD38* knockdown is *THY1/CD90*
10 upregulation. \log_2 fold-change cutoff = $|1.5|$.

11

12 **Figure 5. ATRA drives CD38 upregulation with limited additional cellular impact while Aza**
13 **leads to a broad interferon-mediated response. A.** Integrated mRNA-seq ($n = 2$ per drug
14 treatment) and cell surface proteomics ($n = 2$ per drug treatment) across RPMI-8226 treatment with
15 10 nM all-trans retinoic acid (ATRA), 2 μ M azacytidine (Aza), and 10 nM panobinostat (Pano).
16 All plots are in comparison to control replicates treated with 0.1% DMSO. Doses chosen are based
17 on those previously published to lead to CD38 upregulation for each agent. Data points shown are
18 for proteins and genes corresponding to Uniprot-annotated membrane-spanning proteins. \log_2 -fold
19 change cutoffs shown at $|0.5|$ for ATRA and $|2.0|$ for Aza and Pano to increase clarity of plots given
20 many fewer changed genes with ATRA treatment. **B.** RNA-seq for same samples with ATRA or
21 Aza treatment vs. DMSO but here showing all mapped genes, not just those annotated as
22 membrane-spanning. Significance cutoff at $p < 0.05$ with \log_2 fold-change cutoff set at $|0.8|$ to
23 illustrate prominent differences above this level in transcriptome alteration after either ATRA or

1 Aza treatment. **C.** KEGG analysis of genes from RNA-seq dataset meeting cutoff criteria of $p <$
2 0.05 and \log_2 fold-change >0.8 after Aza treatment.

3

4 **Figure 6. Unbiased phosphoproteomics reveals downregulation of proliferative signaling after**
5 **daratumumab treatment. A.** RPMI-8226 cells were treated with 20 μ M daratumumab or IgG1
6 isotype control for 20 minutes ($n = 3$ each) and then harvested for unbiased phosphoproteomics with
7 immobilized metal affinity chromatography enrichment for phosphopeptide enrichment. Plot
8 displays results of Kinase Substrate Enrichment Analysis, indicating modest decrease in
9 phosphorylation of numerous predicted substrates of MAPK pathway kinases as well as cyclin-
10 dependent kinases (cutoffs of $p < 0.05$, \log_2 fold-change $>|0.5|$). **B.** Western blot in RPMI-8226 of
11 MAPK (ERK1/2) (Thr202/Tyr204) relative to total MAPK demonstrates modest decrease in MAPK
12 phosphorylation after 5, 10, or 15 min daratumumab (Dara) treatment; magnitude of change
13 normalized to IgG1 control at each time point (red) appears consistent with phosphoproteomic data.
14 **C.** Western blot of MM.1S cells treated with daratumumab and blotted for p-AKT (Ser473) and
15 total AKT, with quantification of p-AKT relative to total AKT and normalized to IgG1 at each time
16 point. All images representative of two independent Western blots.

FIGURE 1

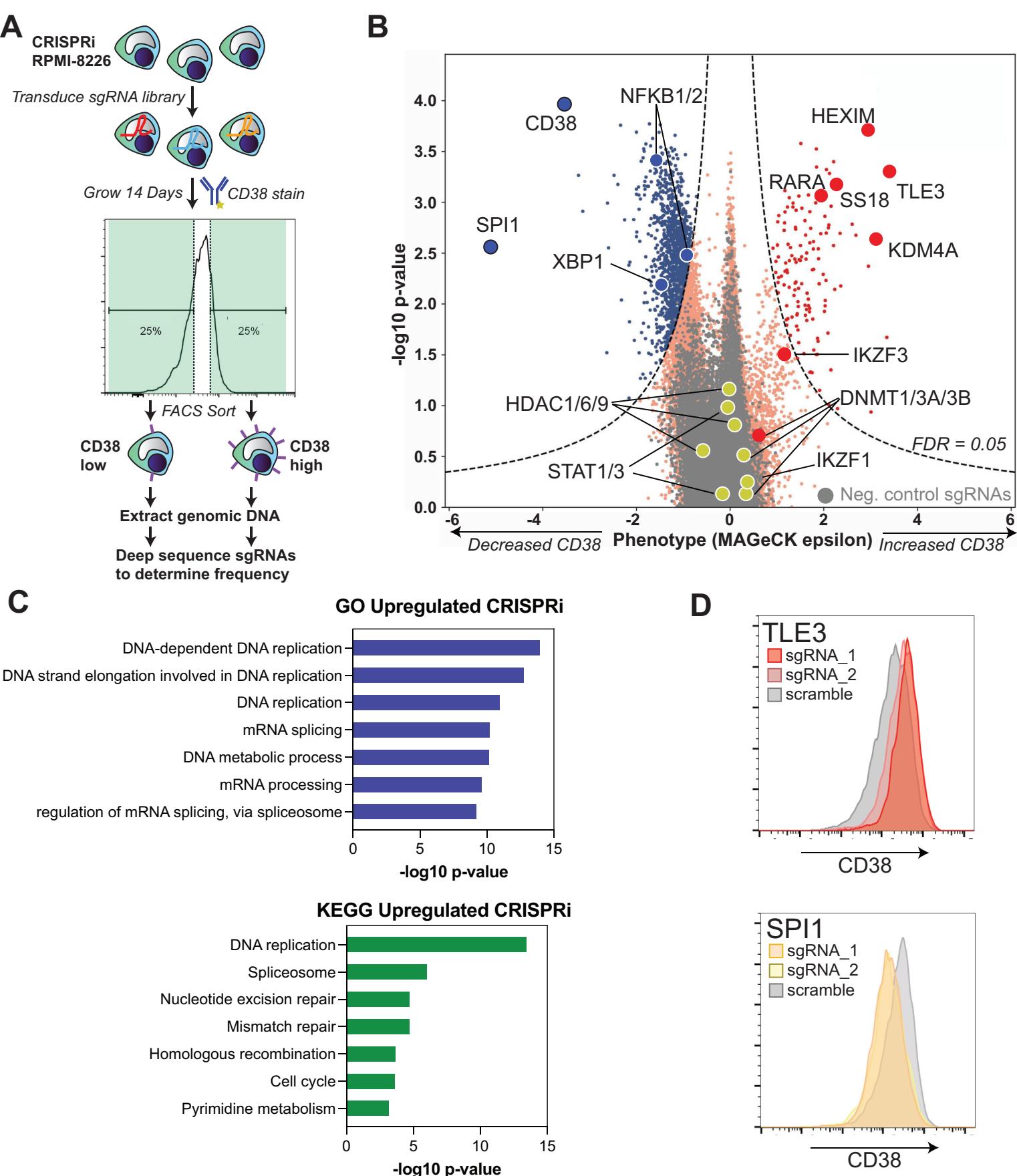
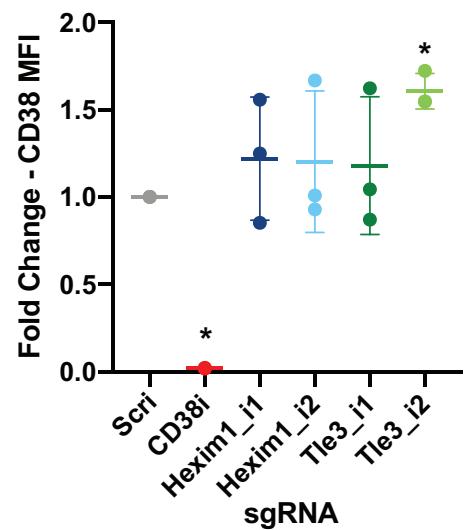
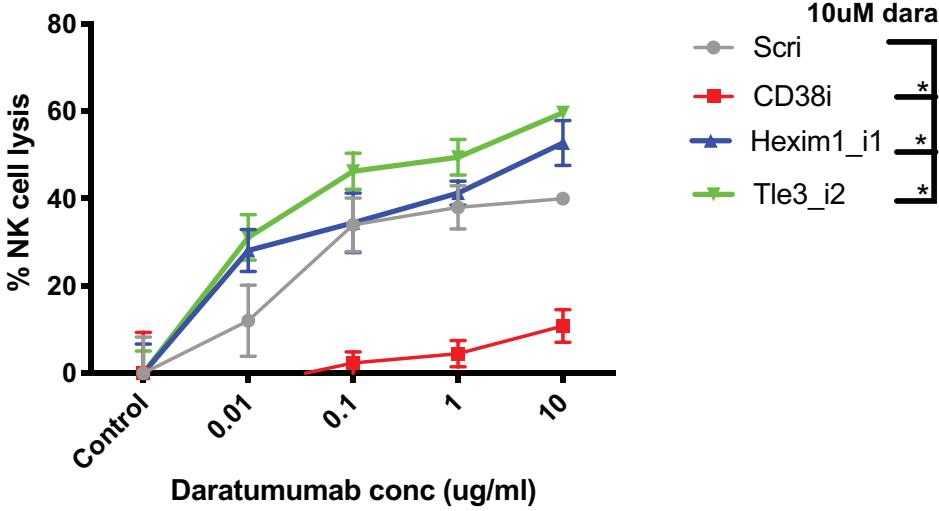


FIGURE 2

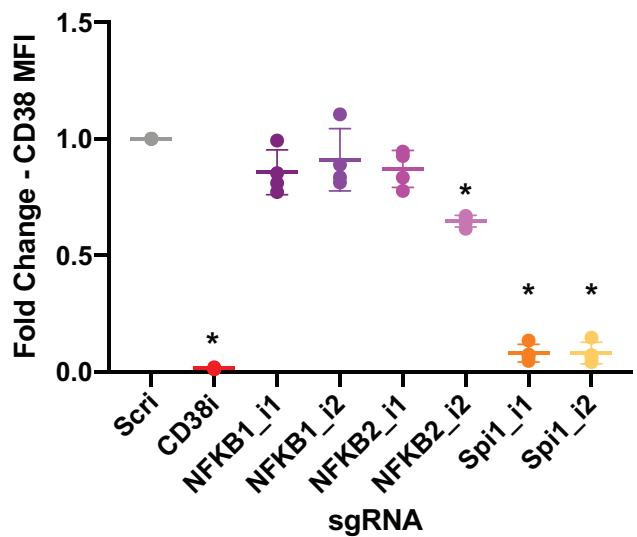
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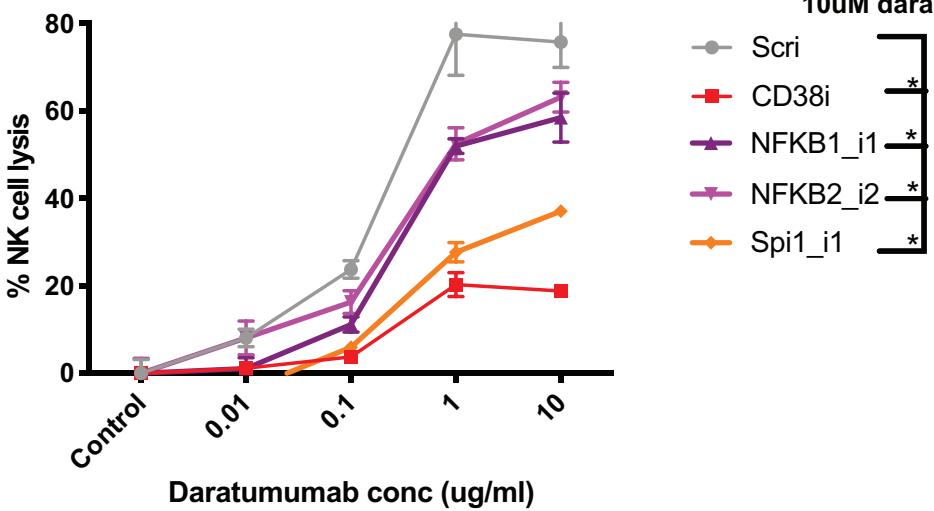
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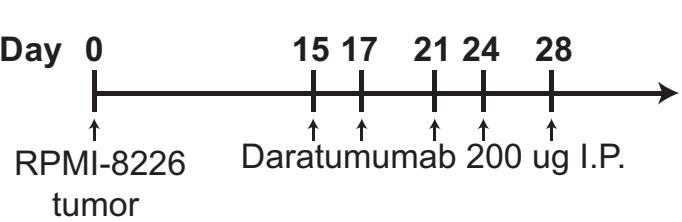
C



D



E



8 NSG x scr-i (ctl)
8 NSG x CD38i
8 NSG x SPI1i

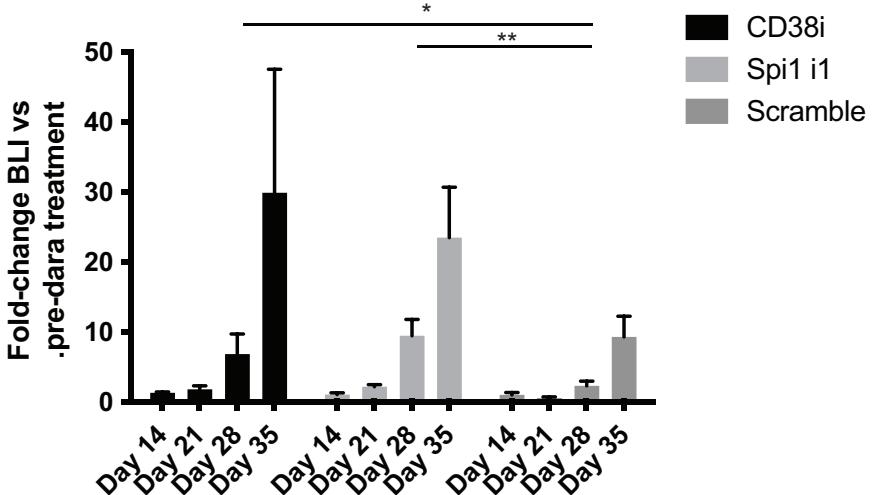
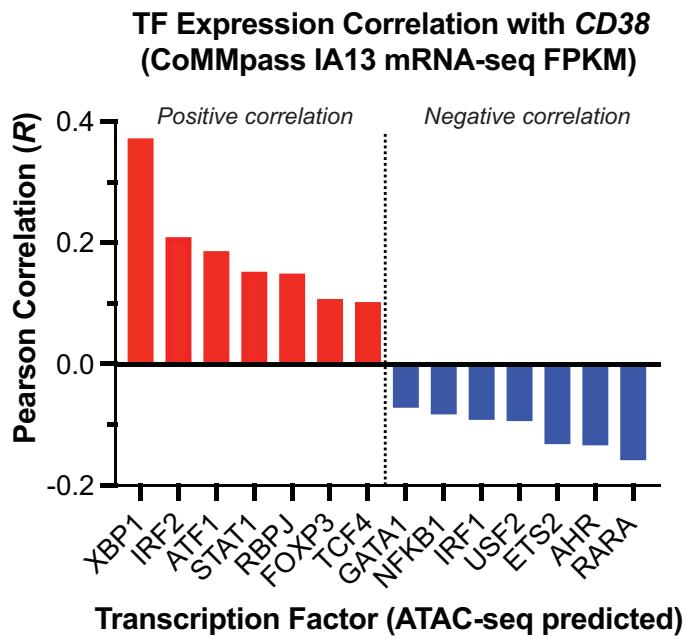
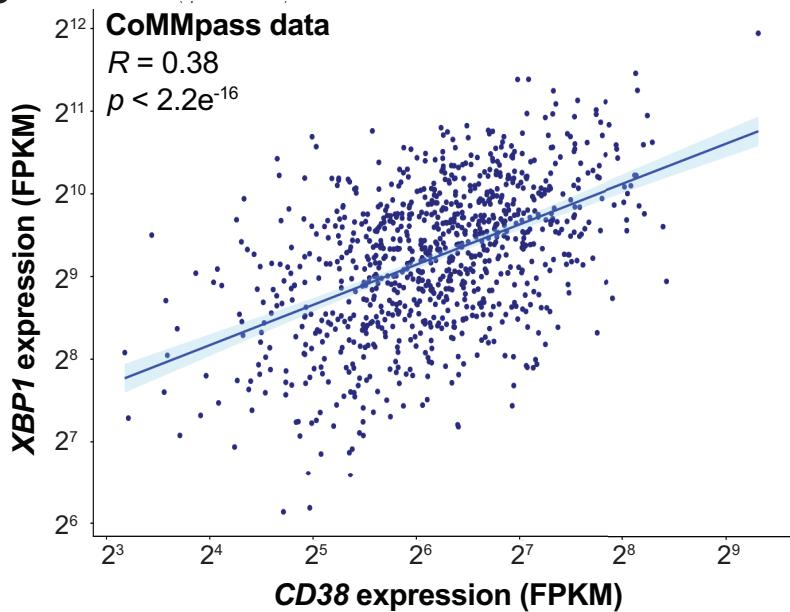


FIGURE 3

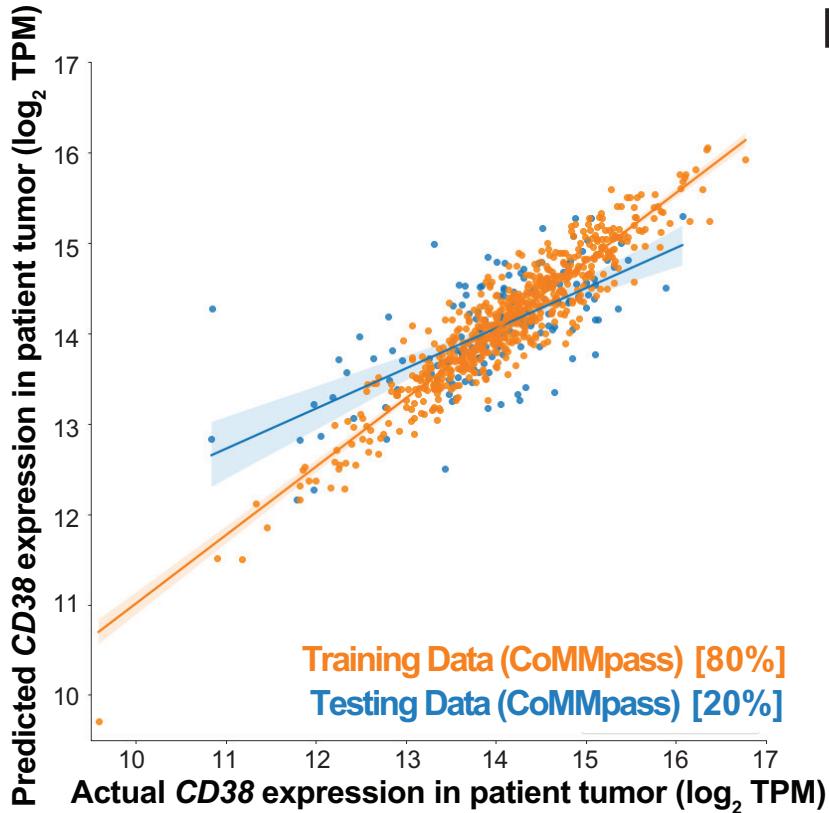
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B



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D

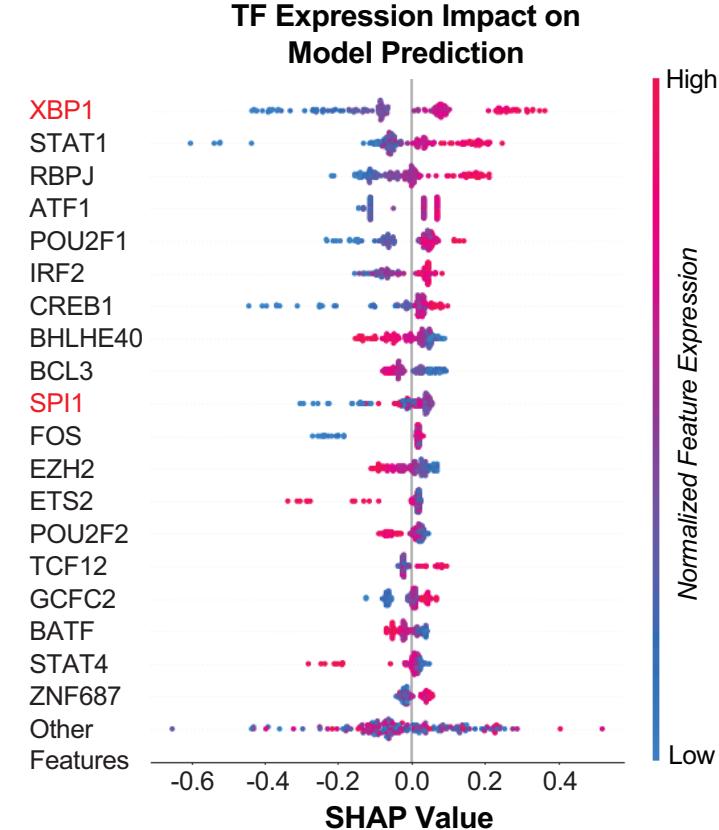
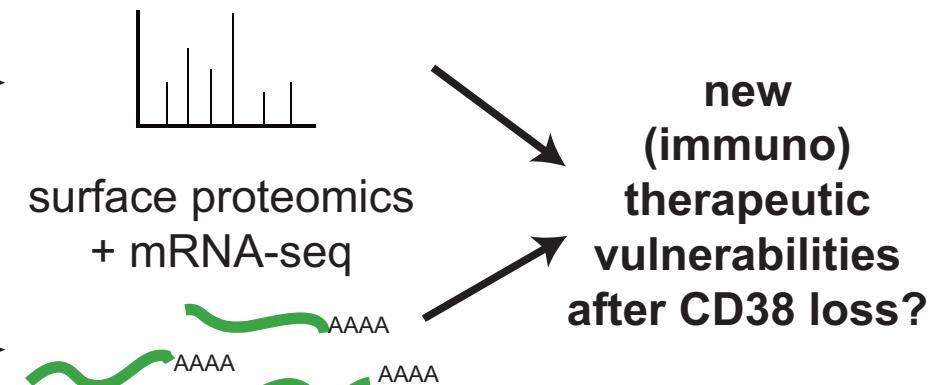
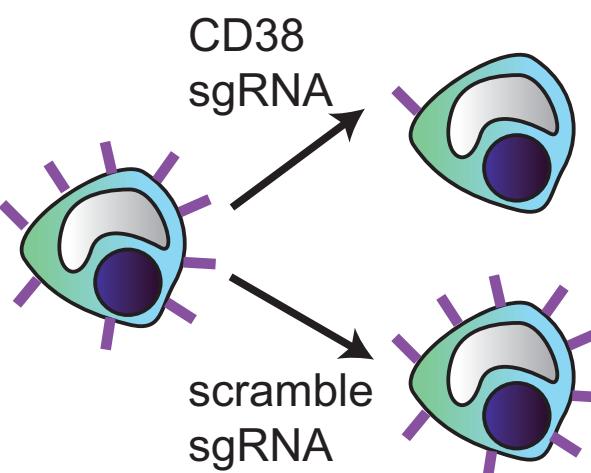


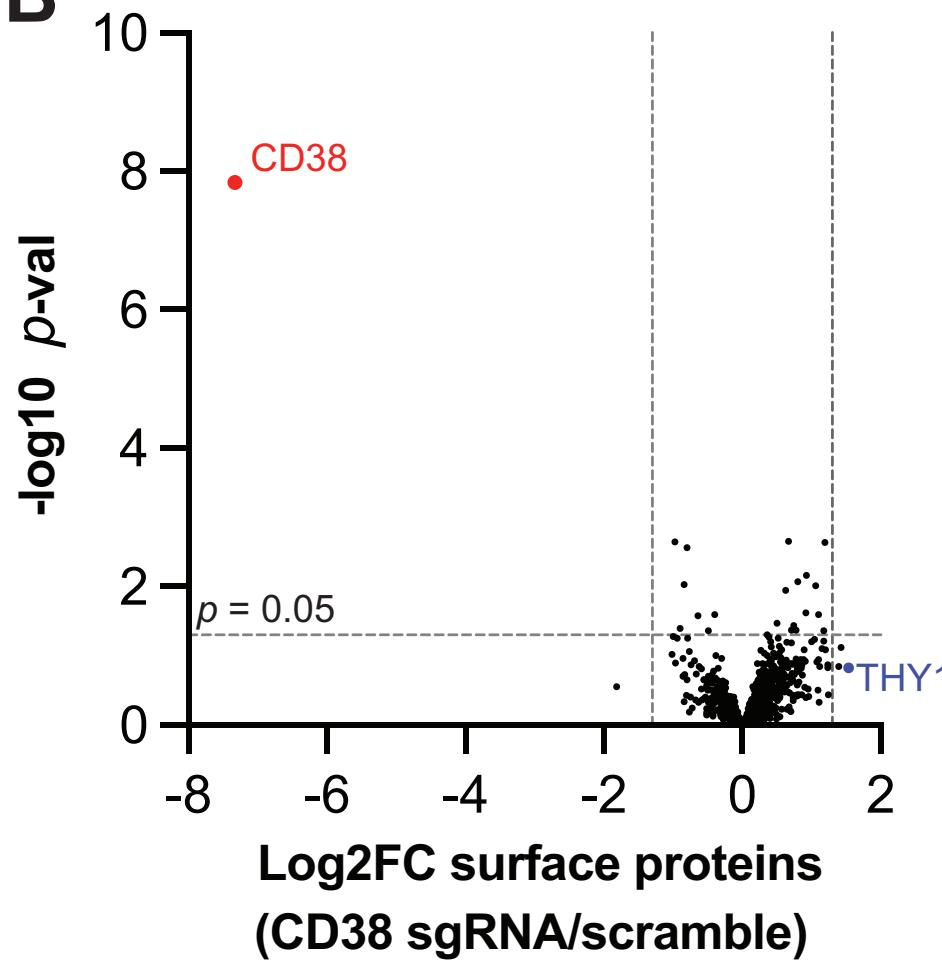
FIGURE 4

A

CRISPRi:
RPMI-8226
AMO1
KMS12-PE



B



C

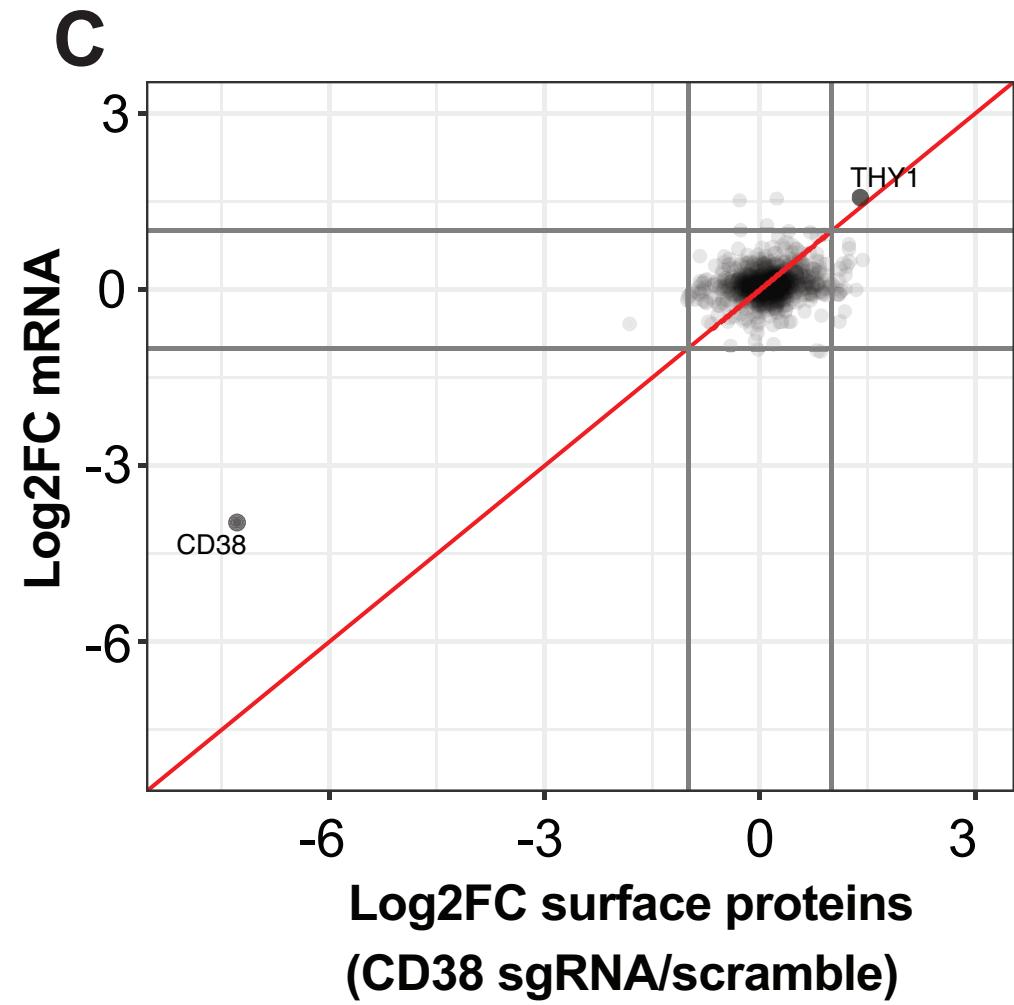
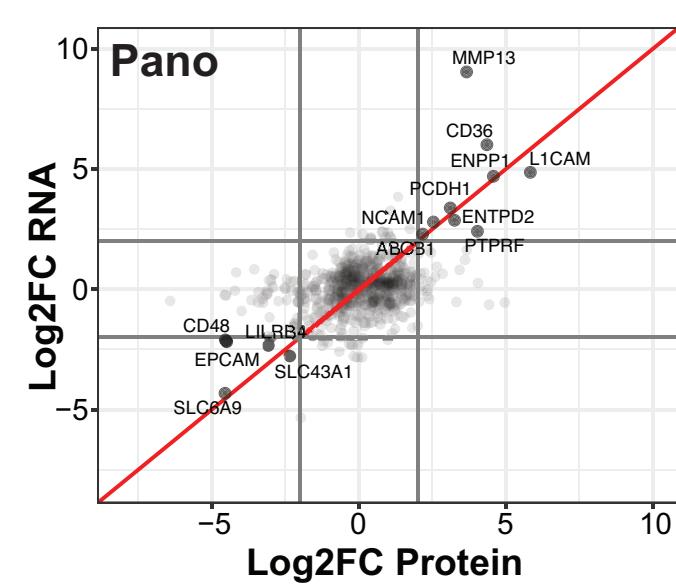
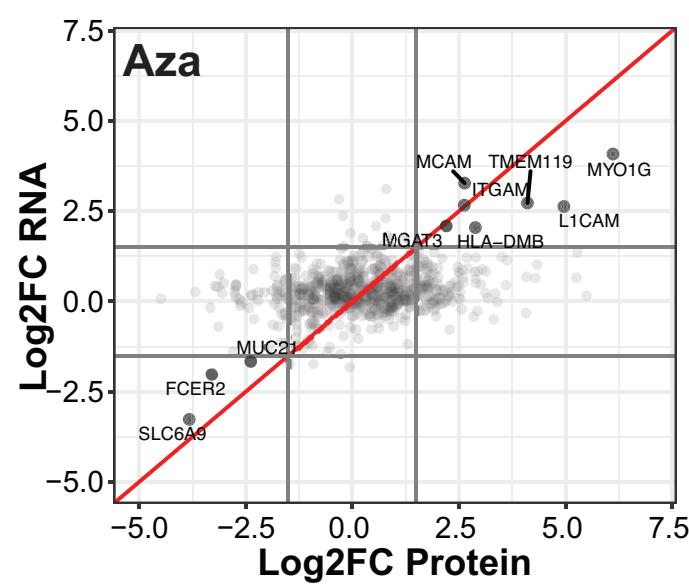
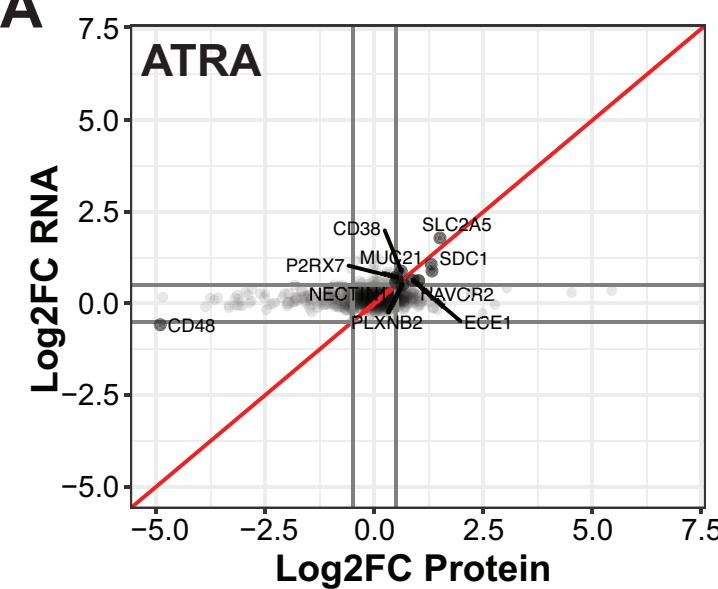
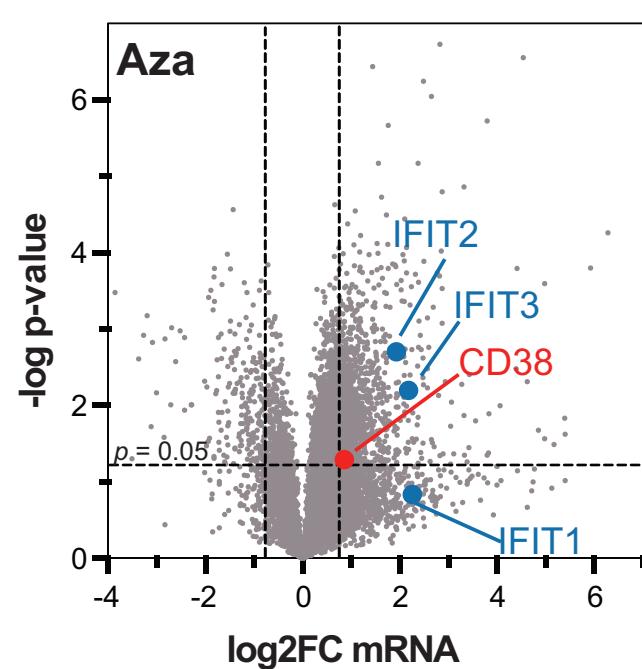
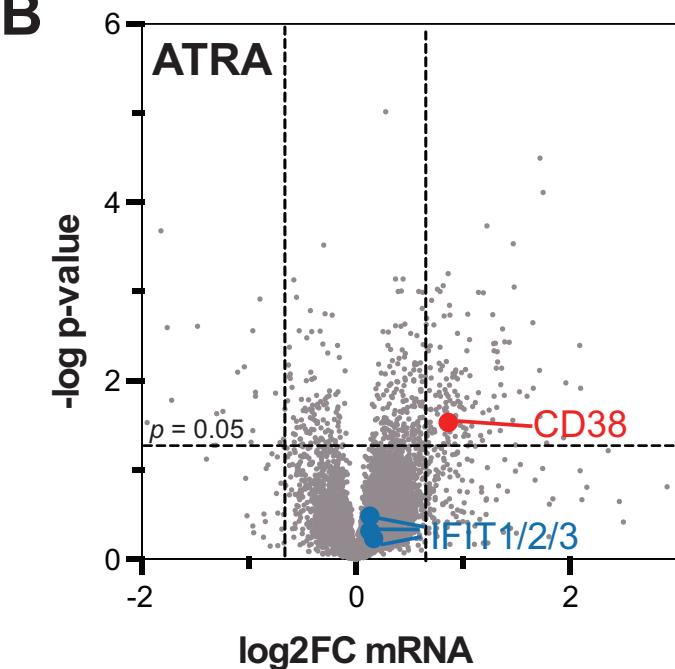


FIGURE 5

A



B



C

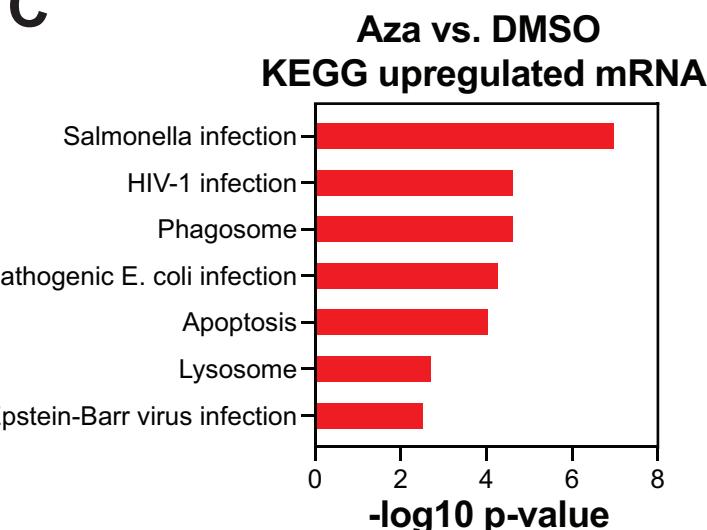
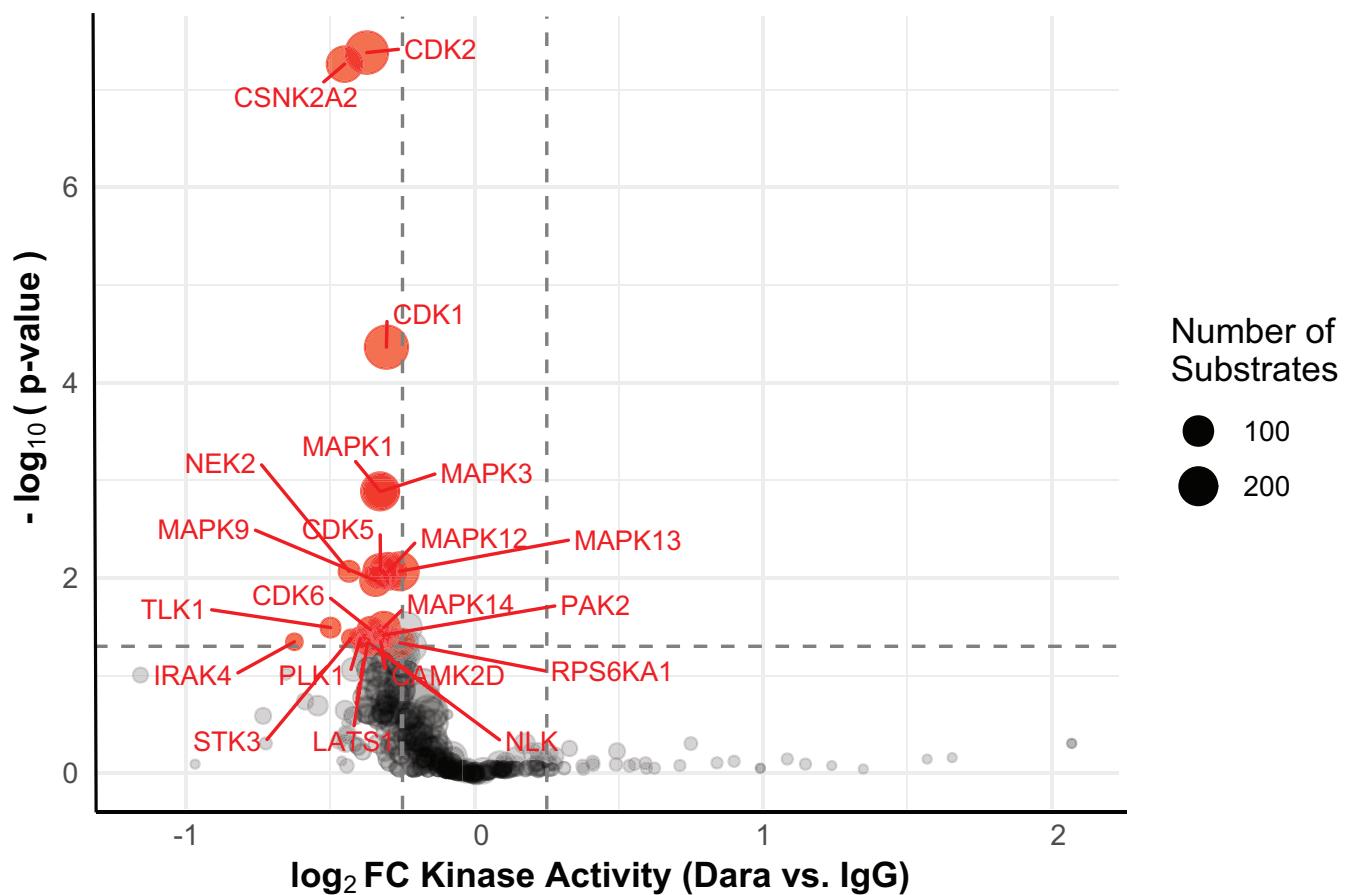


FIGURE 6

A



B

