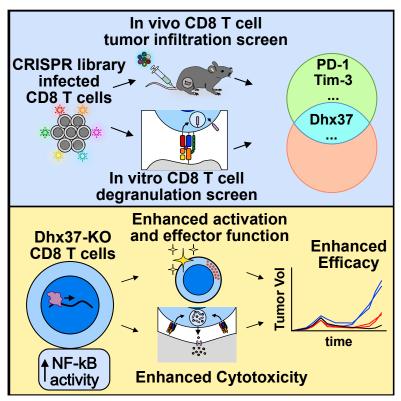
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Systematic Immunotherapy Target Discovery Using Genome-Scale *In Vivo* CRISPR Screens in CD8 T Cells

Graphical Abstract



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In Brief

Genome-wide CRISPR screening in CD8 T cells in the context of immunotherapy identifies genes that modulate T cell effector functions, including *Dhx37*, an RNA helicase that affects NF-κB signaling, T cell activation, and cytotoxicity.

Highlights

- In vivo and in vitro genome-scale CD8 T cell CRISPR screen in immunotherapy contexts
- Dhx37 knockout in CD8 T cells enhances adoptive transfer efficacy
- Dhx37 modulates CD8 T cell activation, cytokine production, and cytotoxicity
- DHX37 interacts with PDCD11 and influences NF-κB activity





Systematic Immunotherapy Target Discovery Using Genome-Scale In Vivo CRISPR Screens in CD8 T Cells

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SUMMARY

CD8 T cells play essential roles in anti-tumor immune responses. Here, we performed genomescale CRISPR screens in CD8 T cells directly under cancer immunotherapy settings and identified regulators of tumor infiltration and degranulation. The in vivo screen robustly re-identified canonical immunotherapy targets such as PD-1 and Tim-3, along with genes that have not been characterized in T cells. The infiltration and degranulation screens converged on an RNA helicase Dhx37. Dhx37 knockout enhanced the efficacy of antigen-specific CD8 T cells against triple-negative breast cancer in vivo. Immunological characterization in mouse and human CD8 T cells revealed that DHX37 suppresses effector functions, cytokine production, and T cell activation. Transcriptomic profiling and biochemical interrogation revealed a role for DHX37 in modulating NF-κB. These data demonstrate high-throughput in vivo genetic screens for immunotherapy target discovery and

establishes DHX37 as a functional regulator of CD8 T cells.

INTRODUCTION

T cells play a central role in maintaining the cellular integrity of the body against intracellular pathogens and tumors (Kaech et al., 2002; Tscharke et al., 2015; Zhang and Bevan, 2011). Activation of antigen-specific CD8 T cells is mediated by T cell receptor (TCR) recognition of cognate antigen on surface major histocompatibility complex (MHC) class I (MHC-I) (Blum et al., 2013; Cox et al., 2011; Tscharke et al., 2015), resulting in T cell proliferation, cytokine production, and killing of target cells (Wong and Pamer, 2003). Defects in T cells can lead to recurrent infections (Dumontet et al., 2015; Hersperger et al., 2011; Kurktschiev et al., 2014; Zhang et al., 2009) or cancer (Chia et al., 2009; Clementi et al., 2005; Mortaz et al., 2016), while dysregulated activation of CD8+ T cells can result in immunopathology and autoimmunity (Gravano and Hoyer, 2013; Holzelova et al., 2004; Valori et al., 2017; Walter and Santamaria, 2005). CD8 T cells have become the central focus of new cancer therapeutics (Blankenstein et al., 2012; Chen and Mellman, 2013; Fridman et al., 2012;



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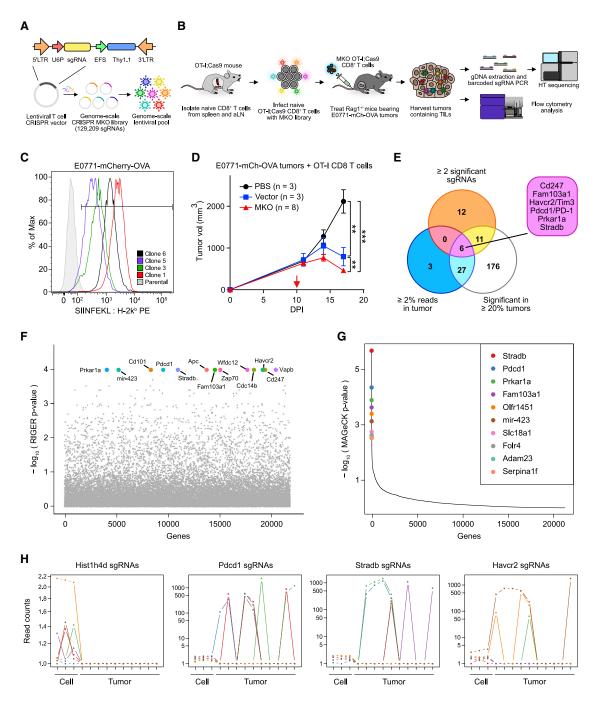


Figure 1. Genome-Scale In Vivo CD8 T Cell Screen Identified Genes Regulating Tumor Infiltration

- (A) Schematics of the design of a T cell CRISPR vector, MKO library cloning, and lentiviral library production.
- (B) Schematics of experiment design.
- (C) Measurement of antigen presentation in E0771-mCh-OVA clonal cell lines with a SIINFEKL:H-2k^b antibody.
- (D) Growth curve of intramammary fat pad tumors from transplanted E0771-mCh-OVA cells in $Rag1^{-/-}$ mice following different treatments. PBS control (n = 3), adoptive transfer of OT-I;Cas9 CD8 T cells infected with vector (n = 3) or MKO (n = 8). Red arrow indicates the time of adoptive transfer. Data are shown as mean \pm SEM. Note that some error bars are not visible because the values are small. *p < 0.05, **p < 0.01, ***p < 0.001 by two-way ANOVA.
- (E) Venn diagram of the three enrichment criteria to identify the top gene hits (\geq 2% read abundance in one sample [n = 36], significant in \geq 20% of samples [n = 220], and \geq 2 independent enriched sgRNAs [n = 26]).
- (F) Meta-analysis of infiltration screen using RIGER with the second-best sgRNA method.

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Spranger et al., 2013). Approved immune checkpoint inhibitors enhance anti-tumor response of CD8 T cells by neutralizing CTLA-4 or PD-1/PD-L1 (Brahmer et al., 2010; Kvistborg et al., 2014; Pardoll, 2012; Ribas, 2012; Swart et al., 2016; Tumeh et al., 2014). These drugs have been effective in multiple oncology indications (Mellman et al., 2011; Sharon et al., 2014), either as monotherapies or in combination with other therapies (Larkin et al., 2015; Postow et al., 2015). Newer checkpoint inhibitors are now under active investigation, such as TIM-3, LAG3, 4-1BB, CD28, ICOS, OX-40, and VISTA (Johnson and June, 2017; Mahoney et al., 2015). Meanwhile, cellular therapeutics utilizing CD8 T cells with chimeric antigen receptors (CARs) (Fesnak et al., 2016; Srivastava and Riddell, 2015) have demonstrated clinical success against hematopoietic malignancies (Esensten et al., 2016; Jackson et al., 2016; Johnson and June, 2017) and have been recently approved. Discovery of previously unknown genes that modulate T cell function is of urgent need to open different avenues for immunotherapies (Mellman et al., 2011), as a large fraction of patients still do not respond to, or have undesired side effects to currently approved therapies (Kelderman et al., 2014; Restifo et al., 2016). Systematic approaches to identify new regulators of T cell functions in vivo can provide potentially orthogonal or complementary opportunities (Ghoneim et al., 2016).

Studies using RNAi or shRNA (short hairpin RNA) libraries have been used to identify CD8 T cell regulators (Chen et al., 2014; Zhou et al., 2014). RNAi operate by degrading mRNA or suppressing their translation but have substantial off-target effects (Nijman, 2015). CRISPR technologies have dramatically enhanced our ability in genome editing (Cong et al., 2013; Hsu et al., 2014; Jinek et al., 2012; Komor et al., 2017; Mali et al., 2013; Wright et al., 2016). High-throughput CRISPR screens have been developed for discovery of novel genes (Shalem et al., 2014; Wang et al., 2014; Shalem et al., 2015). Application of CRISPR technologies has enabled the manipulation of the T cell genome (Beil-Wagner et al., 2016; Singer et al., 2016; Su et al., 2016). Genetic screens have been performed in vitro for discovery of regulators for T cell proliferation and differentiation (Henriksson et al., 2019; Shifrut et al., 2018). Given the complexity of physiological cues, it is critical to perform in vivo screens to better identify clinically relevant targets. High-throughput in vivo CRISPR screens in primary CD8 T cells would enable rapid identification of negative regulators of T cell anti-tumor function, thereby providing potential targets for immunotherapy.

In this study, we performed in vivo genome-scale CRISPR screens in CD8 T cells for tumor infiltration in mouse models of triple-negative breast cancer (TNBC). Our screen re-discovered prime immunotherapy targets such as PD-1 and Tim-3, as well as previously undocumented targets. We also performed an orthogonal screen to identify factors that modulate CD8 T cell degranulation ability. The screens converged on a small number of hits that included Dhx37, a previously uncharacterized target in T cells. We subsequently interrogated the function of Dhx37/ DHX37 in both mouse and human CD8 T cells. Our data revealed that Dhx37 suppresses CD8 T cell activity in response to antigenic stimulation. Biochemical experiments along with transcriptome profiling revealed that DHX37 interacts with both nuclear factor κB (NF- κB) p65 and PDCD11 and modulates NF-κB pathway activity in CD8 T cells.

RESULTS

Development of Genetic Screens of In Vivo Tumor Infiltration and Cytotoxic Function in Primary CD8 T Cells

We designed and generated a lentiviral CRISPR vector that contains an sgRNA expression cassette and Thy1.1 (STAR Methods) (Figure 1A). Next, we cloned into this vector a mouse genomescale single guide RNA (sgRNA) library (MKO) containing 128,209 gene-specific sgRNAs and 1,000 non-targeting controls (NTCs) (Table S1). We generated lentivirus from this library (Figures 1A and 1B) (STAR Methods), which enabled T cell CRISPR screens.

To set up a model system for T cell immunotherapy screens, we utilized E0771 cells, a murine TNBC line, as a transplant tumor model (Figure 1B). To enable T cell recognition of cognate antigen in cancer cells, we generated clonal E0771 cell lines that constitutively express chicken ovalbumin (OVA) (STAR Methods) (Figure 1C). Using an antibody that recognizes the SIINFEKL: H-2Kb complex (Porgador et al., 1997), we confirmed that the SIINFEKL epitope of OVA was presented on surface H-2Kb (Figure 1C). Clone 3 of E0771-mCherry-OVA (E0771mCh-OVA) cell lines was chosen for further studies because it presented a low-level, single-peak SIINFEKL: H-2Kb (Figure 1C). We then isolated naive CD8+ T cells from OT-I;Cas9 mice, cultured for 3 days, and then adoptively transferred these cells into Rag1-deficient mice (Figure S1A). Prior to adoptive transfer, these cells were analyzed by flow cytometry and found to have upregulated the T cell activation markers, CD44 and CD69 (Figures S1B and S1C). 7 days post-transfer, spleens were isolated and CD8⁺ T cells were purified and co-cultured with either E0771 or E0771-mCh-OVA cells in the presence of CD107a antibody, a granule-specific marker that is transiently deposited on the surface of T cells in response to antigenic stimulation (Betts et al., 2003), to quantitatively measure the degree of T cell degranulation (degranulation assay) (STAR Methods) (Figure S1D). We found that the transferred OT-I;Cas9 CD8 T cells degranulated in the presence of OVA, demonstrating that the transferred T cells retained antigenic specificity (Figure S1E). To determine whether the observed anti-tumor responses were OVA specific, we performed an in vivo experiment using E0771 cells or E0771mCh-OVA cells transplanted into the mammary fat pad of Rag1^{-/-} female mice. After 10 days, these mice were treated with either PBS or OT-I; Cas9 CD8 T cells (Figure S1F). We found that tumor regression occurred only in mice bearing E0771mCh-OVA tumors treated with OT-I;Cas9 CD8+ T cells but not

⁽G) Meta-analysis of infiltration screen using MAGeCK analysis of survival screen.

⁽H) Single gene-level analysis of individual for their sgRNA distributions across each sample, showing representative genes depleted (His1h4d) or enriched (Pdcd1, Stradb, and Havcr2) in tumors as compared to cells.

in the groups treated with either PBS or in mice carrying E0771 tumors (Figure S1G). Together, these data confirmed the antigenic specificity of this model system, providing a normalized baseline for *in vivo* genetic screens in primary CD8 T cells.

We then performed the in vivo T cell tumor infiltration screen (Figure 1B), using a mammary fat pad orthotopic tumor model (Figure 1D), as well as a subcutaneous model in parallel (Figures S2A and S2B) (STAR Methods). We transduced CD8+ T cells from OT-I;Cas9 mice with the MKO sgRNA library and adoptively transferred them into E0771-OVA tumor-bearing mice bearing tumors (STAR Methods). In both orthotopic and subcutaneous models, we observed that T cell injections (either vector or MKO transduced) mitigated tumor growth and that the MKO transduced population had a stronger therapeutic effect compared to vector control (Figure 1D). Histological and pathological analysis revealed the presence of tumor-infiltrating lymphocytes (TILs) from mice receiving adoptive transfer but not in PBS-treated mice (Figure S2B). Flow cytometric analysis of single-cell suspensions of spleens, lymph nodes, and tumors also detected adoptively transferred CD8⁺ T cells (Figure S2C). Tumors from a cohort of mice were subjected to high-throughput sgRNA library sequencing (STAR Methods), which revealed the sgRNA representations of MKO transduced OT-I;Cas9 CD8 T cells before injection and in all tumor samples (Table S1; Figures S2D and S2E). While the library representation of preinjected T cells follows a log-normal distribution for both genetargeting sgRNAs and NTCs, the sgRNA representation in organs is characterized by the dominance of a small fraction of sgRNAs (Figure S2D). At the individual mouse level, we found that the sgRNA representation between different mice are correlated with each other (Figure S2E).

Convergent T Cell Screens Identified Known and Undocumented Immunotherapy Targets

Using stringent criteria (false discovery rate [FDR] <0.5%), we identified significantly enriched sgRNAs in each tumor (STAR Methods). A composite of multiple ranking approaches, including sgRNA abundance, multiple independent sgRNAs and multiple mouse replicates (STAR Methods), revealed rank lists of top scoring sgRNAs and genes (Figure 1E; Figures S3A ad S3B). The infiltration screen revealed high-rank genes representing three classes of genes: immune regulators, growth controlling genes, and largely uncharacterized genes (Figures 1E-1G; Figures S3A and S3B). A total of 6 genes were significantly enriched with all three criteria fulfilled (Cd247, Fam103a1, Hacvr2/Tim-3, Pdcd1/PD-1, Prkar1a, and Stradb) (Figure 1E). To further assess the robustness of screen performance, we first applied RIGER (STAR Methods), an analysis method considering the performance of multiple sgRNAs for the same gene. RIGER analysis with the 2nd-best sgRNA method revealed 13 genes at toplevel statistical significance (p \leq 1e-4) (Figure 1F), while the weighted-sum method scored 7 genes at top significance (Figure S3C), where 6 of them were shared with the 2nd-best sgRNA method. In fact, these six genes (Cd247, Fam103a1, Hacvr2, Pdcd1, Prkar1a, and Stradb) are exactly identical to the 3-way shared hits identified by our custom FDR method. We also performed a parallel analysis with MAGeCK, a model-based algorithm for genome-scale CRISPR screens (Li et al., 2014), which

again revealed *Stradb*, *Pdcd1*, *Prkar1a*, and *Fam103a1* on top of the list (Figure 1G; Figure S3D). Examination of individual genes revealed strongly depleted sgRNAs targeting essential genes (e.g., *Hist1h4d*) and enriched sgRNAs targeting known or potentially new T cell phenotype modulators (Figure 1H). PD-1 and Tim-3 are well-known immunotherapy targets, thereby benchmarking the success and technical rigor of the screen. Collectively, these analyses demonstrate the rigor of *in vivo* CRISPR screen for CD8 T cell tumor infiltration and revealed targets for cancer immunotherapy.

We set out to independently identify genes that could modulate the ability of CD8 T cells to kill cancer cells. We performed a screen using degranulation assay (Figure 2A). We pulsed E0771 cells with varying concentrations of SIINFEKL peptide and found that they presented SIINFEKL peptide on surface MHC-I in a dose-dependent manner (Figure 2B). We isolated naive OT-I;Cas9 CD8 T cells and transduced them with MKO library, stimulated them for 6 days, rested them for 12 h, and then co-cultured them with SIINFEKL-pulsed E0771 cells at 1:1 of T cell: cancer cell (Effector: Target, or E: T) ratio in the presence of CD107a antibody (STAR Methods) (Betts et al., 2003). With three biological replicates, we sorted the top 5% CD107a+ cells (Figure 2C) and then subjected them to screen readout and data analysis (STAR Methods). Using the FDR <0.5% significance cutoff, we identified significantly enriched sgRNAs and their corresponding genes (Figure 2D), which represented potential negative regulators of T cell degranulation. Three genes emerged as common hits between the degranulation screen and the tumor infiltration screen (Dhx37, Lyn, and Odc1) (Figure 2E).

Immunotherapeutic Potential of *Dhx37* Gene-Edited CD8 T Cells

To examine the function of these genes in a model of immunotherapy, we cloned sgRNAs targeting Dhx37 and Odc1 that scored in both screens, along with Pdcd1 (which only scored in the infiltration screen) into the T cell lentiviral CRISPR vector. We did not pursue Lyn because it is not normally expressed in T cells (Tsantikos et al., 2009; Yamanashi et al., 1989). We then prepared lentivirus, transduced T cells, and adoptively transferred them into mice bearing E0771-mCh-OVA tumors (STAR Methods). Despite initially growing for 3 days post-adoptive transfer, the tumors treated by adoptive T cells regressed over the following 2.5 weeks (Figures 2F and 2G). All OT-I;Cas9 CD8 T cell-treatment groups demonstrated strong anti-tumor effects beginning 7 days after adoptive transfer (Figure 2G). Notably, sgDhx37 infected OT-I;Cas9 CD8 T cells significantly suppressed relapse when compared to mice treated with vector-infected T cells (Figure 2G). We observed that (5/5, 100%) of the mice in the sgDhx37 group had smaller tumors as compared to the vector group, compared to 2/5 in sgPdcd1 and 3/5 in sgOdc1 groups (Figure 2H). These data suggest that CRISPR targeting of Dhx37 enhanced the anti-tumor effects of OT-I;Cas9 CD8 T cells against E0771-mCh-OVA tumors.

Analysis of Gene Expression Signatures in *Dhx37*Tumor-Infiltrating Lymphocytes by Single-Cell RNA-Seq Dhx37 is a highly conserved DEAH box RNA helicase that has

Dhx3/ is a highly conserved DEAH box RNA helicase that has been reported to be involved with biogenesis of the small

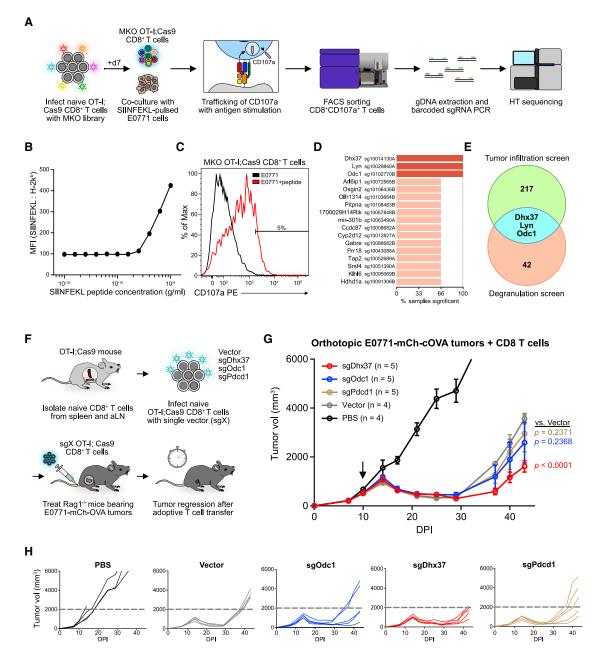


Figure 2. High-Throughput CRISPR Screen of CD8 T Cell Degranulation Identified Dhx37 and Odc1 as Top Hits Convergent with Infiltration

- (A) Schematics of experiment.
- (B) Titration of SIINFEKL peptide on MHC-I in E0771 cells.
- (C) Representative histogram of CD107a⁺ T cells analyzed from the co-culture of OT-I;Cas9 CD8 T cells and E0771 cancer cells.
- (D) Waterfall plot of the top-ranked sgRNAs across all sorted cell samples.
- (E) Venn diagram comparing the hits from the degranulation screen and from the tumor infiltration screen.
- (F) Schematics of efficacy testing experiments with adoptive transfer of single-gene knockout CD8 T cells
- (G) Growth curves of mammary fat pad E0771-mCh-OVA tumors in Rag1^{-/-} mice following different treatments: PBS control, adoptive transfer of OT-I;Cas9 CD8 T cells infected with lentiviral vector, lenti-sgDhx37, lenti-sgOdc1, and lenti-sgPdcd1. Note: arrow indicates the time of adoptive transfer of MKO or vector transduced OT-I; Cas9 CD8T cells. Data are shown as mean ± SEM. Statistical comparisons were made with two-way ANOVA for each group against vector, with
- (H) Spider plots of (G) separated by treatment group for visibility. See also Figures S4 and S5.

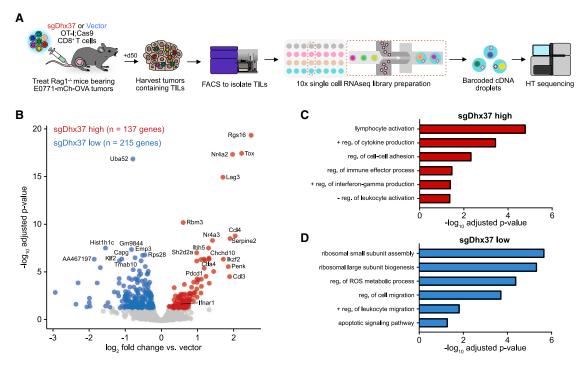


Figure 3. Single-Cell Transcriptomics of *Dhx37* Knockout CD8 Tumor-Infiltrating Lymphocytes

(A) Schematics of experiment.

(B) Volcano plot of differentially expressed genes in tumor-infiltrating CD8 cells treated with sgDhx37 compared to vector control.

(C) Gene Ontology (GO) analysis of significantly upregulated genes in sgDhx37-treated tumor-infiltrating CD8 cells.

(D) GO analysis of significantly downregulated genes in sgDhx37-treated tumor-infiltrating CD8 cells.

Note: for (C) and (D), significantly enriched (Bonferroni adjusted p < 0.05) GO categories were shown. See also Table S2.

ribosomal subunit through its release of U3 small nucleolar RNA (snoRNA) (Roychowdhury et al., 2019). Its homologs have also been found to regulate escape behavior via glycine receptor expression in zebrafish (Hirata et al., 2013) and cause microcephaly in rare gene variants (Karaca et al., 2015). However, this gene has not been previously associated with T cell function. The putative ATP-dependent RNA helicase domain and conservation implies that it might affect gene expression (Abdelhaleem, 2005). Because the heterogeneous tumor microenvironment might influence the functional state of individual TILs, we performed single-cell RNA sequencing (scRNA-seq) to investigate the transcriptomes of adoptively transferred sgDhx37-transduced OT-I;Cas9 CD8 TILs (sgDhx37 TILs) (Figure 3A) (STAR Methods).

After processing, stringent filtering and normalizing the raw scRNA-seq data (STAR Methods), the TIL scRNA-seq dataset was composed of 552 cells (sgDhx37, n = 191 cells; vector, n = 361 cells), collectively measuring 8,244 expressed genes in TILs (Table S2). We performed differential expression analysis between sgDhx37- and vector-treated TILs, identifying sets of significantly upregulated and downregulated genes (Table S2). 215 genes were significantly downregulated in sgDhx37 TILs, while 137 genes were significantly upregulated (Benjamini-Hochberg adjusted p < 0.05), with the mostly highly upregulated genes being *Rgs16*, *Tox*, and *Nr4a2* (Figure 3B). *Tox* has been recently implicated as a critical regulator of CD8 T cell and TILs (Alfei et al., 2019; Khan et al., 2019; Scott et al., 2019).

NR4A family members were found to regulate CAR-T potency against solid tumors (Chen et al., 2019). Other significantly upregulated genes included known immune-related genes such as Eomes, Nr4a3, Ccl4, Ifnar1, and Ikzf2 (Figure 3B; Table S2). The sgDhx37-upregulated genes also include genes involved in negative regulation of leukocyte activation such as Ctla4, Lag3, and Pdcd1, albeit to a lesser extent, which mirrors the positive correlation of T cell cytotoxic scores with these markers in TILs from human melanoma patients (Tirosh et al., 2016). As a gene set, gene ontology (GO) analysis revealed multiple immune-related pathways that were significantly upregulated in sgDhx37 TILs (adjusted p < 0.05), including lymphocyte activation, positive regulation of cytokine production, regulation of cell-cell adhesion, regulation of immune effector process, and positive regulation of interferon-gamma (IFN-γ) production (Figure 3C). Downregulated genes in sgDhx37 TILs are enriched in ribosomal small subunit assembly and ribosomal large subunit biogenesis (Figure 3D), consistent with its known role in rRNA modification and processing. Taken together, the scRNA-seq data revealed significant changes in the transcriptomes of sgDhx37 TILs in the heterogeneous tumor microenvironment at the single-cell level.

Validation and Immunologic Characterization of *Dhx37* as a Regulator of Primary CD8 T Cells

We generated an adeno-associated virus (AAV) vector (AAV-sgRNA-Thy1.1) capable of producing bona fide gene editing in

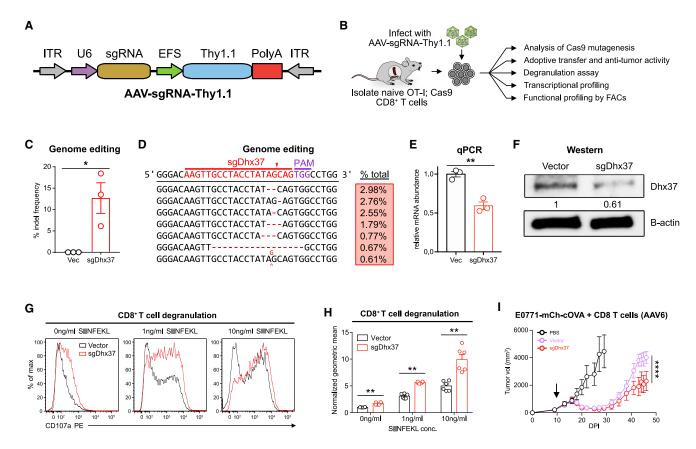


Figure 4. AAV-Mediated Gene Editing in Primary Murine CD8 T Cells and Effect of Dhx37 Perturbation

- (A) Schematics of the AAV-CRISPR T cell knockout vector.
- (B) Schematics of experiment.
- (C) Dhx37 gene editing in mouse CD8 T cells with AAV-CRISPR measured by T7E1 assay.
- (D) Representative Illumina targeted amplicon sequencing of the sgRNA target site 5 days after infection with AAV-sgDhx37. Top most frequent variants were shown, with the associated variant frequencies in the box to the right. PAM and sgRNA spacers were indicated above. Red arrows indicate predicted cleavage sites. Red dashed lines indicate deletions.
- (E) qPCR of Dhx37 mRNA level in murine CD8 T cells with AAV-sgDhx37 transduction.
- (F) Western blot of Dhx37 protein level in murine CD8 T cells with AAV-sqDhx37 transduction.
- (G) Representative flow histograms of CD107a degranulation assay for Dhx37 knockout CD8 T cells.
- (H) Quantification of data from (G) with independent experimental replicates.
- (I) Anti-tumor activity of adoptive transfer of AAV-CRISPR mediated Dhx37 knockout CD8 T cells. Growth curves of mammary fat pad E0771-mCh-OVA tumors in $Rag1^{-/-}$ mice following different treatments.

Note: in all bar plots, data are shown as mean \pm SEM. *p < 0.05, **p < 0.01, ***p < 0.001, by unpaired two-sided t test. In tumor growth curves, arrow indicates the time of adoptive transfer. ****p < 0.0001 by two-way ANOVA.

See also Figures S4 and S5 and Table S3.

primary murine CD8 T cells (Figure 4A) (STAR Methods). By isolating OT-I;Cas9 CD8 T cells and infecting them with respective AAV vectors, we performed molecular, functional, and immunological analysis of *Dhx37* loss, in addition to anti-tumor efficacy testing (Figure 4B). Amplicon sequencing showed that AAV-sgDhx37 potently mutagenized the targeted loci 5 days post-infection (Figures 4C and 4D; Table S3). This observation is corroborated with a reduction of *Dhx37* mRNA and Dhx37 protein levels in T cells (Figures 4E and 4F). The partial knockdown is likely due to incomplete gene editing efficiency, mRNA and Dhx37 protein stability at this time point, and the fact that not all indels lead to effective loss of protein, which together might under-estimate the effect of *Dhx37* mutagenesis by CRISPR.

Despite the incomplete knockout, we observed that compared to vector control, AAV-sgDhx37-treated OT-I;Cas9 CD8 T cells had significantly increased degranulation (surface CD107a) when co-cultured with cognate SIINFEKL-pulsed E0771 cells, at baseline (0 ng/mL), physiological (1 ng/mL), and saturation (10 ng/mL) conditions of anti-CD3 stimulation (Figures 4G and 4H). Moreover, we found no differences in cell growth between sgDhx37 and control T cells in culture (Figure S4A), ruling out the possibility that the degranulation result was simply due to a growth advantage.

Using the same E0771-mCh-OVA tumor model, we again validated that *Dhx37* perturbation by AAV-sgDhx37 showed enhanced efficacy of adoptive T cell therapy (Figure 4I). During

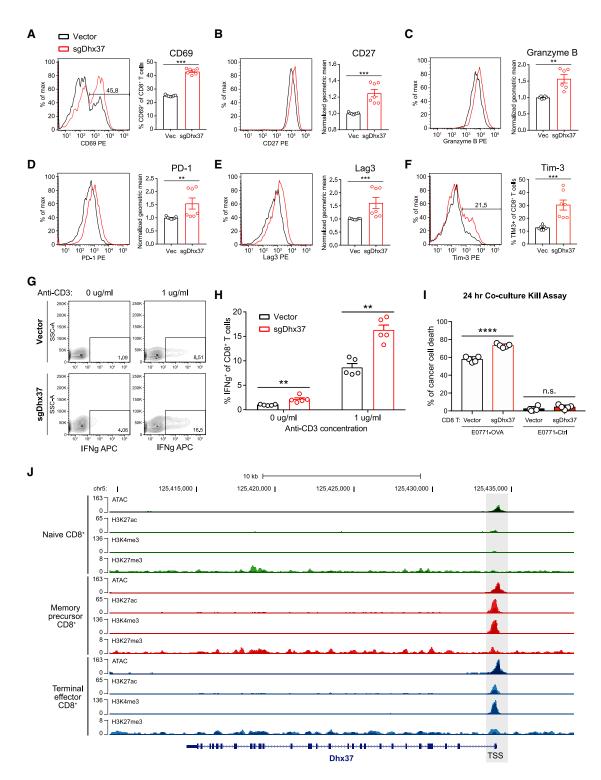


Figure 5. Immunological Characterization of Dhx37 Knockout Mouse CD8 T Cells

(A–F) Flow cytometry analysis of key immune markers of T cell function, including CD69 (A), CD27 (B), Granzyme B (C), PD-1 (D), Lag3 (E), and Tim-3 (F). Left panel in each plot is a representative histogram of sgDhx37 (red) and vector (black) CD8+ T cells. Right panel is a bar plot of quantification for each marker. Data are shown as mean \pm SEM. *p < 0.05, **p < 0.01, ***p < 0.001, by unpaired two-sided Mann-Whitney test.

- (G) Flow cytometry analysis plot of IFN- γ production of sgDhx37 and Vector CD8 T cells with or without anti-CD3 stimulation.
- (H) Quantification bar plot of (G). Data are shown as mean \pm SEM. **p < 0.01, by unpaired two-sided Mann-Whitney test.

(legend continued on next page)

the long-term in vivo tumor growth (6.5 week time course), we observed antigen decrease (Figures S4B-S4D), decrease of effective TILs (Figures S4F-S4H), and loss of T cell stimulation ability in cancer cells (Figures S5A-S5D), which is consistent with the current model of cancer immunoediting that might partly explain the ultimate rebound of tumors after adoptive transfer (Figure 2G and Figure 4I). These data validated that acute genetic perturbation of Dhx37 led to enhanced killing by CD8 T cells, both in vitro and in vivo, together further validating the screen results.

We then characterized the immunological features of Dhx37 mutant CD8 T cells. We found that Dhx37 mutant CD8 T cells have higher levels of an activation marker CD69, as well as a co-stimulatory receptor CD27 (Figures 5A and 5B). Concordantly, Dhx37 mutant cells produced higher level of granzyme B (Figure 5C). In line with the TIL scRNA-seq result, Dhx37 mutant have higher levels of surface PD-1, Lag3, and Tim-3 (Figures 5D-5F), which are known immune checkpoints and also linked to intrinsic T cell activation (Lee et al., 2011). We then measured the ability of *Dhx*37 mutant T cell's ability to produce IFN- γ , a major effector cytokine, and found that Dhx37 mutant produced higher level of IFN-γ both at baseline and upon 1 μg/mL anti-CD3 stimulation (Figures 5G and 5H). In a co-culture assay, we found that Dhx37 mutant (AAV-sgDhx37 treated) CD8 T cells are more potent in killing E0771-mCh-OVA cancer cell as compared to wild-type (vector treated) T cells (Figure 5I). The enhanced killing does not occur in parental E0771 cells without OVA, indicating that the effect is antigen specific (Figure 5I). By examining the epigenetic status of the Dhx37 locus, we also found that memory precursor and terminal effector CD8 T cells have high levels of chromatin accessibility (assay for transposase-accessible chromatin using sequencing [ATAC-seq] peaks), promoter activity (H3K4me3) and enhancer activity (H3K27ac), while naive CD8 T cells have low levels of these markers (Figure 5J). These data together link Dhx37 to T cell activation and effector function.

Of note, we also performed AAV-mediated gene editing and immunological assays for Odc1 (Figures S6A and S6B; Table S3). We found that Odc1 mutant T cells were able to degranulate in higher abundance in response to 1 ng/μL of SIINFEKL-pulsed E0771 cells compared to vector controls (Figures S6C and S6D). However, this increased responsiveness was not observed at a saturating dose of SIINFEKL peptide (10 ng/µL) (Figures S6C and S6D). Moreover, we revealed that Odc1 mutant T cells also as showed increased effector functions in terms of granzyme and IFN-γ production (Figures S6E–S6G) as well as higher levels of CD69, CD27, PD-1, Lag3, and Tim-3 (Figure S6G).

Bulk Transcriptome Profiling of *Dhx37* **Mutant CD8 T Cells Revealed Immunologic Signatures of T Cell** Activation and Effector Function along an NF-kB Axis

To identify genes downstream of Dhx37, we performed transcriptome profiling of *Dhx37* mutant CD8 T cells as compared to control T cells, using bulk mRNA-seq with 4 experimental replicates each (STAR Methods) (Table S4). We found a total of 97 significantly upregulated genes in Dhx37 mutant CD8 T cells (sgDhx37 high genes) and 69 downregulated genes (sgDhx37 low genes) (Figure 6A) (Table S5). Consistent with the immune phenotyping results, among the Dhx37 mutant upregulated genes are genes encoding effector cytokines such as various granzymes and IFN-γ, Cd69, Tim-3 (Havcr2), and other immune genes Serpinb6b and Serpinb9b (Figure 6B). Examples of genes downregulated upon Dhx37 loss include Foxp3, Il6ra, Lyz2, and Ly6c2 (Figure 6B). We validated 5 individual genes by single-gene qPCR, and confirmed that all 5, i.e., Gzmc, Gzmd, Sepinb9b, Tim-3, and Il6ra, are differentially expressed (Figure 6C).

Because there are no existing studies of *Dhx37* perturbation in T cells, we compared the differential expression signatures to the expression profiles of naive, memory, and effector T cells, as well as effector cells under induced activation. We found that the upregulated gene set in Dhx37 mutant is enriched in genes associated with T cell activation (Figure 6D; Table S6). We also found that the downregulated gene set is depleted in effector genes as compared to naive or memory T cells (Figure 6E; Table S6). Furthermore, DAVID pathway analysis (STAR Methods) revealed that the top two enriched pathways are immune response and cytolysis for Dhx37 mutant upregulated gene set (Figure 6F), with the next tier of significant terms being inflammatory response, cell-surface protein, and extracellular exosome (Table S6). These analyses further linked Dhx37 to T cell activation and effector function. Interestingly, when we examined the regulatory sequences in the promoters of genes upregulated upon Dhx37 loss, the strongest sequence motif corresponded to the NF-κB motif (p = 1.053e-5) (Figure 6G), suggesting that Dhx37 may act, at least in part, through NF-κB components.

Characterization of DHX37 in Human T Cells

Human DHX37 encodes a 1,150 amino acid protein with several conserved domains, including the DEAH box (DEXDc), HrpA, NTPase, H2A, and the helicase (HELICc) domains (Figure S7A). Sequence alignment of DHX37 homologs from 32 species showed that this protein is highly conserved, especially in the DEXDc and helicase domains (Figure S7A). We made N- and C-terminal reporter fusion constructs and used them together with endogenous anti-DHX37 antibody to perform staining and imaging (STAR Methods). We found that both exogenous GFP-DHX37 and DHX37-GFP fusions overlap with those of the endogenous DHX37 protein, and they localize in nucleoli (Figures S7B and S7C). By western blot, we found that DHX37 protein is expressed in primary human CD4⁺ and CD8⁺ T cells, as well as patient TILs (Figure 7A). Based on the transcriptomic signature, pathway analysis, and motif analysis, we hypothesized that DHX37 interacts with NF-κB components. We performed immunoprecipitation (IP) on endogenous DHX37 protein with lysate

⁽I) Co-culture luciferase assay that measures the ability of sgDhx37 and Vector-treated CD8 T cells to kill antigen-expressing (E0771-OVA) and parental control (E0771-Ctrl) cells. Data are shown as mean ± SEM. ****p < 0.0001, by unpaired two-sided t test.

⁽J) Normalized signal tracks of ATAC-seq, H3K27ac, H3K4me3, and H3K27me3 chromatin immunoprecipitation sequencing (ChIP-seq) at the Dhx37 locus in naive, memory precursor, and terminal effector CD8+ T cells. See also Figure S6.

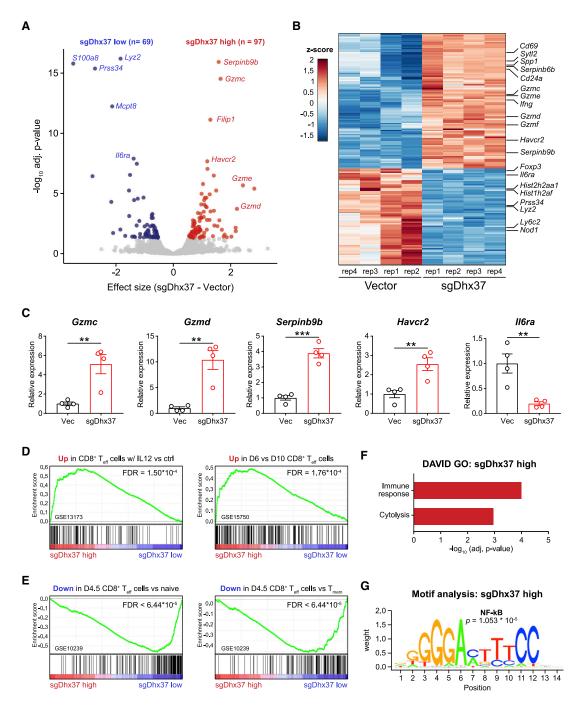


Figure 6. Transcriptome Analysis of Dhx37 Knockout Mouse CD8 T Cells by mRNA-Seq

(A) Volcano plot of differentially expressed genes between *Dhx37* knockout and Vector control CD8 T cells, as quantified by bulk mRNA-seq (n = 4 biological replicates each).

- (B) Heatmap of differentially mRNA-seq expressed genes in between *Dhx37* knockout and Vector control CD8 T cells.
- (C) qPCR single-gene validation of representative highly upregulated or downregulated genes, including Gzmc, Gzmd, Serpinb9b, Tim-3, and Il6ra. Data are shown as mean \pm SEM. *p < 0.05, **p < 0.01, ***p < 0.001, by unpaired two-sided t test.
- (D and E) GSEA analysis of differentially expressed (DE) genes in *Dhx37-DE* upregulated gene set in (left) gene signature of IL-12 treatment versus control T effector cell and in (right) D6 versus D12 T effector cell in culture. (E) Analysis of *Dhx37-DE* downregulated gene set in (left) gene signature of effector versus naive CD8 T cell and in (right) effector versus memory CD8 T cell. (F) DAVID GO analysis of *Dhx37-DE* upregulated gene set.
- (G) Motif analysis of the regulatory DNA elements of the *Dhx37-DE* upregulated gene set, identifying the consensus NF-κB binding site as the top enriched motif. See also Tables S4, S5, S6.

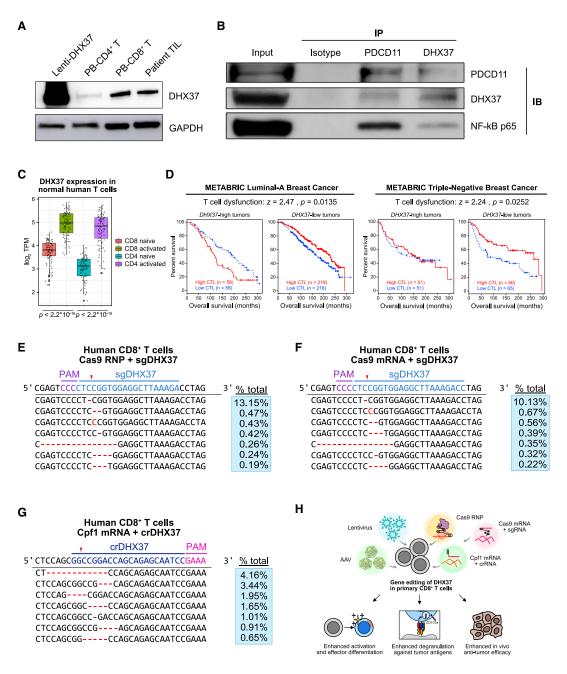


Figure 7. Characterization of DHX37 in Human T Cells

(A) Western blot analysis of DHX37 protein in human cells. From left to right, HEK293FT cells with DHX37 cDNA overexpression as a positive control; peripheral blood CD4+ T cells from a healthy donor; peripheral blood CD8+ T cells from a healthy donor; TILs from a tumor biopsy freshly isolated from an non-small-cell lung carcinoma (NSCLC) patient.

- (B) CoIP western experiment of DHX37 and NF-kB pathway components (PDCD11 and p65) in human primary CD8 T cells. Anti-DHX37 and Anti-PDCD11 reciprocal IPs were blotted with endogenous antibody against PDCD11, DHX37, and also along with p65.
- (C) DHX37 mRNA expression in normal human CD8+ and CD4+ T cells in activated and naive states. Statistical comparisons were made with two-sided unpaired
- (D) TIDE analyses of DHX37 expression in T cell dysfunction signatures linked to survival benefits in patients, with luminal-A or TNBC.
- (E-G) Gene editing of the DHX37 locus in primary human CD8 T cells, with Cas9 RNP (E), Cas9 mRNA (F), or Cpf1 mRNA (G) systems. Representative Illumina targeted amplicon sequencing of the sgRNA target site. Top most frequent variants were shown, with the associated variant frequencies in the box to the right. PAM and sgRNA spacers were indicated above. Red arrows indicate predicted cleavage sites. Red dashed lines indicate deletions. Red A/C/G/T where applicable indicate insertions.
- (H) Schematics summarizing DHX37 targeting. See also Figure S7 and Table S3.

from primary human CD8 T cells. We found that the NF- κ B binding protein (PDCD11) and the NF- κ B transcription factor p65 were detected in the DHX37-IP from CD8 T cell lysate (Figure 7B). We then performed a reciprocal IP with anti-PDCD11 antibody and detected both p65 and DHX37 in the PDCD11-IP fraction (Figure 7B).

Human DHX37 is expressed in most organs, with highest expression in lymphoid tissues such as the bone marrow, lymph nodes, spleen, and appendix (Fagerberg et al., 2014). By analyzing DHX37 expression in normal human T cells according to activation state, we found that activated T cells, both CD8+ and CD4+, have higher levels of DHX37 (Figure 7C). By analyzing scRNA-seq data of human CD8+ TILs from cancer patients, we found that DHX37 is expressed at higher level in exhausted TILs compared to non-exhausted (Figure S7D). To examine the relevant roles of DHX37 in human T cells, we further analyzed a recent single-cell transcriptomics dataset of human TILs (Zheng et al., 2017a). DHX37 mRNA expression was detected in a fraction of peripheral blood T cells, tissue-resident T cells, and tumor-infiltrating T cells (Figure S7E). DHX37 was expressed in CD3⁺/CD8⁺ (cytotoxic T cell), CD3⁺/CD4⁺/CD25⁻ (T helper cell), and CD3⁺/CD4⁺/CD25⁺ (Treg) populations (Figure S7E), in those T cells from peripheral blood, resident tissue, tumornormal junction, or infiltrated in the tumor (Figure S7E). To assess the clinical relevance of DHX37 in cancer, we used Tumor Immune Dysfunction and Exclusion (TIDE) analysis to examine whether DHX37 expression in T cells is linked to breast cancer outcomes (STAR Methods). While high cytotoxic T lymphocyte (CTL) score is associated with an overall survival benefit, highlevel of DHX37 expression abolishes or negates this benefit, in both luminal-A and TNBC subtypes (Figure 7D). Together, these data showed that DHX37 is expressed in normal and tumorassociated human T cells, interacts with NF-κB core components, and is associated with T cell dysfunction related clinical outcomes in breast cancer patients.

DISCUSSION

Tumor-derived TILs possess the ability to recognize and kill cancer cells in autologous transplantation studies (Rosenberg et al., 1985). The discovery of CTLA-4 and PD-1 pathways revealed how inhibitory signals of CD8 T cells can be exploited by tumors to evade immune recognition (Ishida et al., 1992; Leach et al., 1996). Targeting T cell inhibitory factors has led to the paradigm of checkpoint blockade immunotherapy (Fesnak et al., 2016; Maude et al., 2014; Maus and June, 2016; Mellman et al., 2011; Ribas, 2012), suggesting that better understanding of the genetic landscape governing T cell function will facilitate the development of better immunotherapies. CRISPR technology has provided a set of tools to open unique opportunities for direct manipulation of the T cell genome (Eyquem et al., 2017), and its scalability enabled genetic screens in T cells, which to date have been performed in vitro (Henriksson et al., 2019; Shifrut et al., 2018). Here, we performed convergent in vivo tumor infiltration and in vitro degranulation screens to systematically study the anti-tumor potential of CD8 T cells. Our screens generated genome-scale quantitative measurements of genetic factors modulating CD8 T cell phenotypes directly in immunotherapy-relevant settings, and identified enriched targets of various functional categories including known and undocumented.

A particularly exciting aspect of the infiltration screen is that, unlike other screens to date, PD-1 and Tim-3 emerged as top hits in an unbiased manner. These markers highlight the sensitivity of the in vivo screen in physiologically relevant conditions. This study can also serve as a reference for high-throughput genetic interrogation of T cells in vivo for immunotherapy target discovery. Possible alternative applications include a variety of cancer models such as genetically engineered mouse models as well as genome-editing based cancer models for diverse oncology indications, with or without other treatments in conjunction. Other CRISPR technologies such as CRISPR activation or repression, other vector or screening systems, and orthogonal genetic manipulation strategies may further empower the study of cancer immunotherapy with available and evolving toolkits (Dominguez et al., 2016; Esvelt et al., 2013; Gilbert et al., 2013; Thakore et al., 2016).

Validation and characterization of a top convergent hit, *Dhx37*, showed that modulation of this gene in CD8 T cells can lead to enhanced anti-tumor activity *in vitro* and *in vivo* by adoptive transfer. Due to the current lack of available compounds targeting DHX37, we reasoned that genetically editing *DHX37* in human CD8 T cells presented an approach for attenuating its activity. We performed ribonucleoprotein (RNP)-mediated and mRNA-mediated editing of *DHX37* in human CD8 T cells, using both Cas9 (Figures 7E and 7F; Table S3), and Cas12a-Cpf1 (Figure 7G; Table S3) systems. Multiple modes of delivery can be applied, including lentivirus, AAV, RNP, or mRNA (Figure 7H). Future studies may develop compounds targeting DHX37 by screening small molecule libraries or may use structure-based drug design.

STAR*METHODS

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AUTHOR CONTRIBUTIONS

S.C. and M.B.D. conceived, designed, and initiated the study. M.B.D. and S.C. generated T cell lentiCRISPR vectors and libraries. S.C., M.B.D., and G.W. generated relevant mouse strains. L.Y. generated T cell AAV-CRISPR system. M.B.D., G.W., and L.Y. performed most *in vivo* experiments. M.B.D. performed *in vitro* T cell immunology assays. R.D.C., J.J.P., and S.C. analyzed CRISPR screen data. R.D.C. analyzed mRNA-seq, single-cell, and immunogenomics data. L.Z. performed biochemistry experiments. X.D. isolated patient TILs and assisted human T cell experiments. H.R.K. assisted MD for various experiments. Y.E. performed histology experiments. C.D.G. assisted T cell library and sgRNA cloning. K.Y.C., X.Z., P.R., Y.D., J.S., S.Z.L., and J.J.Z. assisted parts of experiments. D.R.L. and R.S.H. provided clinical insights and patient samples for analysis. M.B.D., R.D.C., and S.C. jointly prepared the manuscript with inputs from all authors. S.C. secured funding and supervised the work.

DECLARATION OF INTERESTS

R.S.H. is a consultant/advisory board member for Abbvie Pharmaceuticals, AstraZeneca, Biodesix, Bristol-Myers Squibb, Eli Lilly and Company, EMD Serrano, Genentech/Roche, Heat Biologics, Infinity Pharmaceuticals, Junshi Pharmaceuticals, Loxo Oncology, Merck and Company, Nektar, Neon Therapeutics, NextCure, Novartis, Pfizer, Sanofi, Seattle Genetics, Shire PLC, Spectrum Pharmaceuticals, Symphogen, and TESARO; has sponsored research grants from AstraZeneca, Eli Lilly and Company, and Merck and Company; and is a board member (non-executive/independent) for Junshi Pharmaceuticals. Yale filed a patent application related to the data described in this study.

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STAR***METHODS**

KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
Antibodies		
VeriBlot for IP Detection Reagent (HRP)	AbCam	Cat#ab131366-500ul
DHX37 Polyclonal Antibody	Bethyl	Cat#A300-856A; RRID:AB_2292875
PDCD11 Polyclonal Antibody	Bethyl	Cat#A303805A; RRID:AB_11218382
APC anti-human CD3 Antibody (Clone: HIT3a)	Biolegend	Cat#300312; RRID:AB_314048
APC anti-mouse CD3ε Antibody (clone: 145-2C11)	Biolegend	Cat#100312; RRID:AB_312677
APC anti-mouse IFN-γ Antibody (clone: XMG1.2)	Biolegend	Cat#505810; RRID:AB_315404
FITC anti-mouse CD8a Antibody (clone: 53-6.7)	Biolegend	Cat#100706; RRID:AB_312745
PE anti-human IgG Fc Antibody (clone: HP6017)	Biolegend	Cat#409304; RRID:AB_10895907
PE anti-human/mouse Granzyme B Recombinant Antibody (clone: QA16A02)	Biolegend	Cat#372208; RRID:AB_2687032
PE anti-mouse CD107a (LAMP-1) Antibody (clone: 1D4B)	Biolegend	Cat#121612; RRID:AB_2134487
PE anti-mouse CD223 (LAG-3) Antibody (clone: C9B7W)	Biolegend	Cat#125208; RRID:AB_2133343
PE anti-mouse CD279 (PD-1) Antibody (clone: 29F.1A12)	Biolegend	Cat#135206; RRID:AB_1877231
PE anti-mouse CD366 (Tim-3) Antibody (clone: B8.2C12)	Biolegend	Cat#134003; RRID:AB_1626181
PE anti-mouse CD62L Antibody (clone: MEL-14)	Biolegend	Cat#104408; RRID:AB_313095
PE anti-mouse CD69 Antibody (clone: H1.2F3)	Biolegend	Cat#104508; RRID:AB_313111
PE anti-mouse H-2Kb bound to SIINFEKL Antibody (clone: 25-D1.16)	Biolegend	Cat#141604; RRID:AB_10895905
PE anti-mouse/human CD44 Antibody (clone: IM7)	Biolegend	Cat#103008; RRID:AB_312959
PE anti-mouse/rat/human CD27 Antibody (clone: LG.3A10)	Biolegend	Cat#124210; RRID:AB_1236459
PE anti-rat CD90/mouse CD90.1 (Thy-1.1) Antibody (clone: OX-7)	Biolegend	Cat#202524; RRID:AB_1595524
PE/Cy7 anti-mouse CD3epsilon Antibody (clone: 145-2C11)	Biolegend	Cat#100320; RRID:AB_312685
PerCP/Cy5.5 anti-mouse CD45.2 Antibody (clone: 104)	Biolegend	Cat#109828; RRID:AB_893350
Ultra-LEAF Purified anti-mouse CD3ε Antibody (clone: 145-2C11)	Biolegend	Cat#100340; RRID:AB_11149115
Ultra-LEAF(TM) Purified anti-mouse CD28 Antibody (clone: 37.51)	Biolegend	Cat#102116; RRID:AB_11147170
Ultra-LEAF Purified anti-mouse CD16/32 Antibody (clone: 93)	Biolegend	Cat#101330; RRID:AB_2561482
beta actin loading control (clone: BA3R)	Invitrogen	Cat#MA5-15739; RRID:AB_10979409
Dynabeads Human T-Activator CD3/CD28	Invitrogen	Cat#11131D
LIVE/DEAD Fixable Near-IR Dead Cell Stain Kit, for 633 or 635 nm excitation	Invitrogen	Cat#L34976
NFkB p65 Antibody (4-2HCLC), ABfinity Rabbit Oligoclonal (lot:)	Invitrogen	Cat#710048; RRID:AB_2532539
Rabbit polyclonal anti-human/mouse Dhx37 (Lot: R44921)	Novus	Cat#NBP2-13922
GAPDH Antibody (clone: 6C5)	Santa Cruz	Cat#sc-32233; RRID:AB_627679
anti-GFP	AbCam	Cat#ab290; RRID:AB_303395
Anti-Nucleolin antibody	AbCam	Cat#ab22758; RRID:AB_776878
Bacterial and Virus Strains		
One Shot Stbl3 Chemical Competent E. coli	ThermoFisher	Cat#C737303
Endura ElectroCompetent Cells	Lucigen	Cat#60242-2
Chemicals, Peptides, and Recombinant Proteins		
DPBS, no calcium, no magnesium	GIBCO	Cat#14190250
RPMI 1640 Medium	GIBCO	Cat#11875-093
DMEM, high glucose, pyruvate	GIBCO	Cat#11995065
Fetal Bovine Serum	Sigma Aldrich	Cat#F4135-500ML
Penicillin-Streptomycin (10,000 U/mL)	GIBCO	Cat#15140122

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REAGENT or RESOURCE	SOURCE	IDENTIFIER
Lonza BioWhittaker L-Glutamine (200mM)	Lonza	Cat#BW17605E
2-Mercaptoethanol	Sigma Aldrich	Cat#M6250-10ML
X-VIVO 15 Serum-free Hematopoietic Cell Medium	Lonza	Cat#BE02-060F
Corning; Human AB Serum; Male Donors; type AB; US; 100mL, 35-060-CI 1/EA	Corning	Cat#MT35060CI
ACK Lysing Buffer	Lonza	Cat#10-548E
Naive CD8a+ T Cell Isolation Kit, mouse	Miltenyi	Cat#130-096-543
LS Columns	Miltenyi	Cat#130-042-401
Recombinant Mouse IL-2 (carrier-free)	Biolegend	Cat#575404
Recombinant Mouse IL-7 (carrier-free)	Biolegend	Cat#577802
Recombinant Mouse IL-12 (p70) (carrier-free)	Biolegend	Cat#
Recombinant Mouse IL-15 (carrier-free)	Biolegend	Cat#566302
Recombinant Human IL-2 (carrier-free)	Biolegend	Cat#589104
Dynabeads Human T-Activator CD3/CD28	Invitrogen	Cat#11131D
Ficoll-Paque PLUS	GE Healthcare Life Sciences	Cat#17144002
Collagenase IV	ThermoFisher	Cat#17104019
OVA 257-264	Invivogen	Cat#vac-sin
Monensin Solution (1,000x)	Biolegend	Cat#420701
QuickExtract DNA Extraction Solution	Epicenter	Cat#QE09050
QIAamp Fast DNA Tissue Kit	QIAGEN	Cat#51404
T7 Endonuclease I	NEB	Cat#M0302L
Proteinase K	QIAGEN	Cat#19131
RNase A	QIAGEN	Cat#19101
Gibson Assembly Master Mix	NEB	Cat#E2611
Phusion Flash High-Fidelity PCR Master Mix	ThermoFisher	Cat#F548L
DreamTaq Green PCR Master Mix (2X)	ThermoFisher	Cat#K1082
QIAquick Gel Extraction Kit	QIAGEN	Cat#28706
E-Gel Low Range Quantitative DNA Ladder	ThermoFisher	Cat#12373031
Fixation/Permeabilization Solution Kit	BD	Cat#554714
Nextera DNA Library Prep Kit	Illumina	Cat#FC-121-1030
Nextera Index Kit	Illumina	Cat#FC-121-1011
TRIzol Reagent	Invitrogen	Cat#15596026
PureLink RNA Mini Kit	Invitrogen	Cat#12183018A
PureLink DNase Set	Invitrogen	Cat#12185010
SuperScript IV Reverse Transcriptase	Invitrogen	Cat#18090050
Random Hexamers (50 μM)	Invitrogen	Cat#N8080127
TaqMan Fast Advanced Master Mix	Invitrogen	Cat#4444557
Bpil (Bbsl) (10 U/μL)	ThermoFisher	Cat#ER1012
Esp3I (BsmBI) (10 U/μL)	ThermoFisher	Cat#ER0451
Benzonase Nuclease	ThermoFisher	Cat#E1014-25KU
Dynabeads Protein A for Immunoprecipitation	Invitrogen	Cat#10002D
NuPAGE LDS Sample Buffer (4X)	Invitrogen	Cat#NP0008
NuPAGE 3-8% Tris-Acetate Protein Gels, 1.5 mm, 10-well	Invitrogen	Cat#EA0378BOX
4–20% Mini-PROTEAN TGX Precast Protein Gels, 10-well	BioRad	Cat#4561094
Bovine Serum Albumin	Sigma Aldrich	Cat#A9418-100G
Pierce ECL Western Blotting Substrate	ThermoFisher	Cat#32106
EDTA	Sigma Aldrich	Cat#E8008-100ML
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REAGENT or RESOURCE	SOURCE	IDENTIFIER
XenoLight D-Luciferin - K+ Salt Bioluminescent Substrate	Perkin Elmer	Cat#122799
Neon Transfection System 100 μL Kit	Invitrogen	Cat#MPK10025
Cas9 protein	IDT	Cat#1081058
Cas9 mRNA	Trilink	Cat#L-7206
Cpf1 mRNA	Trilink	Cat#L-7009
Cas9 tracrRNA	IDT	Cat#1072532
DHX37 crRNA (Cas9)	IDT	Custom, sequence specific
DHX37 crRNA (Cpf1)	Sigma	Custom, sequence specific
Critical Commercial Assays		
Qubit dsDNA HS Assay Kit	ThermoFisher	Cat#Q32854
Chromium Single Cell	10x Genomics	V2 discontinued, alternative is v3, Cat#1000092
Deposited Data		
Human T cell single cell RNA-seq	Zheng et al., 2017a	GEO: GSE98638
Dhx37 single cell RNaseq	This paper, GEO	GEO: GSE132960
Dhx37 mRNaseq	This paper, GEO	GEO: GSE132960
Dhx37 and Odc1 CRISPR Cas9 and Cas12a/Cpf1 mutagenesis indels	This paper, SRA	SRA: PRJNA549266
qPCR data	This paper, Mendeley Data	https://doi.org/10.17632/2ffxmf3d7k.1
Flow data	This paper, Mendeley Data	https://doi.org/10.17632/2ffxmf3d7k.1
Experimental Models: Cell Lines		
HEK293FT	ThermoFisher	Catalog Number: R70007
E0771	CH3	Catalog Number: 940001
Experimental Models: Organisms/Strains		
OT-I	Jackson Laboratory	Catalog Number: 003831
Cas9	Chu et al., 2016; Platt et al., 2014	Jackson Lab
OT-I;Cas9	This paper	N/A
Rag1-/-	Jackson Laboratory	Catalog Number: 002216
Oligonucleotides		
guide RNA spacer oligos	This paper	Table S7
amplicon primers for surveyor or Nextera	This paper	Table S7
Recombinant DNA		
MGC Mouse Dhx37 cDNA	Dharmacon	MMM1013-202797609
DHX37-GFP	This paper	Addgene, N/A
GFP-DHX37	This paper	Addgene, N/A
T cell LentiCRISPR vector (pSC008)	This paper	Addgene, N/A
T cell AAVCRISPR vector (pLY05)	This paper	Addgene, N/A
mGeCKOa	Sanjana et al., 2014	Discontinued, alternative is Pooled Library #100000052
mGeCKOb	Sanjana et al., 2014	Discontinued, alternative is Pooled Library #1000000052
Software and Algorithms		
Software and Algorithms FlowJo software 9.9.6	FlowJo	https://www.flowio.com
Software and Algorithms FlowJo software 9.9.6 Cutadapt	FlowJo Martin, 2011	https://www.flowjo.com https://cutadapt.readthedocs.io/en/stable/

(Continued on next page)

Continued		
REAGENT or RESOURCE	SOURCE	IDENTIFIER
DAVID	Huang et al., 2009	https://david.ncifcrf.gov
Cell Ranger 1.3	10x Genomics (Zheng et al., 2017b)	https://www.10xgenomics.com/solutions/single-cell/
Seurat	Butler et al., 2018	https://satijalab.org/seurat/
Scran R package	Lun et al., 2016	https://www.r-project.org
Rtsne R package	van der Maaten, 2014; van der Maaten and Hinton, 2008	https://www.r-project.org
Edge R package	Robinson et al., 2010	https://www.r-project.org
RIGER	Luo et al., 2008	https://software.broadinstitute.org/ GENE-E/extensions.html
MAGeCK	Li et al., 2014	https://sourceforge.net/p/mageck/wiki/Home/
Kallisto	Bray et al., 2016	https://pachterlab.github.io/kallisto/
Sleuth	Pimentel et al., 2017	https://pachterlab.github.io/sleuth/
GSEA	Subramanian et al., 2005	http://software.broadinstitute.org/gsea/index.jsp
PSCAN	Zambelli et al., 2009	http://159.149.160.88/pscan/

LEAD CONTACT AND MATERIALS AVAILABILITY

Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Sidi Chen (sidi.chen@yale.edu). A list of critical reagents (key resources) is included in the Key Resources Table. Relevant plasmids are deposited to Addgene and available to the academic community. For additional materials, please email the lead contact for requests. Some material may require requests to collaborators and/or agreements with various entities. Requests are reviewed by Yale University regarding intellectual property or confidentiality obligations. Material that can be shared will be released via a Material Transfer Agreement.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

All animal work was approved by IACUC and performed with approved protocols (Chen-2015-20068; Chen-2018-20068). Mice, both sexes, between the ages of 6-12 weeks of age were used for the study. OT-I TCR transgenic mice (OT-I mice) were described by (Hogquist et al., 1994). Constitutive Cas9-2A-EGFP mice (Cas9 mice) were described by (Chu et al., 2016; Platt et al., 2014). Rag1^{-/-} were described by (Mombaerts et al., 1992). OT-I; Cas9 mice were generated by breeding OT-I and Cas9 mice and genotyped according to Jackson Lab protocol. Naive CD8 T cells were isolated from OT-I mice, Cas9 mice, and OT-I;Cas9 mice using Miltenyi bead-based kits. Tumor-transplantation studies were conducted in Rag1-/- mice. All animals were housed in standard individually ventilated, pathogen-free conditions, with 12h:12h or 13h:11h light cycle, room temperature (21-23°C) and 40%-60% relative humidity. When a cohort of animals were receiving multiple treatments, animals were randomized by 1) randomly assign animals to different groups using littermates, 2) random mixing of females prior to treatment, maximizing the evenness or representation of mice from different cages in each group, and/or 3) random assignment of mice to each group, in order to minimize the effect of gender, litter, small difference in age, cage, housing position, where applicable. Average tumor sizes were consistent between treatment groups to account for selection bias.

METHOD DETAILS

Generation of a T cell CRISPR vector

A lentiviral T cell CRISPR vector, lenti-pLKO-U6-sgRNA(BsmBI)-EFS-Thy1.1CO-spA, was generated by codon-optimizing and subcloning Thy1.1 and sgRNA expression cassette into a lentiviral vector via Gibson assembly. Vector expresses an sgRNA with a non-targeting spacer.

Genome-scale mouse T cell CRISPR library cloning

Two sub-libraries (mGeCKOa and mGeCKOb) from (Sanjana et al., 2014) were sub-cloned in equal molar ratios by Gibson assembly into the T cell CRISPR vector to generate the genome-scale mouse T cell CRISPR library (MKO). The resulting library consisted of 129,209 sgRNAs including 1,000 non-targeting controls (NTCs) and 128,209 unique sgRNAs targeting every gene in the genome. An estimated library coverage of > 50x ($\sim 7 \times 10^6$ total colonies) was achieved by electroporation. The library was subsequently sequence-verified by Illumina sequencing.

Viral library production

The MKO library plasmid was transfected into low-passage HEK293FT cells at 80% confluency in 15cm tissue culture plates. Viral supernatant was collected at 48h and 72h post-transfection, filtered via a 0.45 µm filtration unit (Fisher/VWR), and concentrated using AmiconUltra 100kD ultracentrifugation units (Millipore). The concentrated supernatant was subsequently aliquoted and stored in −80°C until use. Virus for empty vector was produced in a similar manner.

T cell isolation and culture

Spleens and mesenteric lymph nodes (mLNs) were isolated from indicated mouse strains and placed in ice-cold 2% FBS [FBS (Sigma) + PBS (GIBCO)]. Organs were prepared by mashing organs through a 100 μm filter. Lymphocytes were suspended in 2% FBS. RBCs were lysed with ACK Lysis Buffer (Lonza), incubated for 2 mins at room temperature, and washed with 2% FBS. Lymphocytes were filtered through a 40 μm filter and resuspended with MACS Buffer (PBS + 0.5% BSA + 2 μM EDTA). Naive CD8 T cells were isolated using naive mouse CD8 T cell isolation kit from Miltenyi. Naive CD8 T cells were resuspended with cRPMI (RPMI-1640 + 10% FBS + 2mM L-Glutamine + 100U Pen/Strep (Fisher) + 49nM β-mercaptoethanol (Sigma)) to a final concentration of 1 × 10⁶ cells/ml. Medium for in vivo experiments was supplemented with 2ng/ml IL-2 + 2.5ng/ml IL-7 + 50ng/ml IL-15 + 1 μg/ml anti-CD28. Medium for in vitro experiments was supplemented with 2ng/ml IL-2 + 2ng/ml IL-12p70 + 1 µg/ml anti-CD28. Cells were cultured on plates pretreated with 5 µg/ml anti-CD3. Cytokines and antibodies mentioned above were purchased from BioLegend.

T cell transduction, virus titration

T cells were infected in culture immediately after isolation by directly adding concentrated virus into the media. Cells were cultured for 3 days after viral transduction to provide sufficient time for sgRNA expression. T cells were stained for Thy1.1 expression to determine viral transduction efficiency and analyzed on FACS. Viral titer was determined for each batch by the number of Thy1.1+T cells normalized to total T cells divided by the volume of virus used. At least 3 doses of viruses with experimental duplicates were used for determining viral titer.

Antibody and Flow Cytometry

Infectivity of CD8 T cells was assessed via surface staining with anti-CD3ε APC, anti-CD8α FITC, and anti-Thy1.1 PE. Biolegend antibodies used in experiments anti-CD3ε PE/Cy7, anti-CD8α APC, anti-CD16/32, anti-CD27 PE, anti-CD44 PE, anti-CD45 PerCP/ Cy5.5, anti-CD62L PE, anti-CD69 PE, anti-CD107a PE, anti-CD279 PE, anti-IFNγ APC, anti-Granzyme B PE, anti-Lag3 PE, anti-SIINFEKL:H-2Kb PE, anti-Thy1.1 PE, and anti-Tim3 PE. Anti-human/mouse DHX37/Dhx37 purchased from Novus Biologicals and anti-NF-kB p65 purchased from Invitrogen were used for immunoblotting. Anti-human DHX37 and anti-human PDCD11 purchased from Bethyl Laboratory was used for immunoprecipitation. Anti-β-actin purchased from Invitrogen and anti-GAPDH purchased from Santa Cruz Biotechnology was used for loading control. For surface stains, cells were stained on ice for 30 mins. Samples were collected on a BD FACSAria cell sorter with 3 lasers, and analyzed using FlowJo software 9.9.6 (Treestar, Ashland, OR) on a MAC workstation.

Library-scale viral transduction of T cells

T cells were isolated and cultured as described previously. Similar to previous studies (Chen et al., 2014; Zhou et al., 2014), for each infection replicate, a total of > 1 × 108 Cas9 or naive OT-I; Cas9 CD8 T cells were transduced with concentrated lentivirus containing the MKO library described above, to achieve an initial library coverage of > 700x. Transduction with the virus containing the empty vector was performed in parallel with a total of > 1 × 10⁷ naive CD8 T cells. Multiple computational strategies were used to generate quantitative ranked list of genes and to distinguish mis-associations or false-positives from potential true hits with strong selection, including usage of independent infection replicates, different mice, independent screens, different TCRs, and independent sgRNAs (later in methods). In each experiment, 3 infection replicates were applied for generation of T cell library unless otherwise noted.

Adoptive transfer of viral library infected T cells

At day 0 of the culture, naive CD8 T cells were infected with the lentiviral MKO library and incubated at 37°C for 3 days. On day 3 of culture, T cells were collected, washed with ice-cold PBS, and resuspended to a final concentration of 5×10^7 cells/ml. 1×10^7 cells were injected intravenously into each mouse. All adoptive cell transfer experiments in this study were done without lymphodepletion. On 7-day post-transfer, mice were euthanized, and relevant organs were isolated.

Generation of a model antigen expression vector

A lentiviral mCherry-OVA (mCh-OVA) vector, lenti-pLKO-U6-sg(BsmBl)-EFS-mCherry-2A-OVA, was generated by subcloning OVA into a mCherry lentiviral vector via Gibson assembly.

Generation of mCherry-OVA cell line

E0771 murine breast cancer cells were transduced with mCh-OVA-expressing lentivirus. After 3 days post-transduction, transduced E0771 cells were cultured individually in 96-well plate by resuspending cells to 10 cells/ml and culturing 100 μL of cell suspension in each well. 2 weeks later, clonal mCh⁺ E0771 clones were identified by fluorescence light microscopy. mCh⁺ E0771 clones were stained with established anti-mouse SIINFEKL: H-2K^b antibody (Porgador et al., 1997) to determine OVA expression. Different E0771-mCh-OVA clones were selected based on OVA expression. Clone 3 was chosen for in vivo experiments because of its low, uniform expression of OVA to select for genes with stronger phenotypes.

Tumor transplantation and tissue processing

 5×10^6 E0771-mCh-OVA cells were either injected either subcutaneously or into the mammary fat pad of Rag1^{-/-} mice. 10 days post-transplantation, viral library infected T cells, prepared as previously described, were intravenously injected in tumor-bearing Rag1^{-/-} mice. After 7 days, draining lymph nodes, non-draining lymph nodes, spleens, lungs, and tumors were isolated. Samples were prepared for DNA extraction or FACS analysis. Spleens and lymph nodes were prepared as previously described. Tumors were broken down into smaller fragments, about the size of lentils. Tumors were then dissociated with 1 μg/ml Collagenase IV for 30 minutes using GentleMacs Octo Dissociator from Miltenyi, and cell suspensions were passed through 100 μm filter twice before staining.

Antigen specificity testing for OT-I T cells

Naive CD8 T cells were isolated from OT-I; Cas9 mice using Naive mouse CD8 T cell kit from Miltenyi. Cells were cultured for 3 days in the presence of plate bound anti-CD3 (5ug/ml), soluble anti-CD28 (1ug/ml), recombinant mouse IL-2 (2ng/ml), recombinant mouse IL-7 (5ng/ml), and recombinant mouse IL-15 (50ng/ml). Cells were analyzed for their activation status through flow cytometric analysis of anti-CD44 PE and anti-CD69 PE. 5 × 10⁶ cells were adoptively transferred into Rag1^{-/-} mice. After 7 days, these mice were sacrificed, spleens were harvested, and prepared similarly to methods described previously. CD8 T cells were purified using mouse CD8 T cell kit from Miltenyi. Cells were then resuspended to a final concentration of 1×10^6 cells/ml in cRPMI supplemented with IL-2, IL-2, monensin, and anti-CD107a. Cells were then cocultured with either E0771 cells or E0771-mChOVA cells for 2 hours. After 2 hours, samples were stained with anti-CD3 PE/Cy7 and anti-CD8 APC, and analyzed via flow cytometry.

Degranulation screen and validation

Experiments were first optimized by pulsing E0771 cells with varying concentrations of SIINFEKL peptide for 4 hours at 37°C, and subsequently stained with the anti-mouse SIINFEKL: H-2K^b antibody and analyzed on flow cytometry. The dose of 1ng/ml was chosen as it represents the maximum concentration tested without being detected by anti-SIINFEKL: H-2Kb. Naive OT-I; Cas9 CD8 T cells were isolated and transduced with MKO lentiviral library as previously described. Infected OT-I; Cas9 CD8 T cells were incubated on plates pretreated with 5 μg/ml anti-CD3ε in cRPMI supplemented with 2ng/ml IL-2 + 2ng/ml IL-12p70 + 1 μg/ml anti-CD28 for 6 days. 12 hours before assay, infected OT-I; Cas9 CD8 T cells were incubated on untreated plates in the presence of 2ng/ml IL-2 + 2ng/ml IL-12p70 to rest cells. On day 6, 12 hours before assay, 1 × 10⁷ E0771 cells were also plated on 10cm plate in D10 media (DMEM + 10% FBS + 100U Pen/Strep). The following day, E0771 cells were incubated with warm D10 media supplemented with either 0 or 1ng/ml SIINFEKL peptide for 4 hours. Meanwhile, infected OT-I; Cas9 CD8 T cells were resuspended to a final concentration 1 × 10⁶ cells/ml with cRPMI + 2nM monensin + anti-CD107a PE antibody (1:400), and added to E0771 cells at a T cell: seeding cancer cell ratio = 1:1. Cells were co-incubated at 37°C for 2 hours. Cells were then stained with anti-CD8 APC for 30 minutes on ice, and cells were sorted via BD FACSAria. 1×10^7 T cells were analyzed, and the top 5% CD107a⁺ cells were sorted, and subjected to genomic DNA extraction, CRISPR library readout, and screen data analysis (described below). A total of three biological replicates were performed. In Dhx37 single gene perturbation degranulation assay, both T cell and E0771 cells were scaled down accordingly, while maintaining T cell: seeding cancer cell ratio = 1:1. Briefly, cells are infected and cultured at 37°C for 7 days. On day 6, T cells are rested overnight and E0771 cells are plated in 24-well plate at 5×10^5 cells/ml/well. E0771 cells are then incubated with media, 1ng/ml SIINFEKL or 10ng/ml SIINFEKL for 4 hours in D10 media. 5 x 10⁵ OT-l; Cas9 CD8 T cells were added to the pulsed E0771 cells. During co-culture, IL-2, IL-12, anti-CD107a-PE (1:400), and monensin was added. After 2 hours of incubation, cells were isolated, stained, and analyzed by flow cytometry.

Genomic DNA extraction

For gDNA extraction, three methods were used. Method 1: for samples with a total number of less than or equal to 1 × 10⁵ cells, 100 μL of QuickExtract solution (Epicenter) was directly added to cells and incubated at 65°C for 30 to 60 minutes until the cell pellets were completely dissolved followed by a 5 minute incubation at 95°C to denature proteins found in QuickExtract. Method 2: for cellular samples with a total number of 1×10^5 to 2×10^6 cells, or tissue samples from mouse lymph nodes, samples were subjected to QIAamp Fast DNA Tissue Kit (QIAGEN) following the manufacturer's protocol. Method 3: for cellular samples with a total number greater than 2×10^6 cells, or tissue samples from mouse organs or tumor samples, a custom Puregene protocol was used. Briefly, 50-200 mg of frozen ground tissue were resuspended in 6 mL of Lysis Buffer (50 mM Tris, 50 mM EDTA, 1% SDS, pH 8) in a 15 mL conical tube, and 30 µL of 20 mg/ml Proteinase K (QIAGEN) were added to the tissue/cell sample and incubated at 55°C overnight. The next day, 30 μL of 10 mg/ml RNase A (QIAGEN) was added to the lysed sample, which was then inverted 25 times and incubated at 37° C for 30 minutes. Samples were cooled on ice before addition of 2 mL of pre-chilled 7.5M ammonium acetate (Sigma) to precipitate proteins. The samples were vortexed at high speed for 20 s and then centrifuged at $\geq 4,000 \times g$ for 10 minutes. Then, a tight pellet was visible in each tube and the supernatant was carefully decanted into a new 15 mL conical tube. Then 6 mL 100% isopropanol was added to the tube, inverted 50 times and centrifuged at $\geq 4,000 \times g$ for 10 minutes. Genomic DNA was visible as a small white pellet in each tube. The supernatant was discarded, 6 mL of freshly prepared 70% ethanol was added, the tube was inverted 10 times, and then centrifuged at $\geq 4,000 \times g$ for 1 minute. The supernatant was discarded by pouring; the tube was briefly spun, and remaining ethanol was removed using a P200 pipette. After air-drying for 10-30 minutes, the DNA changed appearance from a milky white pellet to slightly translucent. Then, 500 μ L of ddH $_2$ O was added, the tube was incubated at 65°C for 1 hour and at room temperature overnight to fully resuspend the DNA. The next day, the gDNA samples were vortexed briefly. The gDNA concentration was measured using a Nanodrop (Thermo Scientific).

sgRNA library readout by deep sequencing

The sgRNA library readout was performed using a two-steps PCR strategy, where the first PCR includes enough genomic DNA to preserve full library complexity and the second PCR adds appropriate sequencing adapters to the products from the first PCR. For PCR#1, a region containing sgRNA cassette was amplified using primers specific to the T cell CRISPR vector:

Forward CCCGAGGGGACCCAGAGAG Reverse CAATTCCCACTCCTTTCAAGAC

PCR was performed using Phusion Flash High Fidelity Master Mix (PF) or DreamTaq Green PCR Master Mix (DT) (ThermoFisher). For reactions using PF, in PCR#1, the thermocycling parameters were: 98° C for 2min, 18-24 cycles of (98° C for 1 s, 62° C for 5 s, 72° C for 30 s), and 72° C for 2 minutes. For reactions using DT, the thermocycling parameters were adjusted according to manufacturer's protocol. In each PCR#1 reaction, we used 3 μ g of total gDNA. For each sample, the appropriate number of PCR#1 reactions was used to capture the full representation of the screen. For example, at \sim 200x coverage of our 129,209 MKO sgRNA library, gDNA from 2.5×10^7 cells was used. Assuming 6.6 pg of gDNA per cell, \sim 160 μ g of gDNA was used per sample, in approximately 50 PCR#1 reactions (with \sim 3 μ g of gDNA per reaction).

PCR#1 products for each biological sample were pooled and used for amplification with barcoded second PCR primers (Supplementary Tables). For each sample, we performed at least 4 PCR#2 reactions using 2 μL of the pooled PCR#1 product per PCR#2 reaction. Second PCR products were pooled and then normalized for each biological sample before combining uniquely barcoded separate biological samples. The pooled product was then gel purified from a 2% E-gel EX (Life Technologies) using the QiaQuick kit (QIAGEN). The purified pooled library was then quantified with a gel-based method using the Low-Range Quantitative Ladder Life Technologies, dsDNA High-Sensitivity Qubit (Life Technologies), BioAnalyzer (Agilent) and/or qPCR. Diluted libraries with 5%–20% PhiX were sequenced with MiSeq, HiSeq 2500 or HiSeq 4000 systems (Illumina).

T cell adoptive transfer anti-tumor efficacy

sgRNAs targeting individual genes were cloned into the T celal CRISPR vector. Two independent sgRNAs targeting each gene (e.g., Dhx37) were used. Virus prep and T cell infection were performed as described above. 5×10^6 E0771-mCh-OVA cells were injected either subcutaneously or into the intramammary fat pad of $Rag1^{-/-}$ mice. 7 days post-transplantation, freshly isolated naive OT-I; Cas9 CD8 T cells were plated on plates pretreated with $5\,\mu g/ml$ anti-CD3 ϵ in cRPMI supplemented with 2ng/ml IL-2 + 2.5 ng/mL IL-7 + 50 ng/mL IL-15 + 1 $\mu g/ml$ anti-CD28, infected with these sgRNA-containing lentiviruses (at MOI of \sim = 1) as described above, and cultured for 3 days. 10 days post-transplantation, 5×10^6 virally infected T cells were intravenously injected in tumor-bearing $Rag1^{-/-}$ mice (T cell: initial cancer cell ratio = 1:1). PBS and empty vector infected T cells were used as adoptive transfer controls. Tumor sizes were measured by caliper once to twice per week. 6 weeks after adoptive transfer, tumors were dissected, and samples were subjected to molecular, cellular, histology analysis, or single-cell RNA-seq. For statistical comparison of tumor growth curves, multiple t tests were performed (Benjamini, Krieger and Yekutieli FDR method) on each time point.

Tumor cell preparation and Tumor Infiltration Lymphocyte Isolation

Tumor bearing mice were euthanized at designated time points, and their tumors were collected and kept in ice cold 2% FBS. Tumors were minced into 1-3 mm size pieces using scalpel and then digested in 1 μ g/ml Collagenase IV for 30-60 min using Miltenyi GentleMACS Octo Dissociator. Tumor suspensions were filtered twice through 100μ m cell strainer, and again through 40μ m cell strainer to remove large bulk. For antigenic retention experiments, 10^8 cells were stained with 4ml of MACS Buffer and Live/Dead fixable near-IR dead cell stain from Invitrogen for 15 minutes on ice. Samples were washed twice with MACS Buffer. Tumor cell suspensions were filtered twice through 40μ m cell strainer before FACS data acquisition.

For tumor infiltration lymphocyte (TIL) isolation, tumor suspensions were carefully layered onto Ficoll-Paque media (GE Healthcare) and centrifuge at 400 g for 30 min to enrich lymphocytes at the bilayer interface. Cells at the interface were carefully collected, and washed twice with 2% FBS, counted, and stained with indicated antibodies for 30 minutes on ice. $CD3^+CD8^+$ TILs were then sorted on BD FACSAria. A total of 3×10^3 to 2×10^4 TILs were collected per tumor.

TIL single cell RNA-seq

TILs sorted from freshly isolated tumors were subjected to single-cell RNA-seq library prep. We followed a protocol by 10x Genomics to perform single cell RNA-seq (scRNA-seq). In brief, Single Cell Master Mix was prepared fresh containing RT reagent mix, RT primer, additive A, and RT enzyme mix. A Single Cell 3' Chip was placed in a 10x Chip Holder. 50% glycerol solution to each unused well accordingly, TIL solution at ~100 cell / μl was added together with the master mix. The Single Cell 3' Gel Bead Strip was placed into a 10x Vortex Adaptor and vortex for 30 s. Then, Single Cell 3' Gel Bead suspension and Partitioning Oil were dispensed into the bottom of the wells in the specified rows. The fully loaded chip was then inserted into Chromium Controller to generate emulsion. The emulsion was then transferred to a 96-well PCR plate for GEM-RT reaction, RT clean up, cDNA amplification, cDNA clean-up, quantification and QC, and subjected to Illumina library construction. In library construction, clean input cDNA was then subjected to fragmentation, end repair & A-tailing. After that, double sided size selection was performed using SPRI Select, followed by adaptor ligation, clean up, and sample indexing PCR, pooling and PCR cleanup, resulting a single-cell RNA-seq library. Enzymatic Fragmentation and Size Selection were used to optimize the cDNA amplicon size prior to library construction per manufacturer's protocols. R1 (read 1 primer sequence) are added to the molecules during GEM incubation. P5, P7, a sample index and R2 (read 2 primer sequence) are added during library construction via end repair, A-tailing, adaptor ligation and PCR. The Single Cell 3' Protocol produces Illumina-ready sequencing libraries contain the P5 and P7 primers used in Illumina bridge amplification. This final library was then QC'ed and quantified using BioAnalyzer, and loaded on a Hiseq 2500 RapidRun for standard Illumina paired-end sequencing, where Barcode and 10 bp randomer (UMI) is encoded in Read 1, while Read 2 is used to sequence the cDNA fragment. Sample index sequences are incorporated as the i7 index read.

AAV T cell knockout vector

An adeno-associated virus (AAV) CRISPR T cell knockout vector, pAAV-U6-sgBbsI-EFS-Thy1.1-PolyA, was generated by subcloning Thy1.1 and sgRNA expression cassette into an AAV vector via Gibson Assembly (NEB). Vector expresses an sgRNA with a nontargeting spacer. For individual gene targeting, the AAV knockout vector was digested with Bbsl. Oligonuleotides encoding specific sgRNAs were ligated into the digested AAV knockout vector using T4 ligase (NEB).

AAV virus production

Various AAV knockout plasmids used in this study as described in the figures and results were subjected to AAV production and chemical purification. Briefly, HEK293FT cells (ThermoFisher) were transiently transfected with transfer (AAV-plasmids), serotype (AAV9 or AAV6) and packaging (pDF6) plasmids using polyethyleneimine (PEI). Approximately 72 h post-transfection, cells were dislodged and transferred to a conical tube in sterile PBS. 1/10 volume of pure chloroform was added to the mixture and incubated at 37°C for 1 h. NaCl was added to a final concentration of 1 M. The mixture was subsequently shaken until dissolved and then pelleted at 20,000 gx at 4°C for 15 min. The chloroform layer was discarded while the aqueous layer was transferred to another tube. PEG8000 was added to 10% (w/v) and shaken until dissolved. The mixture was incubated at 4°C for 1 h and spun at 20,000 gx at 4°C for 15 min. After the supernatant was discarded, the pellet was resuspended in DPBS + MgCl₂, treated with benzonase (Sigma), and then incubated at 37°C for 30 min. Chloroform (1:1 volume) was then added, shaken and spun down at 12,000 g at 4°C for 15 min. The aqueous layer was isolated and passed through a 100-kDa MWCO (Millipore). The concentrated solution was washed with PBS and the filtration process was repeated. Virus was titered by qPCR using custom Taqman assays (ThermoFisher).

Immunological characterization

For immunological characterization, naive OT-I; Cas9 CD8 T cells were isolated and prepared similarly to methods described previously. At the start of culture, AAV-Vector or AAV-Dhx37 was added to the start of culture. Cells were incubated for 6 days. On the 6th day, cells were rested overnight in the presence of IL-2 (2ng/ml) and IL-7 (5ng/ml) for 12 hours. The next day cells were subjected to both functional (degranulation and IFN_γ production) and phenotypic analysis (surface and endosomal staining).

Determining gene editing efficiency

Cutting efficiency was analyzed from cells cultured for 5 days in the presence of AAV. DNA from cells was extracted by incubating cells with Epicenter QuickExtract for 65°C for 30 minutes, and subsequently incubated at 98°C for 5 minutes. gDNA from samples was amplified by PCR. Purified PCR product was subjected to both T7 endonuclease surveyor (NEB) and Nextera sequencing according to Illumina protocol. A list of oligos for T cell gene editing and determination of efficiency can be found at Table S7.

Western Blot of Dhx37 knockout murine T cells

Pellets from 5 x 10⁶ cells AAV-Vector or AAV-Dhx37 treated OT-I; Cas9 CD8 T cells were weighed. 10x volume of Invitrogen Lysis Buffer was added to cell pellets. Samples were incubated on ice for at least 10 minutes and then subjected to BCA analysis. 20ug of protein was used for western blot analysis. Dhx37 was analyzed using antibody purchased from Novus Biologicals. Beta-actin was analyzed using antibody purchased from Invitrogen.

Co-immunoprecipitation

Human PB CD4 $^+$, CD8 $^+$ T cells, as well as patient TILs were analyzed for DHX37 protein expression by western blot with a rabbit polyclonal antibody (Novus, NBP2-13922). For immunoprecipitation studies, 10^8 human CD8 T cells were isolated and wash twice with ice cold PBS. The cell pellet was subsequently weighed and 10x volume of Invitrogen Lysis Buffer supplemented with protease inhibitor cocktail (Roche) was added to cell pellets. Samples were incubated on ice for at least 10 minutes and then subjected to BCA analysis. 1-2mg of protein was used for immunoprecipitation studies, 10ug of anti-human DHX37 (Bethyl Laboratories), anti-human PDDC11 (Bethyl Laboratories), or IgG isotype control (Cell Signaling), and volume was resuspended to a final concentration of 400ul Invitrogen Lysis Buffer. Samples were tumbled overnight at 4° C for at least 12 hours. Immunocomplexes were pulled down using Protein A dynabeads from Invitrogen for 6 hours at 4° C. Afterward, beads were washed 3 times with ice cold PBS and resuspended with 50ul of NuPAGE LDS sample buffer with β-ME. Samples were heated at 70° C for 10 minutes. Beads were removed by magnet and eluent transferred to fresh tube. Samples were then heated at 95° C for 3 minutes and placed on ice for 1 minute before immunoblotting. For proteins > 200kDa, $3\sim8$ %Tris-Acetate gels were used for gel electrophoresis. For smaller proteins, $4\sim20$ % Tris-Glycine gels were used. Proteins were then transferred to $0.45~\mu m$ PVDF membrane overnight. Membranes were blocked with 5% BSA, stained with primary antibody overnight at 4° C, and stained with VeriBlot for IP secondary antibody (Abcam) for 1 hour at room temperature. Proteins were detected using Pierce ECL substrate.

qPCR and bulk mRNA-seq

4 x 10⁶ cells/sample were washed twice with ice cold PBS. Cells were then treated with TRIzol for 5 minutes at RT. RNA was purified via PureLink RNA mini kit from Invitrogen. cDNA for qPCR was generated using SuperScript Reverse Transcriptase IV users' protocol. In brief, at least 500ng of RNA was used per reaction. Random hexamers were used for cDNA generation. Afterward, cDNA was subjected to qPCR using TaqMan Real-Time PCR Master Mixes. The following probes and their sequences are provided: *Actb* (Mm02619580_g1), *Dhx37* (Mm01202964_g1), *Gapdh* (Mm99999915_g1), *Gzmc* (Mm01313651_m1), *Gzmd* (Mm01722569_g1), *Havcr2* (Mm00454540_m1), *Il6ra* (Mm00439653_m1), *Serpinb9b* (Mm00488405_m1), *Sytl2* (Mm01317927_m1). Samples were processed using Applied Bioscience StepOnePlus Real Time machine and relative mRNA expression was normalized to *Actb* or *Gapdh* controls. Quadruplicate total RNA samples from each group were subjected to mRNA-seq library prep. TruSeq strand-specific kit (Illumina) was used for mRNA-seq library preparation and sequenced using Illumina HiSeq 4000 or Novaseq platforms.

Endosomal staining

Endosomal staining was performed using BD Cytofix/Cytoperm staining kit and protocol. Prior to endosomal staining, cells were stained with surface markers for 30 minutes on ice. Afterward, cells were fixed with BD Cytofix buffer for 10 minutes on ice. Cells were washed with Cytoperm/Wash Buffer. Cells were then subjected to FcBlock using anti-CD16/32 (1:250) in Cytoperm/Wash Buffer. Samples were subsequently stained with endosomal stain for 30 minutes on ice, washed twice with Cytoperm/Wash Buffer, and finally resuspended in MACS Buffer prior to FACS analysis.

De novo IFN γ production

AAV-treated OT-I;Cas9 CD8 T cells were incubated for 6 days to allow for sufficient cutting and protein knock down. On day 6, cells were rested for at least 12 hours in cRPMI supplemented with IL-2 (2ng/ml) and IL-7 (5ng/ml). On day of experiment, 96-well plates were coated with PBS or 1ug/ml anti-CD3 for 1 hour at 37°C. (It is important to note that the outside wells were not used in the assay to account for potential edge effects. To avoid this issue, these wells were filled with PBS.) Wells were washed twice with cRPMI. Cells were resuspended to a final concentration of 10⁶ cells/ml of cRPMI supplemented with IL-2, IL-12, anti-CD28, and Brefeldin A. 100ul of cell suspension was added to each well and incubated for 4 hours at 37°C. Cells were then stained for anti-CD3 PE/Cy7 and anti-CD8 FITC. Cells were then fixed with BD Cytofix/Cytoperm kit as mentioned above and stained intracellularly with anti-IFN_Y APC. Samples were analyzed via FACS.

Co-culture killing

E0771 cells or E0771-mCh-OVA cells were infected with lentivirus bearing Luciferase transgene. Luciferase-positive cells were selected under hygromycin selection. Luciferase-positive cells were plated at a 1:1 ratio with AAV-treated OT-I;Cas9 CD8 T cells at a density of 5 × 10⁵ cells/ml in cRPMI supplemented with IL-2 and IL-12. For controls E0771-luciferase or E0771-mCh-OVA-luciferase cells were plated in the absence of T cells. After 24 hours, 150ug/ml luciferin was added to each well. Luciferase activity was measured using GE Typhoon 9500. Samples were normalized to E0771-luciferase or E0771-mCh-OVA-luciferase controls.

Human T cell culture and manipulation

Human CD8 T cells were purified from human PBMCs using Human Naive CD8 T cell kit. Human CD8 T cells were cultured at a concentration of 1-2 × 10⁶ cells/ml in X-VIVO 15 media supplemented with 5% Human AB serum + IL-2 (100U/ml). Every 14 days, cells were stimulated with anti-CD3/CD28 dynabeads for 48 hours and removed using magnet.

DHX37 western blot in human T cells

Human PB CD4⁺, CD8⁺ T cells, as well as patient TILs were analyzed for DHX37 protein expression by western blot with a rabbit polyclonal antibody (Novus, NBP2-13922).

GFP-DHX37 tagging

The lentiviral DHX37 overexpression construct was further engineered by Gibson Assembly to insert either an N-terminal or C-terminal EGFP fusion. GGSG linker was inserted between DHX37 and EGFP to prevent potential steric hinderance. Plasmids were used to transfect or transduce cells to overexpress DHX37 protein.

Immunofluorescence and microscopy

Cells were fixed by 4% PFA for 5 minutes, washed with 0.1% Triton PBS, and stained. Primary antibodies used: chicken anti-GFP (1:500, Abcam ab290), rabbit anti-nucleolin (1:500, Abcam ab22758), and rabbit anti-Dhx37 (1:200, NBP2-13922). Fluorescent images were captured on a Leica inverted microscope.

Gene editing in human CD8 T cells

Human CD8 T cells were treated according to methods recently developed in the lab (Dai et al., 2019 Mar;). Briefly, $2-3 \times 10^6$ cells were washed two times with ice cold PBS. Cells were then resuspended with 100ul of Buffer R from Neon Electroporation Kit.

For Cas9 RNP based gene editing, T cells were prepared at a density of 2×10^5 cells per $10 \, \mu L$ tip reaction in electroporation Buffer R (Neon Transfection System Kits). RNPs were produced by complexing a two-component gRNA to Cas9. In brief, Cas9 sgRNA was designed to target *DHX37* (sg4-TCTTTAAGCCTCCACCGGAG) using Benchling. Cas9 crRNAs and tracrRNAs were chemically synthesized (Dharmacon or IDT), and suspended in nuclease-free IDTE buffer at a concentration of $160 \, \mu M$. The crRNA and tracrRNA were mixed at 1:1 ratio and annealed as a guide RNA in Nuclease-Free IDTE buffer at $95 \, ^{\circ}$ C for $5 \, ^{\circ}$ min and $37 \, ^{\circ}$ C for $10 \, ^{\circ}$ min. RNPs were formed by the addition of SpCas9 nuclease (Dharmacon, IDT) with the $80 \, \mu M$ gRNA (1:2 Cas9 to sgRNA molar ratio) at benchtop for $15 \, ^{\circ}$ min. RNPs were electroporated immediately after complexing at program $24 \, (1,600V, 10 \, ^{\circ}$ ms and three pulses). After electroporation, the cells were transferred into $1 \, ^{\circ}$ mL of pre-warmed X-VIVO media (without antibiotics) immediately.

For Cas9 mRNA T cell electroporation, Cas9 mRNA T cell electroporation was similar with Cas9 RNP T cell electroporation. After T cells activation, T cells were mixed with 1 μ g Cas9 mRNA (TriLink) and 1 μ L annealed DHX37 sg4 (80 μ M) per 10 μ L tip reaction in electroporation Buffer R (Neon Transfection System Kits). Electric shocked were performed at program 24 (1,600V, 10ms and three pulses). After electroporation, the cells were transferred into 1mL of pre-warmed X-VIVO media (without antibiotics) immediately.

For Cpf1 gene editing, 5ug of Cpf1 mRNA and 5ug of crDHX37 was added to cell suspension. Cell mixture was pipetted up and down several times. The mRNA/crRNA complex were electroporated into human CD8 T cells using 3 pulses of 1600V for a duration of 10ms each pulse. After 5 days in culture, genomic DNA isolated per protocols established above. Cutting efficiency was determined by T7E1 surveyor and Nextera.

QUANTIFICATION AND STATISTICAL ANALYSIS

Statistics summary

Statistical analyses and software used are described in this section in detail. All of the statistical details of experiments can be found in the figures, figure legends, results, supplemental tables, and/or Mendeley Data (see also: DATA AND CODE AVAILABILITY), including the statistical tests used, exact value of n, what n represents (e.g., number of animals, number of experimental replicates), definition of center, and dispersion and precision measures (e.g., mean and SEM).

Demultiplexing and read preprocessing

Raw single-end fastq read files were filtered and demultiplexed using Cutadapt (Martin, 2011). To remove extra sequences down-stream (i.e., 3' end) of the sgRNA spacer sequences, we used the following settings: cutadapt-discard-untrimmed -a GTTTTAGAGC TAGAAATGGC. As the forward PCR primers used to readout sgRNA representation were designed to have a variety of barcodes to facilitate multiplexed sequencing, we then demultiplexed these filtered reads with the following settings: cutadapt -g file:fbc.fasta-no-trim, where fbc.fasta contained the 12 possible barcode sequences within the forward primers. Finally, to remove extra sequences upstream (i.e., 5' end) of the sgRNA spacers, we used the following settings: cutadapt-discard-untrimmed -g GTGGAAAGG ACGAAACACCG. Through this procedure, the raw fastq read files could be pared down to the 20bp sgRNA spacer sequences.

Mapping of spacers and quantitation of sgRNAs

Having extracted the 20 bp sgRNA spacer sequences from each demulitplexed sample, we then mapped the sgRNA spacers to the MKO library (Table S1). To do so, we first generated a bowtie index of either sgRNA library using the bowtie-build command in Bowtie 1.1.2 (Langmead et al., 2009). Using these bowtie indexes, we mapped the filtered fastq read files using the following settings: bowtie -v 1-suppress 4,5,6,7-chunkmbs 2000 -best. Using the resultant mapping output, we quantitated the number of reads that had mapped to each sgRNA within the library. To generate sgRNA representation bar plots, we set a detection threshold of 1 read, and counted the number of unique sgRNAs present in each sample.

Normalization and summary analysis of sgRNAs

We normalized the number of reads in each sample by converting raw sgRNA counts to reads per million (rpm). The rpm values were then subject to log₂ transformation for certain analyses. To generate correlation heatmaps, we used the *NMF* R package (Gaujoux and Seoighe, 2010) and calculated the Pearson correlations between individual samples using log₂ rpm counts. To calculate the cumulative distribution function for each sample group, we first averaged the normalized sgRNA counts across all samples within a given group. We then used the ecdfplot function in the *latticeExtra* R package to generate empirical cumulative distribution plots.

Enrichment analysis of sgRNAs

We used three criteria to identify the top candidate genes: 1) if an sgRNA comprised $\geq 2\%$ of the total reads in at least one organ sample; 2) if an sgRNA was deemed statistically significantly enriched in $\geq 20\%$ of all organ samples using a false-discovery rate (FDR) threshold of 0.5% based on the abundances of all non-targeting controls; or 3) if ≥ 2 independent sgRNAs targeting the same gene were each found to be statistically significant at FDR < 0.5% in at least one sample each. For the first and second criteria, individual sgRNA hits were collapsed to genes to facilitate comparisons with the hits from the third criteria. Due to the number of T cells trafficking *in vivo* in mice, depletion screen analysis was not performed. To enhance the statistical power, the infiltration screen data from the orthotopic (fat pad) model and the subcutaneous model were combined for hit identification.

Heatmap sgRNA library representation

Heatmaps of the top enriched sgRNAs were generated using the *aheatmap* function with default setting (*NMF* R package). Only sgRNAs with a log_2 rpm ≥ 1 were included for visualization in the heatmaps.

Overlap and significance of enriched sgRNAs

To generate Venn diagrams of overlapping enriched sgRNAs or genes, we considered all sgRNAs that were found to be significantly across different statistical calling algorithms, different T cells, or different experiments.

RIGER and MAGeCK analyses

RNAi Gene Enrichment Ranking (RIGER) (Luo et al., 2008) and the Model-based Analysis of Genome-wide CRISPR/Cas9 Knockout (MAGeCK) (Li et al., 2014) algorithms was used as independent methods to quantify enrichment of candidate genes for both the infiltration and survival screens.

For RIGER analysis, read count tables were used to calculate log fold changes for tumor versus cell samples in order to score and rank sgRNAs, with ties in rank broken by random order. This data was then used as input to a Java-based implementation of RIGER (https://github.com/broadinstitute/rigerj) in order to generate p values and gene rankings based on consistent enrichment across multiple sgRNAs for identification of candidate genes (Shalem et al., 2014). Both the second highest-ranking sgRNA and the weighted sum scoring methods were used for computation of gene rankings.

For MAGeCK analysis, read count tables were used as inputs to a command-line-based tool (https://sourceforge.net/p/mageck/wiki/Home/) with the treatment group defined as the tumor samples and the control group defined as the cell pellet samples, with a list of non-targeting control sgRNAs provided for normalization and generation of the null distribution of RRA. Native MAGeCK plotting functions were used for visualization of RRA score and p value distributions and individual sgRNA read counts of selected genes.

Gene ontology and pathway enrichment analysis

Various gene sets were used for gene ontology and pathway enrichment analysis using DAVID functional annotation analysis (Huang et al., 2009). For sgRNA set, sgRNAs were converted to their target genes and then the resultant genes were used for analysis.

scRNA-seq data processing

TIL scRNA-seq fastq data was pre-processed using established and custom pipelines. Briefly, raw Illumina data files were subjected to Cell Ranger, which used cellranger mkfastq to wrap Illumina's bcl2fastq to correctly demultiplex Chromium-prepared sequencing samples and to convert barcode and read data to FASTQ files. Then, cellranger count was used to take FASTQ files and performs alignment to mouse genome (mm10), filtering, and UMI counting. Raw sequencing output was first preprocessed by Cell Ranger 1.3 (10x Genomics) (Zheng et al., 2017b) using cellranger mkfastq, count, and aggr (no normalization mode). Cells passed the initial quality control metrics imposed by the Cell Ranger pipeline were further filtered using a variety of criteria (Lun et al., 2016): 1) We excluded all cells with a total library count (i.e., # of UMIs) that was ≥ 4 standard deviations below the mean; 2) We excluded all cells with library diversity (i.e., # of detected genes/features) that was ≥ 4 standard deviations below the mean; and 3) We excluded all cells in which mitochondrial genes disproportionately comprised the total % of the library (≥ 4 standard deviations above the mean). After applying these 3 filters, we retained a final set of cells for further analysis. We additionally filtered the 27,998 genes/features using a flat cutoff metric: genes with low variance were excluded. Finally, the data was normalized by library size using the *scran* R package (Lun et al., 2016).

scRNA-seq analysis

Using the final normalized and processed dataset (described above), we performed differential expression analysis using the edgeR R package (Robinson et al., 2010). In brief, edgeR first estimates the negative binomial dispersion parameter to model the variance

between cells from the same treatment group. A generalized linear model is then fitted to determine differentially expressed genes between treatment conditions. Multiple hypothesis correction was performed by the Benjamini-Hochberg method. Significantly differentially expressed genes were defined as having a Benjamini-Hochberg adjusted p < 0.05, with upregulated genes having a positive log fold change and downregulated genes having a negative log fold change. Volcano plots were generated using *edgeR* output statistics. Gene ontology enrichment analyses on differentially expressed genes were performed using the PANTHER classification system (Mi et al., 2013). The statistical overrepresentation test was used to identify enriched GO (biological process) categories among the differentially expressed genes. Bonferroni multiple hypothesis correction was performed. To generate an overall view of the top differentially expressed genes, we scaled each row of the dataset (i.e., by gene) to obtain z-scores. Heatmaps were generated using the *NMF* R package (Gaujoux and Seoighe, 2010).

mRNA-seq processing and analysis

Bulk mRNA-seq fastq files were processed by kallisto (Bray et al., 2016), with the settings-rf-stranded –b 100. Differential expression analysis was then performed using Sleuth (Pimentel et al., 2017). Significant differentially expressed genes were defined by a Benjamini-Hochberg adjusted p value < 0.05. Differentially expressed genes were used for DAVID gene ontology analysis (Huang et al., 2009). The full dataset was also used for GSEA (Subramanian et al., 2005). Motif enrichment analysis in the promoters of differentially expressed genes was performed using (Zambelli et al., 2009), using a –450 to +50 bp interval and the TRANSFAC database.

Epigenetic analysis of Dhx37 in CD8 T cells

Normalized ATAC-seq, H3K4me3, H3K27ac, and H3K27me3 tracks were accessed from GSE111092 (Laumont et al., 2018), through the ctlcon package (https://bitbucket.org/robert_amezquita/ctlcon).

Analysis of human T cell (sc)RNA-seq data

Human T cell scRNA-seq data from liver cancer patients ((Zheng et al., 2017a), GSE98638) and melanoma patients undergoing immunotherapy ((Sade-Feldman et al., 2018), GSE120575.) For the GSE98638 dataset, cells were classified according to the original definitions: peripheral blood, tissue-resident, tumor-normal junction, and tumor-infiltrating T cells; CD3+/CD4+/CD25-T cells, CD3+/CD8+T cells, and CD3+/CD4+/CD25+T cells. For the GSE120575 dataset, the published designations of "exhausted" versus "non-exhausted" were utilized. For both datasets, normalized TPM expression values were utilized for downstream analysis. Differential expression of DHX37 in exhausted versus non-exhausted CD8 T cells was assessed by Mann Whitney test. For analysis of DHX37 expression in bulk RNA-seq data from human T cells, the DICE database was queried (Schmiedel et al., 2018) (https://dice-database.org/).

Analysis of CTL function in patient survival

We used the TIDE tool with DHX37 as the query gene (Jiang, 2018) (http://tide.dfci.harvard.edu/query/). Breast cancer datasets were evaluated for evidence of T cell dysfunction, stratified by bulk tumor expression levels of DHX37.

Blinding statement

Investigators were blinded for sequencing data analysis using metadata, but not blinded for tumor engraftment, adoptive transfer, organ and tumor dissection, or flow cytometry.

DATA AND CODE AVAILABILITY

All data for this study are included in this published article and its supplemental materials. Original/source non-genomics data and relevant statistics for figures in the paper are provided in an excel sheet and are in Mendeley Data (Reserved https://doi.org/10.17632/2ffxmf3d7k.1). The genomics data generated during this study are available at public repositories SRA and GEO with Accession numbers. All DNA-level genomic sequencing data are available via NCBI SRA: PRJNA549266. All transcriptomics data are available via GEO (which automatically makes SRA deposit): GSE132960, superseries composed of GSE132926 (bulk RNA) and GSE132959 (scRNA-seq). Custom scripts and additional data related to support the findings of this work will be available to the academic community upon reasonable request to the lead contact.

ADDITIONAL RESOURCES

Additional resources and related information can be found at Key Resources Table.

Supplemental Figures

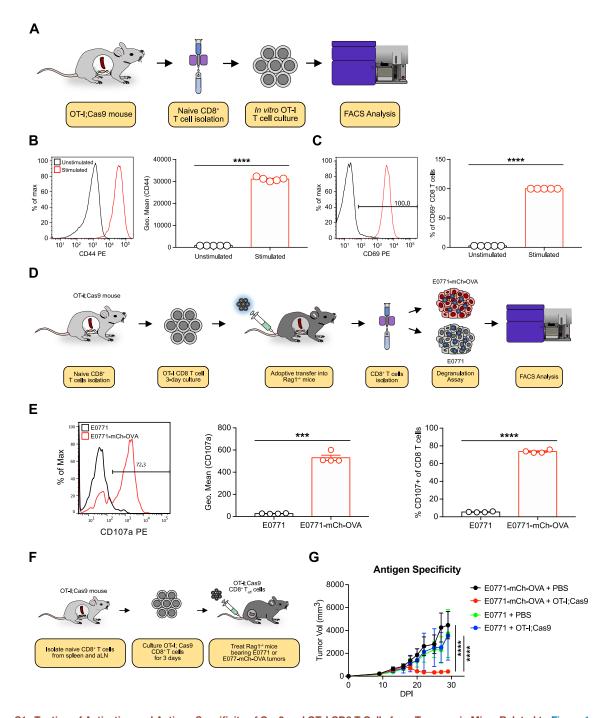


Figure S1. Testing of Activation and Antigen Specificity of Cas9 and OT-I CD8 T Cells from Transgenic Mice, Related to Figure 1

(A) Schematic of flow analysis experiment for T cell activation in OT-I;Cas9 CD8 T cells in culture.

(B and C) Flow cytometry analysis of activation markers of T cells, including CD44 (B) and CD69 (C). Left panel in each plot is a representative histogram of stimulated (red) and unstimulated (black) CD8 T cells. Right panel is a bar plot of quantification for each marker. * = p < 0.05, ** = p < 0.01, *** = p < 0.001, by unpaired two-sided t test.

(D) Schematic of flow analysis experiment for antigen-specificity testing in OT-I;Cas9 CD8 T cells in culture.

(E) Flow cytometry analysis of degranulation of T cells using CD107a. Left panel in each plot is a representative histogram of CD8 T cells co-cultured with antigenexpressing cancer cells E0771-mCh-OVA (red) and with parental control cancer cells E0771 (black). Middle panel is a bar plot of quantification for CD107a as geometric mean. Right panel is a bar plot of quantification for CD107a as percent cell positive. * = p < 0.05, ** = p < 0.01, *** = p < 0.001, by unpaired two-sided t test.

(F) Schematic of antigen-specificity testing for OT-I; Cas9 CD8 T cells in a tumor model in vivo.

⁽G) Tumor growth curves of E0771 (parental control line) or E0771-mChOVA (antigen-expressing line) in mice treated with PBS or adoptive transfer of OT-I;Cas9 CD8 T cells. Only the group of E0771-mChOVA tumors receiving adoptive transfer of OT-I;Cas9 CD8 T cells showed significant tumor regression, supporting antigen-specificity of OT-I;Cas9 CD8 T cells against the OVA antigen. * = p < 0.05, ** = p < 0.01, *** = p < 0.001, by two-way ANOVA.

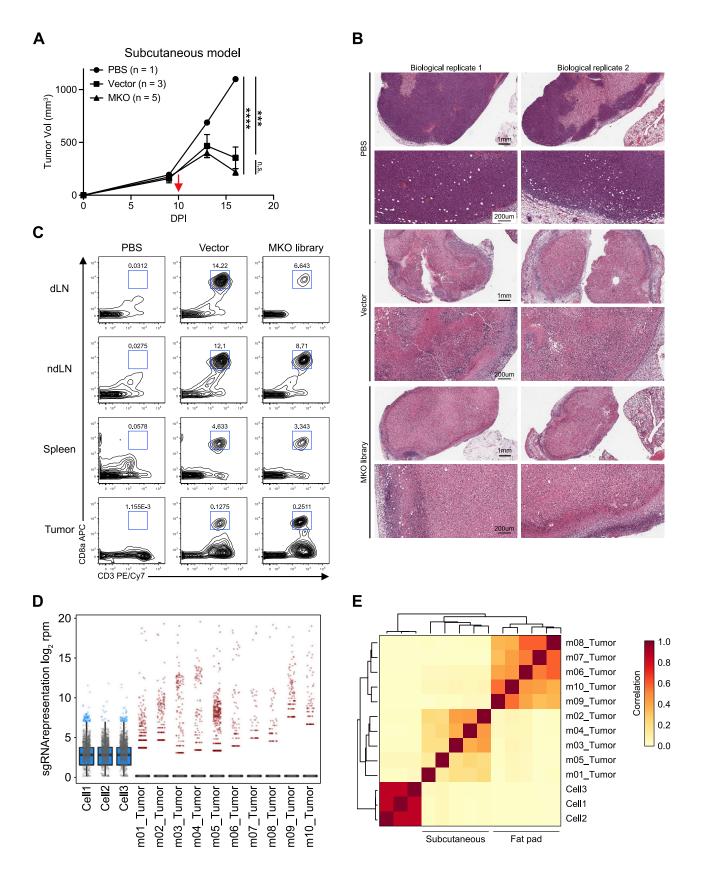




Figure S2. Additional Experiments and Analyses for Adoptive Transfer Tumor Infiltration Screen, Related to Figure 1

(A) Growth curve of subcutaneous tumors from transplanted E0771-mCh-OVA cells in $Rag1^{-/-}$ mice following different treatments. PBS control (n = 1), adoptive transfer of OT-I;Cas9 CD8 T cells infected with vector (n = 3), and adoptive transfer of OT-I;Cas9 CD8 T cells infected with MKO (n = 5). Red arrow indicates the time of adoptive transfer of MKO or vector transduced OT-I;Cas9 CD8 T cells. Error bars for certain data points were invisible because the errors were small. Data are shown as mean \pm s.e.m.. * = p < 0.05, ** = p < 0.01, *** = p < 0.001, by two-way ANOVA. Note: Detailed statistics related to this figure are provided in an excel table.

- (B) Full-slide and high-power histology sections stained by hematoxylin and eosin of tumors derived from E0771 cells expressing OVA antigen in Rag1^{-/-} mice after different treatment conditions. Top group: tumors in mice that were injected with PBS. Middle group: tumors in mice after adoptive transfer of vector-treated activated OT-I; Cas9 CD8 T cells. Bottom group: tumors in mice after adoptive transfer of MKO mutagenized activated OT-I; Cas9 CD8 T cells. In PBS group, tumors were devoid of lymphocytes and showed signatures of rapid proliferation and little cell death. In adoptive transfer groups, tumors were infiltrated by lymphocytes and showed signatures of cell death in large areas. Low magnification image scale bar: 1mm; high magnification image scale bar: 200 μm.
- (C) Representative FACS plots of adoptively transferred T cells in draining and non-draining LNs (dLN and ndLN, respectively), spleen, lung, and tumor (TILs) from E0771-mCh-OVA tumor-bearing Rag1^{-/-} mice. MKO is the genome-scale T cell CRISPR library. Numbers indicate average percentage of total cells. Top row: FACS plots from PBS-treated mice. Middle row: FACS plots from mice treated with vector-infected OT-I;Cas9 CD8 T cells. Bottom row: FACS plots from mice treated with MKO-infected OT-I;Cas9 CD8 T cells.
- (D) Box-dot plot of overall sgRNA library representation in all samples, including cellular libraries of infected OT-I; Cas9 CD8 T cells before injection (blue, n = 3), and tumors from multiple mice (red, n = 10 mice, 10 total tumors). sgRNA representation is depicted in terms of \log_2 rpm.
- (E) Correlation analysis of genome-scale CRISPR perturbation of OT-I; Cas9 CD8 tumor-infiltrating lymphocytes into $Rag1^{-/-}$ mice with E0771-mCh-OVA tumors. Heatmap of pairwise Pearson correlations of sgRNA library representation across 3 cell libraries prior to injection, and all samples in the tumor infiltration screen (n = 10 mice, 10 tumors). Correlations were calculated based on log_2 rpm values. E0771-mCh-OVA cells were transplanted subcutaneously for mice 1-5, and into the mammary fat pad for mice 6-10.

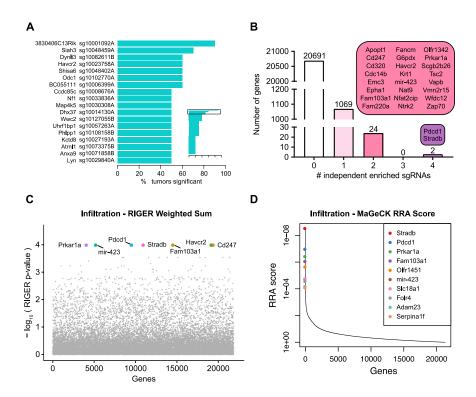


Figure S3. Additional Analyses for Adoptive Transfer Tumor Infiltration Screen, Related to Figure 1

(A) Waterfall plot of the top-ranked sgRNAs across all tumors (21 sgRNAs significantly enriched in \geq 50% of tumors, FDR < 0.5%). Inset, waterfall plot of all sgRNAs that were significantly enriched in \geq 20% of tumors.

(B) Bar plot of the number of genes with 0-4 independent sgRNAs that were significantly enriched in at least one organ sample (FDR < 0.5%). A total of 26 genes were found to have at least 2 independent sgRNAs enriched. *Pdcd1* and *Stradb* were each found to have 4 independently enriched sgRNAs.

- (C) Meta-analysis of infiltration screen using RIGER with Weighted Sum method.
- (D) Meta-analysis of infiltration screen using MAGeCK analysis of survival screen, highlighting several top-scoring genes ranked by RRA score.

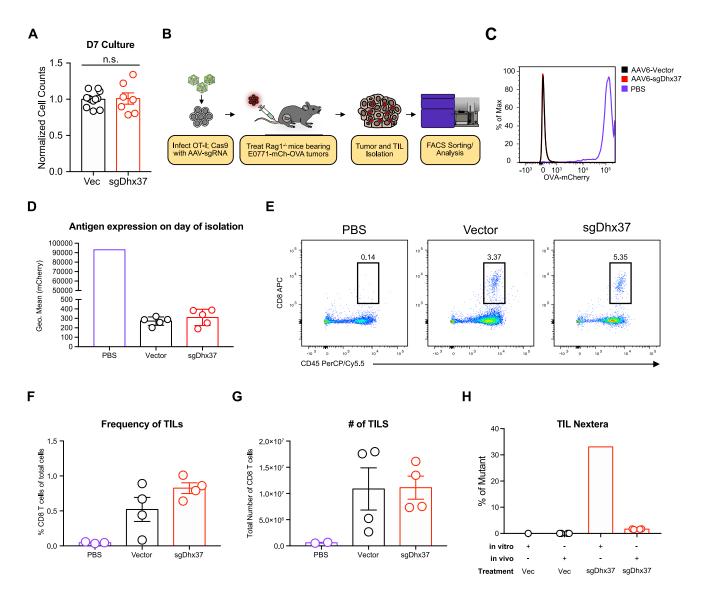


Figure S4. Analysis of Antigen Retention and T Cell Stimulation Ability of Tumor Cells after Adoptive Transfer, Related to Figures 2 and 4 (A) Normalized cell counts of AAV-Vector and AAV-sgDhx37 OT-I;Cas9 CD8 T cells after 7 days of cultures.

- (B) Schematic of flow analysis experiment for antigen retention and T cell stimulation ability of tumor cells after adoptive T cell transfer.
- (C and D) Flow cytometry plot of tumor cells gating on mCherry-OVA expression from tumors induced by E0771-mCh-OVA cancer cells treated with PBS, Vector or sgDhx37 OT-I;Cas9 CD8 T cells. (C) Representative histograms. (D) Bar plot quantification of (C).
- (E) Flow cytometry plot of CD8 TIL gating for PBS, AAV-Vector, and AAV-sgDhx37 OT-I;Cas9 CD8 T cell-treated mice.
- (F and G) Frequency and total numbers of CD8 TILs observed in $Rag1^{-/-}$ mice bearing E0771-mCh-OVA tumors treated with either PBS, AAV-Vector, and AAV-sgDhx37 OT-I;Cas9 CD8 T cells
- (H) Representative Illumina targeted amplicon sequencing of the *Dhx37* sgRNA target site in AAV-treated OT-I;Cas9 CD8 T cells at the time of injection and on the day of tumor sample collection.

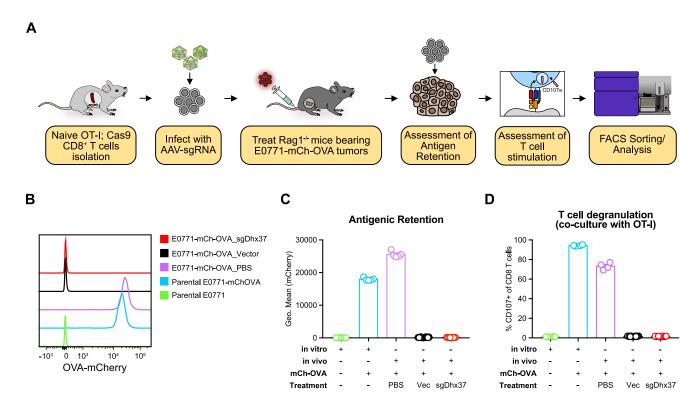


Figure S5. Antigenic Retention and OT-I CD8 T Cell Stimulation by Tumor Cells Ex Vivo, Related to Figures 2 and 4

(A) Schematic of flow analysis experiment for antigenic retention of cancer cells isolated from E0771-mCh-OVA tumor-bearing Rag1^{-/-} mice treated with either PBS, AAV-Vector, or AAV-sgDhx37 OT-I;Cas9 CD8 T cells. Tumors were isolate 50 days after orthotopic tumor injection and cultured for several days in order to acquire sufficient number of cells for *in vitro* assays.

- (B) Histogram representation of antigenic retention of cultured tumor cells on the day of experiment.
- (C) Quantification of (B).
- (D) Quantification of OT-I;Cas9 CD8 T cell degranulation when cocultured with indicated cancer cell.

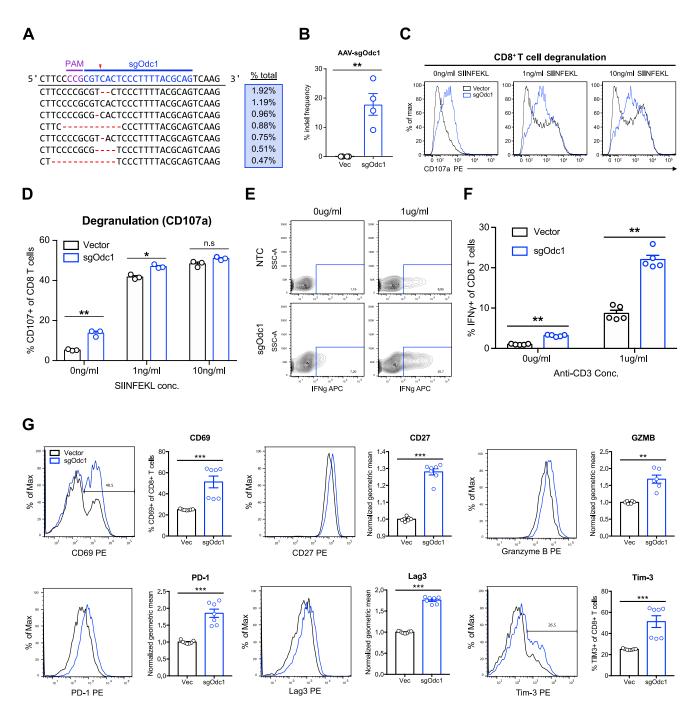


Figure S6. Immunological Characterization of Odc1 Knockout Mouse CD8 T Cells, Related to Figure 5

(A) Representative Illumina targeted amplicon sequencing of the sgRNA target site 5 days after infection with AAV-sgRNA targeting the gene of interest. Top most frequent variants were shown, with the associated variant frequencies in the box to the right. In these Nextera-Indel plots, PAM and sgRNA spacers were indicated above. Red arrows indicate predicted cleavage sites. Red dash lines indicate deletions. Red A/C/G/T where applicable indicate insertions.

- (B) Bona fide gene editing in primary murine CD8 T cells with an AAV-sgOdc1 measured by T7E1 surveyor assay. Data are shown as mean \pm s.e.m.. ** = p < 0.01, by unpaired two-sided Mann Whitney test.
- (C) CD107a degranulation assay for *Odc1* knockout CD8 T cells. Flow cytometry histograms were shown for each treatment group (Vector and sgOdc1) with 3 doses of SIINFEKL peptide pulsing of E0771 cancer cells (0, 1, and 10 ng/ml).
- (D) Quantification of data from (C) with independent experimental replicates. Data are shown as mean \pm s.e.m. ** = p < 0.01, by unpaired two-sided Mann Whitney test.
- (E) Flow cytometry analysis plot of IFN γ production of sgOdc1 and Vector CD8 T cells with or without anti-CD3 stimulation.

⁽F) Quantification bar plot of (E). Data are shown as mean \pm s.e.m.. ** = p < 0.01, by unpaired two-sided Mann Whitney test.

⁽G) Flow cytometry analysis of key immune markers of T cell function, including CD69, CD27, Granzyme B, PD-1, Lag3 and Tim-3. Left panel in each marker's analysis is a representative histogram of sgOdc1 (blue) and Vector (black) CD8 T cells. Right panel is a bar plot of quantification for each marker. Data are shown as mean \pm s.e.m.. *=p < 0.05, **=p < 0.01, ***=p < 0.001, by unpaired two-sided Mann Whitney test.

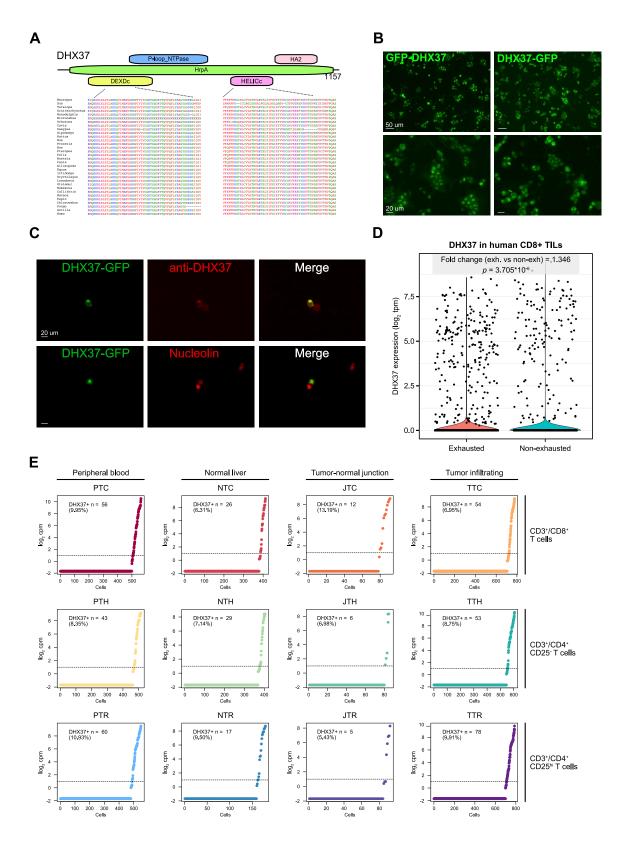


Figure S7. Additional Analysis of Human DHX37, Related to Figure 7

(A) Domain structure and conservation of DHX37 protein, which contains an HrpA domain that can be subdivided into 4 subdomains (DEXDc, P loop NTPase, HELICc and HA2). Shown below are representative alignments of regions in DEXDc and HELICc domains that are conserved between multiple mammalian species including mouse and human.

- (B) Localization of DHX37 in the nucleus, as revealed by microscopic images of GFP-DHX37 and DHX37-GFP fusion proteins, under 20x (top) and 40x (bottom) objectives. Scale bars, 50 μm for 20x, and 20 μm for 40x.
- (C) Co-Localization of DHX37 with nucleolus, as revealed by immunofluorescence of anti-DHX37, DHX37-GFP and Nucleolin staining, under 40x objectives. Scale bar, 20 µm.
- (D) DHX37 mRNA expression in human cancer patients' CD8 TILs. Exhausted TILs express higher level of DHX37 (p = 3.705e-16 by Mann-Whitney test).
- (E) Distribution of *DHX37* expression in human peripheral blood T cells, tissue-resident T cells, tumor-normal junction T cells, and tumor-infiltrating T cells in CD3⁺/CD4⁺/CD25⁻ helper T cells, CD3⁺/CD8 cytotoxic T cells, and CD3⁺/CD4⁺/CD25⁺ regulatory T cells. The number and total percentage of DHX37⁺ cells are noted on each plot.