Regular Article

TET2-mutant clonal hematopoiesis prevents T-cell exhaustion and suppresses cancer
metastasis

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ABSTRACT

Clonal hematopoiesis (CH) is widely regarded as a risk factor for age-associated disease, yet its influence on solid tumor progression remains poorly understood. Here, we uncover an unexpected tumor-suppressive role for *TET2*-mutant CH in solid tumors. Through an analysis of over 16,000 cancer patients, we demonstrate that *TET2* mutations are associated with a significant reduction in metastatic burden. Mechanistically, we show that *Tet2* deficiency prevents terminal exhaustion in CD8+ tumor-infiltrating lymphocytes (CD8+ TILs). This resistance is driven by DNA hypermethylation at regulatory elements of the transcription factor *Tox*, which prevents the induction of the terminal exhaustion program. Consequently, *Tet2*-deficient CD8+ TILs maintain stem-like, effector-competent states that mediate durable anti-tumor immunity and suppress metastatic outgrowth. Our findings establish *TET2*-mutant CH as a natural epigenetic constraint on T-cell exhaustion and uncover a mechanistically defined pathway with translational potential for enhancing anti-tumor immune responses.

INTRODUCTION

Metastasis represents the principal cause of cancer-related mortality. Clinical metastatic disease is present in 4% to 65% of patients with solid malignancies at the time of cancer diagnosis[1]. Metastasis occurs through a complex and multi-step cascade wherein malignant cells spread beyond the primary tumor through the blood and lymph systems, colonize distant organs, and eventually manifest as metastatic lesions[2]. Immune cells recruited to metastatic sites undergo extensive reprogramming, and contribute to the formation of tumor-

promoting or tumor-restraining immune microenvironments[3]. Indeed, metastatic tumor-

infiltrating leukocytes exhibit dual effects on metastasis. Cytotoxic CD8$^+$ tumor-infiltrating

lymphocytes (TILs) and natural killer (NK) cells can restrain metastatic tumor cells under

certain conditions[4-6]. However, their anti-tumor function is frequently compromised by

exhaustion, characterized by upregulation of inhibitory receptors (PD-1, TIM3, LAG3) and

impaired effector functions within the immunosuppressive tumor microenvironment[7-9].

Similarly, while macrophages, neutrophils, IL-17-producing γδ T cells, and regulatory T

cells are commonly associated with promoting metastasis outgrowth[10-12], specific subsets

within these populations can exert anti-metastatic effects depending on their polarization

state and local microenvironmental cues. For instance, M1-polarized macrophages exhibit

anti-tumor activity through production of pro-inflammatory cytokines and direct

cytotoxicity[13,14], and presentation of tumor- antigens[15], N1-polarized neutrophils can

suppress tumor progression through reactive oxygen species production and immune

activation[16,17], and certain γδ T cell subsets demonstrate potent cytotoxic activity against

tumor cells and support anti-tumor immunity[18-20]. These observations indicate that the

phenotypic heterogeneity and functional state of tumor-infiltrating immune cells, shaped by

local microenvironmental cues, critically influence metastatic colonization.

Clonal hematopoiesis (CH), the age-associated acquisition of somatic mutations in

hematopoietic stem and progenitor cells, has emerged as a major source of systemic

inflammation and immune dysregulation. Analysis of patient samples and animal models

indicates that CH-derived immune cells underlie these inflammatory changes[21]. The

prevalence of CH in solid cancer patients is 20-30%, which is significantly higher than in

age-matched healthy individuals[22,23]. *DNMT3A*, *TET2*, and *ASXL1* are the most commonly

mutated genes driving CH in untreated solid cancer patients and in healthy individuals[23],

whereas exposure to cancer therapy heightens the frequency of mutations in DNA damage

response genes, such as *TP53*, *CHEK2,* and *PPM1D*[22]. Indeed, CH-derived mutant alleles

are frequently detected in tumor-infiltrating immune cells of solid cancer patients with CH[23-

26]. Tumor infiltration by CH-derived mutant immune cells shapes the local immune

microenvironment, which has been associated with tumor inflammation and poor

prognosis[27,28]. However, the clinical and biological implications of CH-derived immune

cells on solid cancer remain incompletely understood. Mouse studies revealed conflicting

results regarding whether CH-associated immune alterations accelerate or restrain tumor

growth across different cancer types[29,30]. Pan-human cancer analyses suggest CH-outcome

associations vary based on cancer type[22,23,31,32], and CH-driver specific outcome analyses are

lacking. Recently, *TET2*-mutant CH was shown to potentiate response to immune checkpoint

blockade in colorectal cancer (CRC), non-small cell lung cancer, melanoma, and animal

models[15,29,33]. Therefore, TET2 inactivation in tumor-infiltrating myeloid and lymphoid cells

can enhance anti-tumor immunity. Intriguingly, the potential roles of CH-derived immune

cells on cancer metastasis have not been clarified.

Here, we examined the association between CH and metastasis in a large clinical

cohort and further investigated how *Tet2*-deficient CH-derived immune cells regulate CRC

liver metastasis using conditional *Tet2* knockout mouse models.

RESULTS

112   *TET2*-mutant CH is associated with lower metastatic cancer burden

113   We analyzed CH in 16,744 non-hematological cancer patients from the MSK-IMPACT

114   cohort with matching demographic and clinical data including treatment history, cancer stage,

115   and metastatic burden[22,34]. Driver clonal hematopoiesis (CH) was defined as putative CH-

116   driver mutations with a variant allele frequency (VAF) ≥ 0.02 (Methods). Reproducing the

117   observations from the larger cohort of Bolton et al. 2020 [22], overall CH prevalence was 19.6%,

118   most commonly with *DNMT3A* (10.2%), *TET2* (3.1%), and *PPM1D* (2.6%) driver mutations

119   (Figure S1A). Prevalence increased with age, and patients with driver CH were significantly

120   older (Figures S1B and S1C). Known associations between *ASXL1* and smoking history, and

121   with *CHEK2*, *TP53*, *PPM1D* and prior exposure to cancer therapy, were recapitulated

122   (Figures S1D and S1E). Metastatic cases were more likely to have received therapy (Figure

123   S1F), and CH prevalence patterns varied by cancer type in accordance with prior studies

124   (Figures S1G and S1H).

125   To investigate a potential association between CH and cancer metastasis, we

126   compared CH prevalence in patients with and without metastatic cancer. In multivariate

127   adjusted analysis, there was a trend towards higher prevalence of driver CH in patients

128   without metastatic disease (OR 0.89, 95%CI [0.78–1.01], FDR-adjusted *P* = 0.109, Figure

129   1A). CH with a *TET2*-driver mutation was significantly more common in patients with local

130   compared to metastatic cancers (OR 0.64, 95%CI [0.46–0.90], *P* = 0.029; Figure 1A), while

131   *DNMT3A*-driven CH and other forms of CH with non-*DNMT3A*, non-*TET2* drivers were not

132   associated with metastatic status (Figure 1A). Non-driver somatic *TET2* variants were not

133   associated with metastatic status (Figure S1I). By cancer type, *TET2*-driver CH was

associated with lower odds of having metastatic breast cancer (multivariate logistic

regression, OR 0.36, P = 0.022), with similar trends in non–small cell lung cancer (NSCLC)

(OR 0.52, P = 0.064) and in CRC (OR 0.53, P = 0.419) (Figure 1B, Table S1). *TET2*-CH

was numerically but not statistically more common in patients with metastatic prostate cancer

(multivariate logistic regression, OR 1.98, P = 0.419). When examining metastatic status at

specific sites, patients with *TET2*-driver CH showed significantly reduced odds of having

liver (OR 0.67, P = 0.016) or brain metastasis (OR 0.54, P = 0.016), and showed a trend

towards lower odds of bone metastasis (OR 0.77, P = 0.114; Figure 1C).

When examining patients with metastatic disease**,** *TET2*-driver CH was linked to lower

metastatic disease burden as quantified by number of metastases (median and interquartile

range [IQR] metastasis number for *TET2*-CH, 3 [2-6], versus no CH and non-*TET2*-CH, 4

[2-7], P = 0.035; multivariate negative binomial generalized linear model IRR = 0.92, P =

0.052; Figures 1D and S1J) and involvement of fewer metastatic sites (median and IQR

metastasis site number for *TET2*-CH, 3 [2-5], versus no CH and non-*TET2*-CH, 3 [2-5], P =

0.006; multivariate negative binomial generalized linear model IRR = 0.92, P = 0.035; Figure

1E and S1K).

Overall, patients with driver CH mutations experienced modestly worse survival in

multivariate analysis (HR 1.06, P = 0.0498; Figure S1L). Strikingly, this was not driven by

DNMT3A- and *TET2*- driver CH (Figure 1F). While patients with *TET2*-driver CH trended

towards improved OS in multivariate analysis compared with no CH (HR 0.87, P = 0.152;

Figure 1F), patients with DDR- (HR 1.24, P < 0.001), cohesin- (HR 1.40, P = 0.096), and

spliceosome-driver CH (HR 1.30, P = 0.034) experienced worse survival (Figure 1F).

Therefore, solid tumor patients with *TET2*-CH are more likely to have local versus metastatic

cancer, those with metastatic cancer tend to have fewer metastases, and in contrast to other

forms of CH, *TET2*-mutant CH is not associated with worse overall survival. Based on these

results, we explored whether *TET2*-mutant CH could suppress cancer metastasis in animal

models.

Figure 1. *TET2*-**mutant CH is associated with reduced cancer metastasis**

(A) Adjusted odds of patient having metastatic versus local cancer by CH-mutant status with any CH-driver mutation, or stratified by (bottom) only *DNMT3A* driver, only *TET2* driver, and non-*TET2* non-*DNMT3A* mutant CH, with the reference of no CH. ORs and 95% confidence intervals are reported from logistic GLMs, and *P* values are FDR-adjusted.

(B) Adjusted odds of patient having metastatic versus local cancer given exposure to *TET2*-mutant CH for the four most prevalent cancer types. ORs and 95% confidence intervals are reported from logistic GLMs for each cancer type, and *P* values are FDR-adjusted (see also Table S1).

(C) Adjusted odds of metastasis occurring at specific sites given exposure to *TET2*-mutant CH. ORs and 95% confidence intervals are reported from logistic GLMs for each metastatic site, and *P* values are FDR-adjusted.

(D) The number of metastases in patients with metastatic cancer stratified by *TET2*-mutant CH presence. Differences in distributions are quantified both univariately by Wilcoxon rank sum tests, and multivariately adjusted with IRRs from negative binomial generalized linear models (Supplementary Figure1J).

(E) The distribution of the number of metastatic sites in patients with metastatic cancer stratified by *TET2*-mutant CH presence. Differences in distributions are quantified both univariately by Wilcoxon rank sum tests, and multivariately adjusted with IRRs from negative binomial generalized linear models (Supplementary Figure1K).

(F) Kaplan-Meier (KM) survival curves for overall survival of all patients, stratified by no CH mutation, and *DNMT3A*-, *TET2*-, DNA-damage-response (DDR)-, cohesin-, and spliceosome-mutant CH. Adjusted hazard ratios (HR) are reported from a Cox proportional hazards model, with the reference of no CH, and *P* values are FDR-adjusted.

CH, clonal hematopoiesis; OR, Odds ratio; GLM, generalized linear model; FDR, false discovery rate; IRR, incidence risk ratio. Detailed statistical outputs, test statistics, and model parameters for all panels in Figure 1 are provided in Table S2.

163 *Tet2* deficiency in T cells inhibits CRC liver metastasis in mice

We investigated the effects of hematopoietic *Tet2* deficiency on CRC metastasis to the liver,

since *TET2*-CH was prominently associated with reduced metastasis to this site (Figure 1C)

and CRC is one of the most common causes of liver metastasis. To explore the role *of Tet2*

loss of function in different types of immune cells on CRC metastasis, we crossed mice

carrying *LoxP*-flanked *Tet2* alleles (*Tet2^{fl/fl}*)[35] with *Vav1^{Cre}* [36], *Cd4^{Cre}* [37], *Cd19^{Cre}* [38], and

LysM^{Cre} [39] to generate mice with targeted *Tet2* knockout (KO) in all hematopoietic cells

(*Vav1^{Cre}Tet2KO* mice), B lymphoid cells (*Cd19^{Cre}Tet2KO* mice), T lymphoid cells

(*Cd4^{Cre}Tet2KO* mice), and myeloid cells (*LysM^{Cre}Tet2KO* mice), respectively. Venus-

labeled mouse intestinal tumor-derived organoid cells carrying *Apc^{Δ716}*, *Kras^{+/G12D}*, *Tgfbr2^-*

^{/-} and *Trp^{R270H/LOH}* mutations (Venus-*AKTP^{M/LOH}* cells)[40], which acquired the ability to

metastasize to the liver, were injected into the spleen of various *Tet2* conditional KO mice

and *Tet2^{fl/fl}* mice to generate CRC liver metastases (Figure 2A). After 30 days, liver metastatic

tumor burden (LMTB) was lower in *Vav1^{Cre}Tet2KO* mice and *Cd4^{Cre}Tet2KO* mice than in

Tet2^{fl/fl} mice, while LMTB of *Cd19^{Cre}Tet2KO* and *LysM^{Cre}Tet2KO* mice was comparable

with that of *Tet2^{fl/fl}* mice (Figures 2B and 2C; Table S3). This result suggests that *Tet2*

deficiency in T cells, but not in B cells or myeloid cells, is sufficient to restrict CRC liver

metastasis in mice.

Figure 2

Figure 2. *Tet2* **deficiency in T cells, but not in myeloid and B cells, inhibits the liver metastasis of CRC in mice.**

(A) Schematic of the CRC liver metastasis model using cell-type-specific *Tet2* knockout mice. CRC organoid-derived Venus-*AKTP*$^{M/LOH}$ cells (1×10^6 cells/mouse) were intrasplenically injected into *Vav1*^{Cre}*Tet2KO, Cd4*^{Cre}*Tet2KO*, *Cd19*^{Cre}*Tet2KO*, *LysM*^{Cre}*Tet2KO*, and *Tet2*$^{fl/fl}$ control mice. Liver metastases were analyzed 4 weeks post-injection.

(B) Representative macroscopic (top row), HE-stained (middle rows; 40× and 100× magnification), and GFP fluorescence (bottom row) images of liver metastases. Scale bars: 1 cm (macroscopic and GFP); 5 mm (HE, 40×); 500 μm (HE, 100×).

(C) Quantification and comparisons of LMTB between different Tet2KO models and *Tet2*$^{fl/fl}$ controls. See also Table S3 for detailed quantification and statistical comparisons.

CRC, colorectal cancer; HE, hematoxylin and eosin; GFP, green fluorescent protein; LMTB, liver metastatic tumor burden. Data in (C) are pooled from 3 independent experiments (total n = 11-14 mice per group). Statistical significance was determined by Mann-Whitney U test. *p < 0.05, **p < 0.01, ***p < 0.001, ****p < 0.0001, ns = not significant.

183   ***Tet2* deficiency reshapes CD8$^+$ TILs through the *Tox*-exhaustion axis to enhance anti-**

184   **tumor immunity**

185   To determine which T cell subsets are regulated by *Tet2* in the anti-tumor response, we

186   compared the transcriptional profiles of CD4$^+$ TILs and CD8$^+$ TILs isolated from CRC liver

187   metastases of *Vav1^{Cre}Tet2KO* mice, *Cd4^{Cre}Tet2KO* mice, and *Tet2$^{fl/fl}$* mice (Figure 3A).

188   Principal component analysis (PCA) and heatmap clustering showed that biological

189   replicates of CD8$^+$ TILs from either *Vav1^{Cre}Tet2KO* mice or *Cd4^{Cre}Tet2KO* mice

190   (*Vav1^{Cre}/Cd4^{Cre}Tet2KO* CD8$^+$ TILs) clustered closely together but distinctly from their

191   *Tet2$^{fl/fl}$* counterparts, while *Tet2KO* CD4$^+$ TILs showed no significant separation from *Tet2$^{fl/fl}$*

192   CD4$^+$ TILs (Figures 3B and S2A, Table S4). Comparing *Vav1^{Cre}/Cd4^{Cre}Tet2KO* groups to

193   *Tet2$^{fl/fl}$* group identified substantially more differentially expressed genes (DEGs) in CD8$^+$

194   TILs than in CD4$^+$ TILs (Figure 3C). In the *Vav1^{Cre}Tet2KO* model, CD8$^+$ TILs exhibited 226

195   DEGs compared to 71 DEGs in CD4$^+$ TILs. This disparity was even more pronounced in the

196   *Cd4^{Cre}Tet2KO* model, where CD8$^+$ TILs demonstrated 209 DEGs versus only 8 DEGs in

197   CD4$^+$ TILs (FDR< 0.01, |Log2 FC| > 2) (Figure 3C, Table S5-S8). These findings suggest

198   that *Tet2* deficiency affects the transcriptome of CD8$^+$ TILs more than CD4$^+$ TILs.

199   Notably, multiple exhaustion-associated genes including *Pdcd1*, *Lag3*, *Tim3*, *Tigit*,

200   *Prdm1*, *Tnfrsf9*, and *Tox* were the most significantly downregulated genes in

201   *Vav1^{Cre}/Cd4^{Cre}Tet2KO* CD8$^+$ TILs relative to *Tet2$^{fl/fl}$* CD8$^+$ TILs, suggesting attenuated

202   exhaustion programs in the absence of *Tet2*. (Figure 3D; Table S5 and S6). In CD8$^+$ TILs,

203   DEG analysis showed substantial concordance between *Vav1^{Cre}Tet2KO* and *Cd4^{Cre}Tet2KO*

204   models with 198 commonly upregulated and 89 commonly downregulated genes, including

downregulation of exhaustion markers *Tox* and *Pdcd1* (FDR< 0.05, |Log2 FC| > 1) (Figure

S2B). Using DAVID gene functional classification[41], we clustered 89 overlapping

downregulated genes shared by *Vav1^{Cre}Tet2KO* CD8$^+$ TILs and *Cd4^{Cre}Tet2KO* CD8$^+$ TILs

compared to *Tet2$^{fl/fl}$* CD8$^+$ TILs (Table S9A). Under stringent parameters (Table S9B), these

consistently downregulated genes were classified into a single cluster comprising six genes

associated with immunologic function, with an enrichment score of 2.59 (Figure 3E). Notably,

this downregulated cluster encompassed key genes linked to T-cell exhaustion, including

Pdcd1, *Lag3*, and *Tigit* (Figure 3E; Table S9C).

Gene Set Enrichment Analysis (GSEA)[42] was performed using the ImmuneSigDB

subset C7 from the Molecular Signatures Database (MSigDB)[43] to compare the genome

expression profiles of *Tet2KO* CD8$^+$ TILs and *Tet2$^{fl/fl}$* CD8$^+$ TILs. The analysis identified 92

gene sets significantly enriched in *Tet2$^{fl/fl}$* CD8$^+$ TILs compared to *Cd4^{Cre}Tet2KO* CD8$^+$ TILs,

five of which were specifically associated with T-cell exhaustion (Figures 3F and S3A; Table

S10). In the same approach, 51 gene sets related to CD8 T-cell function were enriched in

Tet2$^{fl/fl}$ CD8$^+$ TILs compared to *Vav1^{Cre}Tet2KO* CD8$^+$ TILs, including five exhaustion-

associated signatures (Figures 3F and S3A, Table S11). Of the 34 gene sets consistently

enriched in *Tet2$^{fl/fl}$* CD8$^+$ TILs relative to both *Vav1^{Cre}Tet2KO* and *Cd4^{Cre}Tet2KO*

counterparts, four were linked to T-cell exhaustion (Figure S3B).

Ingenuity Pathway Analysis (IPA) of 89 consistently downregulated genes in

Vav1^{Cre}Tet2KO CD8$^+$ TILs and *Cd4^{Cre}Tet2KO* CD8$^+$ TILs (Table S12) identified the

transcription factor *Tox* as one of the potential upstream regulators targeting genes associated

with T-cell exhaustion (overlap *p*-value = 0.00164) (Figure 3G; Table S12). Consistent with

this finding, *Tox* exhibited the greatest differential expression with the lowest *P* value among

all tested genes, with levels markedly higher in *Tet2$^{fl/fl}$* CD8$^+$ TILs compared to

Vav1^{Cre}/Cd4^{Cre}Tet2KO CD8$^+$ TILs (*Tet2$^{fl/fl}$* vs. *Vav1^{Cre}Tet2KO*, *P* = 0.0002; *Tet2$^{fl/fl}$* vs.

Cd4^{Cre}Tet2KO, *P* = 0.0001; Figure S3C; Table S13).

Furthermore, flow cytometry analysis of CRC metastatic tumors in the liver revealed

a significantly higher percentage of CD8$^+$ TILs co-expressing exhaustion markers Tim3 and

PD-1 of the *Tet2$^{fl/fl}$* group compared with that of the *Vav1^{Cre}/Cd4^{Cre}Tet2KO* group (Figure

S4A; Table S14A). Indirect immunofluorescence staining further demonstrated a higher

number of CD8α cells co-expressing Tim3, PD-1 and Tox in tumors from the *Tet2$^{fl/fl}$* group

than in the *Vav1^{Cre}/Cd4^{Cre}Tet2KO* groups (Figure S4B; Table S14B). Collectively, these

findings identify *Tet2* as a critical enforcer of the terminal exhaustion program, which is

intrinsically linked to high *Tox* expression.

To further validate the role of *Tet2* in regulating the antitumor functions of CD8$^+$

TILs, we performed CD8 T-cell depletion experiments in *Cd4^{Cre}Tet2KO* and *Tet2$^{fl/fl}$* mice

(Figure 3H). Circulating and liver-infiltrating CD8 T cells were efficiently and specifically

depleted in anti-CD8α antibody-treated groups compared with PBS controls (Figures S5A

and S5B). CD8 T-cell depletion in *Cd4^{Cre}Tet2KO* mice drastically increased LMTB

compared to their PBS-treated controls (*P* = 0.007), while CD8 T-cell depletion in *Tet2$^{fl/fl}$*

mice had no significant effect (*P* = 0.396) (Figures 3I, 3J and S5C; Table S15). As a result,

LMTB of anti-CD8α antibody-treated *Cd4^{Cre}Tet2KO* mice was not significantly different

from that of anti-CD8α antibody-treated *Tet2$^{fl/fl}$* mice (*P* = 1.0) and was comparable to that

of PBS-treated *Tet2$^{fl/fl}$* mice (*P* = 1.0), indicating that CD8 T-cell depletion attenuated the

tumor-restrictive effect observed in *Tet2*-deficient mice (Figures 3I, 3J and S5C; Table S15).

In contrast, PBS-treated *Cd4^Cre^Tet2KO* mice revealed a significantly lower LMTB compared

to PBS-treated *Tet2^fl/fl^* mice ($P = 0.007$) (Figures 3I, 3J and S5C; Table S15), consistent with

our earlier results. Collectively, these data demonstrate that the enhanced tumor control

observed in *Cd4CreTet2KO* mice requires CD8$^+$ TILs.

Figure 3. *Tet2* **deficiency extensively reshapes CD8$^+$ TILs transcriptional profiles through the *Tox*-exhaustion axis to enhance anti-tumor immunity in CRC liver metastasis.**

(A) Schematic of the experimental workflow for transcriptomic profiling. CD45$^+$ tumor-infiltrating leukocytes were isolated via MACS and sorted by FACS into CD4$^+$ and CD8$^+$ T cells, respectively. Total RNA was extracted from sorted populations and subjected to library preparation using SMARTer Stranded Total RNA-seq Kit, followed by high-throughput sequencing (Hi-seq X) and data analysis (CLC workbench).

(B) PCA plots of CD4 and CD8$^+$ TILs supervised by the top 200 DEGs. Gene ranking was based on combined FDR calculated by multiplying the two individual FDR from *Vav1^{Cre}Tet2KO* vs. *Tet2$^{fl/fl}$* FDR and *Cd4^{Cre}Tet2KO* vs. *Tet2$^{fl/fl}$* comparisons (selection criteria: FDR < 0.05, |Log2 FC| > 1 in both comparisons) (see also Table S4).

(C) Stacked bar plots show the number of upregulated and downregulated genes in CD4 and CD8$^+$ TILs from the comparison between *Vav1^{Cre}/Cd4^{Cre}Tet2KO* groups to *Tet2$^{fl/fl}$* group (FDR < 0.01, |Log2 FC| > 2).

(D) Volcano plots show DEGs of *Vav1^{Cre}/Cd4^{Cre}Tet2KO* CD8$^+$ TILs in comparison with *Tet2$^{fl/fl}$* CD8$^+$ TILs (FDR < 0.05, |Log2 FC| > 1).

(E) Venn diagrams display 89 overlapping down-DEGs in the comparison between *Vav1^{Cre}Tet2KO* and *Cd4^{Cre}Tet2KO* with *Tet2$^{fl/fl}$* CD8$^+$ TILs (FDR < 0.05, Log2 FC < -1). Also shown is the DAVID functional classification of these overlapping genes, highlighting a cluster associated with immunologic function (Enrichment Score: 2.59; stringent parameters).

(F) GSEA using ImmuneSigDB subset of C7 showing the enrichment of T-cell exhaustion-related signatures in *Tet2$^{fl/fl}$* CD8$^+$ TILs relative to *Vav1^{Cre}/Cd4^{Cre}Tet2KO* CD8$^+$ TILs (FDR < 0.25, nominal p < 0.05).

(G) Interaction network of *Tox* and T-cell exhaustion-related genes identified by IPA. The analysis was based on the 89 overlapping down-DEGs within CD8$^+$ TILs from *Vav1^{Cre}Tet2KO* and *Cd4^{Cre}Tet2KO* groups compared to the *Tet2$^{fl/fl}$* groups (FDR < 0.05, Log2 FC < -1).

(H) Schematic of the CD8 T-cell depletion experiment. Venus-*AKTP$^{M/LOH}$* cells were intrasplenically transplanted into *Cd4^{Cre}Tet2KO* or *Tet2$^{fl/f}$* mice. Anti-CD8α antibody or PBS was initiated 2 days before transplantation, and continued every 4 days throughout tumor progression.

(I) Representative macroscopic and GFP fluorescence images of liver metastases 30 days post-injection. Scale bars, 1 cm.

(J) Quantification of LMTB based on the ratio of tumor area to total liver area (%) See also Table S15 for detailed quantification and statistical comparisons.

TILs, tumor-infiltrating lymphocytes; CRC, colorectal cancer; MACS, magnetic-activated cell sorting; FACS, fluorescence-activated cell sorting; PCA, principal component analysis; DEG, differentially expressed gene; GSEA, Gene Set Enrichment Analysis; IPA, Ingenuity Pathway Analysis; GFP, green fluorescent protein; LMTB, liver metastatic tumor burden; PBS, phosphate-buffered saline. Data in (J) are compiled of 2 independent experiments (total n = 7-8 mice per group). Statistical significance in (J) was determined by pairwise Mann-Whitney U tests with Bonferroni correction for multiple comparisons. *$P < 0.05$, **$P < 0.01$, ***$P < 0.001$, ****$P < 0.0001$, ns = not significant.

256 ***Tet2* deficiency preserves effector function and restricts terminal differentiation of**

257 **CD8$^+$ TILs**

258 To comprehensively characterize the transcriptional landscape of *Tet2*-deficient CD8$^+$ TILs,

259 we performed single cell RNA sequencing (scRNA-seq) with V(D)J profiling. We profiled

260 16,790 and 17,367 CD45$^+$ tumor-infiltrating leukocytes isolated from liver metastatic tumors

261 of *Tet2$^{fl/fl}$* (*n* = 2) and *Vav1^{Cre}Tet2KO* (*n* = 2), respectively (Figure S6A). CD8$^+$ TIL clusters

262 were identified *in silico* from CD45$^+$ tumor-infiltrating leukocytes (Figures S6B-S6J). *In*

263 *silico*-defined CD8$^+$ TILs were subjected to unsupervised sub-clustering, resulting in the

264 identification of 9 distinct clusters, consisting of stem-like CD8 T cells (T$_{SLC}$, *Ly6c2*^{low}CD8

265 T$_{SLC}$ and *Ly6c2*^{high}CD8 T$_{SLC}$), precursor/progenitor T cells (T$_{PROG}$), cytotoxic/effector-like

266 exhausted T cells (T$_{EFF-1}$ and T$_{EFF-2}$), intermediate exhausted T cells (T$_{INT-1}$ and T$_{INT-2}$),

267 terminally exhausted T cells (T$_{TEX}$), and proliferative exhausted T cells (P-T$_{EX}$) (Figures 4A

268 and 4B). T$_{PROG}$ was characterized by the high expression of stem-like T-cell markers (*Sell*,

269 *Ccr7*, *Tcf7*, *Il7r* and *Id3*) along with the dual expression of effector-like and exhausted T-cell

270 markers at moderate or low levels (Figure 4C, Table S16). T$_{EFF}$ (T4 and T5) cells displayed

271 moderate to high expression of cytolytic/effector-like and exhaustion markers. T$_{TEX}$ (T8) cells

272 were defined by the high expression of multiple exhaustion markers and reduced expression

273 in effector-like markers (Figure 4C, Table S16). Among the spectrum of CD8$^+$ TILs, T$_{PROG}$

274 (T3) and T$_{EFF}$ (T4 and T5) were predominant in *Vav1^{Cre}Tet2KO* mice, while T$_{INT}$ (T6 and

275 T7), T$_{TEX}$ (T8) and P-T$_{EX}$ (T9) were dominant in control *Tet2$^{fl/fl}$* mice (Figure 4B). Among

276 stem-like CD8$^+$ TILs, *Ly6c2*^{high}T$_{SLC}$ (T2) was markedly enriched in *Tet2$^{fl/fl}$* compared to

277 *Vav1^{Cre}Tet2KO* mice (Figure 4B). DEG analysis revealed significant upregulation of

exhaustion-associated genes, including *Tox, Tim3, Tnfrsf9, Eomes,* and *Gzmk,* in *Tet2$^{fl/fl}$*

CD8$^+$ TILs comparing to *Tet2KO* CD8$^+$ TILs, consistent with bulk RNA-seq data (Figure

4D; Table S17). In contrast, cytolytic and effector-related genes, such as *Klrc1, Klrk1, Ifng,*

Tnf, Cd160, Ly6a, Cxcr3, Id2, Nr4a, and *Ccl4*, were significantly upregulated in *Tet2KO*

CD8$^+$ TILs (Figure 4D; Table S17). Gene-set activity measured by AUCell scoring[44] further

demonstrated significant enrichment of effector-versus exhausted-T-cell signatures (p =

0.013), as well as memory- versus exhausted-T-cell signatures ($p < 0.0001$), in *Tet2KO* CD8$^+$

TILs compared with *Tet2$^{fl/fl}$* CD8$^+$ TILs (Figure 4E).

To characterize the exhaustion phenotype, we analyzed the expression of a signature

comprising 42 genes consistently associated with exhausted T cells[45] (Table S18) via

heatmap analysis (Figure 4F). T_{SLC} (T1 and T2) clusters exhibited lower average expression

of most exhaustion-associated markers, consistent with their progenitor state. In contrast,

T_{TEX} (T8) and P-T_{EX} (T9) clusters displayed high expression across multiple exhaustion

markers (Figure 4F). Notably, 32 out of the 42 exhaustion-associated genes were highly

expressed in *Tet2$^{fl/fl}$* CD8$^+$ TILs but downregulated in *Vav1^{Cre}Tet2KO* CD8$^+$ TILs (Figure

4F). To quantify this, we calculated the exhaustion score using this 42 gene signature. The

T-cell exhaustion score delineated the spectrum of exhausted T cells with positive scoring

values increasing toward the T_{TEX} cluster (Figure 4G). *Tet2$^{fl/fl}$* CD8$^+$ TILs exhibited a

significantly higher exhaustion score than *Tet2KO* CD8$^+$ TILs (Figure 4H).

Using *Monocle 3*[46], we reconstructed the pseudotime trajectory of CD8$^+$ TILs to

investigate differentiation dynamics within the metastatic tumor microenvironment (Figure

4I). *Ly6c2^{low}*T_{SLC} cells were located at the root of the trajectory, consistent with their

progenitor-like identity. Along this trajectory, T_{PROG} cells differentiated toward T_{EFF-1} or

diverged toward T_{TEX} after transitioning through T_{INT} states. A subset of T_{EFF-1} cells retained

cytolytic capacity as inferred by effector signatures and progressed into T_{EFF-2}, while others

gradually lost effector function and transitioned to T_{TEX} (Figure 4I). We next examined the

T-cell receptor (TCR) clonotype landscape using strict clonotype calling based on the

combination of V(D)J genes and the nucleotide sequence of the CDR3 region[47] (Figure 4J).

Clonal expansion states were classified by TCR clone size: single (1 cell), small (2–5 cells),

medium (6–20 cells), large (21–100 cells), and hyperexpanded (>100 cells). Hyperexpanded

clones were predominantly enriched within T_{INT} and T_{TEX} clusters of *Tet2^{fl/fl}* CD8$^+$ TILs,

reflecting a biased differentiation toward terminal exhaustion (Figures 4J, S7A and S7B).

Clonal structure analysis revealed significantly greater repertoire diversity in Tet2KO

mice, whereas *Tet2^{fl/fl}* mice exhibited marked oligoclonal dominance, particularly within

exhausted T cell states (Figures S7C and S7D). Clone size distribution across differentiation

states demonstrated that hyperexpanded clones in *Tet2^{fl/fl}* were predominantly concentrated

in exhaustion states (T_{INT} [T6-T7], T_{TEX} [T8], and P-T_{EX} [T9]), while medium and large

clones were underrepresented in effector states (T_{EFF-1} and T_{EFF-2} [T4-T5]). Conversely,

Tet2KO CD8$^+$ TILs displayed more balanced clone size distribution (Figures 4J and S7E-

S7F).

The TCR clonotype overlap analysis highlighted distinct differences in differentiation

dynamics between *Tet2^{fl/fl}* and *Tet2KO* CD8$^+$ TILs. In *Tet2^{fl/fl}* CD8$^+$ TILs, there was an

absence of TCR overlap between the progenitor-like states (*Ly6c2*^high T_{SLC} [T2] and T_{PROG}

[T3]) and cytolytic/effector states (T_{EFF-1} [T4] and T_{EFF-2} [T5]), indicating a restricted clonal

relationship (Figure 4K). Conversely, *Tet2KO* CD8$^+$ TILs exhibited robust TCR overlap

among these populations (Figure 4K). Collectively, these data demonstrate that *Tet2*

deficiency prevents the oligoclonal expansion of terminally exhausted clones, while

preserving a broad and diverse TCR repertoire across the progenitor and effector states.

Figure 4. *Tet2* deficiency preserves effector function and restricts terminal differentiation of CD8$^+$ TILs in CRC liver metastasis.

(A) Nine subclusters were labeled with different colors on the UMAP plot of the total 2,202 CD8$^+$ TILs from both *Tet2$^{fl/fl}$* group ($n = 2$) and *Vav1^{Cre}Tet2KO* group ($n = 2$).

(B) UMAP plot of CD8$^+$ TIL subclusters of each mouse group (1,101 CD8$^+$ TILs for each group). The pie plots show percentages of each CD8$^+$ TIL subcluster in each group.

(C) Bubble plot showing the expression of selected marker genes among the top 50 most significant DEGs across defined clusters. The list of the top 50 most significant DEGs for each cluster are provided in Table S16.

(D) Volcano plot displaying DEGs between *Tet2$^{fl/fl}$* and *Vav1^{Cre}Tet2KO* CD8$^+$ TILs, highlighting T-cell exhaustion markers and cytolytic/effector-associated genes (adjusted *P* value < 0.05). List of DEGs is provided in Table S17.

(E) Box plots comparing the enrichment scores of effector versus exhausted T-cell signatures ($p = 0.013$) and memory versus exhausted T-cell signatures ($p < 0.0001$) between *Tet2$^{fl/fl}$* and *Vav1^{Cre}Tet2KO* CD8$^+$ TILs.

(F) Heatmap displaying the average expression of 42 exhaustion-associated marker genes across subclusters and genotypes.

(G) UMAP plot with cells labeled by exhaustion score characterize the spectrum of exhausted T cells with positive scoring values increasing toward T$_{TEX}$.

(H) Box plot comparing exhaustion scores between *Tet2$^{fl/fl}$* and *Tet2KO* CD8$^+$ TILs ($p = 0.046$).

(I) Pseudotime trajectory of CD8$^+$ TIL subclusters inferred by Monocle3 (left), with cells ordered along the trajectory using *Ly6c2^{low}* T$_{SLC}$ as the root (right). Arrows indicate inferred differentiation paths toward T$_{EFF}$ and T$_{TEX}$ states.

(J) UMAP plot with cells colored by TCR clonal size categories, showing the distribution of clones across states in each group.

(K) TCR repertoire overlap analysis using the Morisita overlap index between progenitor-like, effector, and exhausted states across groups.

TILs, tumor-infiltrating lymphocytes; CRC, colorectal cancer; UMAP, Uniform Manifold Approximation and Projection; DEG, differentially expressed gene; T$_{TEX}$, terminally exhausted T cell; T$_{EFF}$, cytotoxic/effector-like exhausted T cells; T$_{SLC}$, stem-like CD8 T cell; TCR, T-cell receptor.

Statistical significance in (E) and (H) was determined by Mann-Whitney U test. *$p < 0.05$, **$p <$ 0.01, ***$p < 0.001$, ****$p < 0.0001$, ns = not significant.

328 ***Tet2* deficiency cell-intrinsically restrains terminal exhaustion and enables sustained**

329 **effector generation via progenitor preservation**

330 To directly assess whether *Tet2* regulates T-cell exhaustion in a cell-intrinsic manner, we

331 utilized an *in vitro* chronic TCR stimulation model using purified CD8 T cells from

332 *Cd4^{Cre}Tet2KO* and *Tet2$^{fl/fl}$* mice. Following initial activation with plate-bound anti-CD3ε and

333 soluble anti-CD28 antibodies, cells were repeatedly re-stimulated with anti-CD3ε antibody

334 alone to mimic persistent antigen exposure (Figure 5A).

335 Flow cytometric analysis at days 7 and 12 revealed that *Tet2* deficiency selectively

336 altered exhaustion fate decisions (Figures 5B and 5C). *Tet2KO* CD8 T cells displayed

337 significantly lower frequencies of PD-1$^+$Tim3$^+$T$_{TEX}$ and PD-1$^+$Tcf1$^-$T$_{TEX}$, accompanied by a

338 significant enrichment of PD-1$^+$Tcf1$^+$ PD-1$^+$ progenitor exhausted T (T$_{PEX}$) cells at both

339 timepoints (Figures 5B and 5C, Table S19A and S19B). Notably, longitudinal analysis

340 further demonstrated that *Tet2* is required for progressive depletion of the progenitor

341 exhausted compartment. Between days 7 and 12, the PD-1$^+$Tcf1$^+$T$_{PEX}$ pool was markedly

342 reduced in *Tet2$^{fl/fl}$* cultures, whereas this population was maintained in *Tet2KO* CD8 T cells,

343 despite accelerated T$_{TEX}$-cell accumulation in both genotypes (Figure 5D; Table S19C).

344 These findings validate our scRNA-seq-based observations and establish *Tet2* as a cell-

345 intrinsic regulator that promotes terminal exhaustion commitment at the expense of the

346 progenitor population under chronic TCR stimulation.

347 We next examined how *Tet2* deficiency influences memory and effector

348 differentiation states during prolonged TCR signaling. Consistent with their exhaustion

349 profile, *Tet2$^{fl/fl}$* CD8 T cells exhibited significantly higher proportions of CD44$^+$CD62L$^-$

effector/effector memory ($T_{EM/EFF}$) cells, whereas *Tet2KO* CD8 T cells retained significantly

higher frequencies of CD44$^+$CD62L$^+$ central memory-like (T_{CM}) cells at both days 7 and 12

(Figures 5E and 5F; Table S19A and S19B). Notably, although *Tet2KO* cells displayed lower

$T_{EM/EFF}$ proportions at individual timepoints, longitudinal analysis revealed robust

differentiation capacity. From day 7 to day 12, *Tet2KO* cultures exhibited a 6.69-fold

expansion of the $T_{EM/EFF}$ compartment, exceeding the 3.24-fold increase observed in *Tet2$^{fl/fl}$*

cells (Figures 5G and 5H). Importantly, enhanced effector output in *Tet2KO* cells was

achieved without excessive attrition of self-renewing populations. *Tet2KO* naive CD8 T cells

(T_{NAIVE}) were not significantly depleted over time, and while T_{CM} frequencies declined in

both genotypes, *Tet2KO* cells exhibited lower fold reductions, indicating superior

preservation of self-renewing memory compartments (Figures 5G and 5H; Table S19C).

Together, these data demonstrate that *Tet2* deficiency decouples effector generation from

progenitor depletion, enabling sustained differentiation while preserving Tcf1$^+$$T_{PEX}$ and

CD62L$^+$$T_{CM}$-like reservoirs under chronic TCR stimulation.

Figure 5

Figure 5.   *Tet2* **deficiency cell-intrinsically restrains terminal exhaustion while preserving self-renewing progenitor pools under chronic TCR stimulation.**

(A) Experimental design for *in vitro* chronic TCR stimulation assay. CD8 T cells isolated from $Cd4^{Cre}Tet2KO$ and $Tet2^{fl/fl}$ mice ($n = 9$ per group) were activated with anti-CD3ε (2 µg/mL) and soluble anti-CD28 (2 µg/mL) antibodies, then chronically re-stimulated with anti-CD3ε antibody alone (2 µg/mL) on days 4 and 8 to model persistent antigen exposure. Cells were analyzed by flow cytometry at days 7 and 12.

(B) Representative flow cytometry plots showing expression of exhaustion markers PD-1 and Tim3 (left), and PD-1 and Tcf1 (right) at days 7 and 12 in $Tet2^{fl/fl}$ (top row) and *Tet2KO* (bottom row) CD8 T cells.

(C) Quantification of exhaustion phenotypes. Box plots comparing frequencies of $PD-1^+Tim3^+T_{TEX}$, $PD-1^+Tcf1^-T_{TEX}$, and $PD-1^+Tcf1^+T_{PEX}$ between *Tet2KO* and $Tet2^{fl/fl}$ groups at day 7 (top) and day 12 (bottom). See also Table S19A and S19B for detailed quantification and statistical comparisons.

(D) Temporal dynamics of exhaustion differentiation. Longitudinal plots showing longitudinal changes from day 7 to day 12 of $PD-1^+Tcf1^+T_{PEX}$ (top), and $PD-1^+Tcf1^-T_{TEX}$ (bottom). See also Table S19A and S19C for detailed quantification and statistical comparisons.

(E) Representative flow cytometry plots showing CD62L and CD44 expression patterns at days 7 and 12 in $Tet2^{fl/fl}$ (top row) and *Tet2KO* (bottom row) CD8 T cells.

(F) Quantification of memory and effector differentiation states. Box plots comparing frequencies of $CD44^-CD62L^+T_{NAIVE}$, $CD44^+CD62L^+T_{CM}$, and $CD44^+CD62L^-T_{EM/EFF}$ between *Tet2KO* and $Tet2^{fl/fl}$ groups at day 7 and day 12. See also Table S19A and S19B for detailed quantification and statistical comparisons.

(G) Temporal dynamics of T_{NAIVE}, T_{CM}, and $T_{EM/EFF}$ populations. Longitudinal plots show longitudinal changes from day 7 to day 12 of the indicated populations (left to right). See also Table S19A and Table S19C for detailed quantification and statistical comparisons.

(H) Fold changes (day 12/day 7) on log2 scale quantifying differentiation kinetics. Numbers on bars indicate actual fold change values; y-axis is log2-transformed to show proportional changes.

TCR, T-cell receptor; T_{TEX}, terminally exhausted T cells; T_{PEX}, progenitor exhausted T cells; T_{CM}, central memory T cells; $T_{EM/EFF}$, effector memory/ effector T cells; T_{NAIVE}, Naïve T cells; AB, antibody. Data in (C), (D), (F), and (G) are compiled from three independent experiments (total $n =$

9 mice per group). Statistical significance for temporal comparisons (Day 7 vs. Day 12 within each genotype) determined by Wilcoxon signed-rank test (paired, two-tailed). Statistical significance for group comparisons (*Tet2^{fl/fl}* vs. *Tet2KO* at each timepoint) determined by Mann-Whitney U test (unpaired, two-tailed). *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$, ****$p < 0.0001$, ns = not significant.

366 ***Tet2* loss induces genome-wide hypermethylation and epigenetic silencing of the**

367 **exhaustion regulator *Tox* in PD-1$^+$CD8$^+$ TILs**

368 To determine whether *Tet2* loss impairs active DNA demethylation in CD8$^+$ TILs, we

369 performed whole genome bisulfite sequencing (WGBS) on activated PD-1$^+$CD8$^+$ TILs

370 isolated from liver metastatic tumors in *Vav1^{Cre}Tet2KO* and control *Tet2$^{fl/fl}$* mice (Figure 6A).

371 WGBS analysis comparing *Tet2KO* to *Tet2$^{fl/fl}$* PD-1$^+$CD8$^+$ TILs identified 19,492

372 differentially methylated regions (DMRs) across all chromosomes, of which 98.8% were

373 hypermethylated in *Tet2KO* cells (Figures S8A-C). Among these, 5,751 DMRs were mapped

374 to annotated genomic loci corresponding to 4,260 genes (Figure S8D; Table S20). A total of

375 2,644 DMRs were located in regulatory elements, including enhancer, promoter,

376 transcription start site (TSS), or CTCF-binding regions, and were designated as regulatory

377 DMRs (rDMRs) (Figure 6B; Table S21). In addition, the majority of rDMRs were distributed

378 within the enhancer or promoter regions (Figure 6B). Notably, *Tet2KO* PD-1$^+$CD8$^+$ TILs

379 exhibited a marked predominance of hypermethylated rDMRs (hyper-rDMRs; *n* = 2,519)

380 compared with hypomethylated rDMRs (hypo-rDMRs; *n* = 125) (Figure 6C). Integration of

381 WGBS with scRNA-seq revealed 22 genes with predominantly hypermethylated rDMRs that

382 were also transcriptionally downregulated in *Tet2KO* PD-1$^+$CD8$^+$ TILs compared to *Tet2$^{fl/fl}$*

383 PD-1$^+$CD8$^+$ TILs (Figures 6D; Table S22). Among these, *Tox*, a master regulator of T-cell

384 exhaustion, showed one of the most extensive methylation changes genome-wide (Figures

385 6D and S8E). A methylation impact score (total DMR count × mean absolute methylation

386 difference) (see Methods) similarly ranked *Tox* second and highlighted its widespread

387 hypermethylation across multiple regulatory regions (Figure S8F).

Within the *Tox* locus, differentially methylated CpG site (DMC) analysis identified

89 DMCs among 735 CpGs, all showing increased methylation in *Tet2KO* PD-1$^+$CD8$^+$ TILs

relative to *Tet2$^{fl/fl}$* PD-1$^+$CD8$^+$ TILs, with a mean gain of 58.2% (Figures 6E, 6F and S8G-

S8I; Table S23). DMR analysis focusing on the *Tox* locus using 500bp-window screening

revealed 56 significant DMRs with a mean methylation increase of 47.2% (Figures 6G and

S8J; Table S24). Notably, *Tox* showed extensive hypermethylation at regulatory elements,

including 8 enhancer DMRs with mean methylation difference of 45.2%, and 4 CTCF

binding site DMRs with mean methylation difference of 49.3% (Figures 6G and 6H). These

findings indicate that *Tet2* deficiency induces broad hypermethylation at critical regulatory

elements of *Tox*, leading to transcriptional repression of this central exhaustion-program

regulator in *Tet2KO* PD-1$^+$CD8$^+$ TILs.

Figure 6. ***Tet2* loss induces genome-wide hypermethylation and epigenetic silencing of *Tox* in PD-1$^+$CD8$^+$ TILs.**

(A) Schematic overview of WGBS workflow for analyzing DNA methylation in PD-1$^+$CD8$^+$ TILs.

(B) Distribution of DMRs across genomic features in *Tet2KO* vs. *Tet2$^{fl/fl}$* PD-1$^+$CD8$^+$ TILs. DMRs located at enhancer, promoter, TSS, or CTCF-binding regions, and were designated as regulatory DMRs (rDMRs).

(C) Pie chart displaying the proportion of hyper-rDMRs (n = 2519, 95.3%) versus hypo-rDMRs (n = 125, 4.7%).

(D) Venn diagram identifying 22 genes showing coordinated hypermethylation (q-value < 0.01, meth.diff > 0.2) and transcriptional downregulation (min.pct = 0.1, adjusted P value < 0.05) in *Tet2KO* PD-1$^+$CD8$^+$ TILs, including the master exhaustion regulator *Tox*.

(E) Heatmap displaying 89 significant DMCs across the *Tox* locus identified by DMC analysis, with Z-scores indicating methylation levels in *Tet2$^{fl/fl}$* and *Tet2KO* replications.

(F) Genome browser view of the *Tox* locus at CpG resolution showing methylation percentage at 735 CpG sites in *Tet2$^{fl/fl}$* (blue) and *Tet2KO* (red) PD-1$^+$CD8$^+$ TILs, with smoothed methylation trends and methylation difference plot below, the grey bars highlighting 89 significant DMCs. Gene structure indicates TSS, enhancers, CTCF binding sites, and exons.

(G) Scatter plot of methylation percentage differences across the *Tox* gene structure by DMR analysis, displaying 56 rDMRs at enhancer, CTCF, and intronic regions.

(H) Box plots of methylation percentage differences at enhancer, CTCF, and intronic regions within the *Tox* locus.

TIL, tumor-infiltrating lymphocytes; WGBS, whole genome bisulfite sequencing; DMR, differentially methylated region; DMC, differentially methylated cytosine; TSS, transcription start site; CTCF, CCCTC-binding factor; CDS, coding sequence; UTR, untranslated region; min.pct, minimum percentage of cells. For (B)-(H), statistical significance to identify DMRs and DMCs determined by *methylKit* using q-value < 0.01 and |meth.diff| $\geq$ 20%.

Tet2KO:Tet2WT chimeric mouse model of CH demonstrates restricted tumor growth and intrinsic resistance of Tet2-deficient CD8$^+$ TILs to terminal exhaustion

To model the mosaicism of tumor-infiltrating immune cells observed in solid tumor patients with CH, in which mutant and wild-type hematopoietic cells coexist within the tumor microenvironment, we established a competitive bone marrow transplantation (BMT) mouse model of CRC liver metastasis (Figure 7A). Lethally irradiated CD45.1$^+$ wild-type (WT) recipients were reconstituted with mixtures of *Vav1^{Cre+}Tet2$^{fl/fl}$* CD45.2$^+$ BM cells and *Tet2WT* CD45.1$^+$/CD45.2$^+$ competitor BM cells at ratios of 1:2 or 1:3, generating *Tet2KO:Tet2WT* mixed chimeras. Control *Tet2$^{fl/fl}$:Tet2WT* chimeras received a 1:1 mixture of *Tet2$^{fl/fl}$* CD45.2$^+$ and *Tet2WT* CD45.1$^+$/CD45.2$^+$ BM cells. Four weeks post-BMT, flow cytometric analysis of peripheral blood confirmed successful hematopoietic reconstitution, with donor-derived CD45.2$^+$ cells constituting >95% of B220$^+$ B lymphocytes, CD11b$^+$ myeloid cells, and NK1.1$^+$ natural killer cells (Figure S9A). As expected from radio-resistant host T cell survival[48], approximately $25.9 \pm 1.8\%$ recipient-derived CD45.1$^+$ T cells remained within the CD3$^+$ T cell population (Figure S9A). Subsequently, Venus-*AKTP$^{M/LOH}$* CRC cells were injected intrasplenically to induce liver metastases in these reconstituted chimeric mice (Figure 7A).

Twelve weeks post-BMT, *Tet2KO* cells exhibited progressive competitive expansion relative to *Tet2WT* cells across all immune cell subsets compared with the four-week post-BMT baseline, with the most pronounced expansion in myeloid cell populations (Figure S9B; Table S25A and S25B). In contrast, control chimeras showed no competitive advantage of *Tet2$^{fl/fl}$* cells over *Tet2WT* competitor cells, maintaining stable ratios throughout the

observation period (Figure S9B; Table S25A and S25B). This clonal expansion pattern

recapitulates the progressive outgrowth of somatic *TET2* mutant-derived CH observed in

human patients[49]. Within liver metastases of *Tet2KO:Tet2WT* chimeras, *Tet2KO* T cells

accounted for approximately 35–46% of total T cells and 23–32% of the CD8$^+$ TIL subset.

These frequencies were comparable to the *Tet2$^{fl/fl}$* CD45.2$^+$ populations observed in controls

(Figure S9C; Table S25C).

Consistent with the restricted LMTB observed in *Vav1^{Cre}/Cd4^{Cre}Tet2KO* mice,

Tet2KO:Tet2WT chimeric mice demonstrated significantly reduced LMTB compared to

Tet2$^{fl/fl}$:Tet2WT controls (Figures 7B and S9D). Flow cytometric analysis of liver TILs

revealed a markedly higher proportion of PD-1$^+$Tim3$^+$ T$_{TEX}$ in both *Tet2$^{fl/fl}$* CD45.2$^+$ and

residual recipient *Tet2WT* CD45.1$^+$ populations compared to that in *Tet2KO* CD45.2$^+$ CD8$^+$

TILs (Figures 7C and 7D; Table S25D). Collectively, these data demonstrate that even within

a competitive environment containing *Tet2WT* counterparts, *Tet2*-deficient CD8$^+$ TILs

exhibit an intrinsic resistance to terminal exhaustion.

Figure 7

Figure 7 *Tet2KO:Tet2WT* **chimeric mouse model of CH demonstrates restricted tumor growth and intrinsic resistance of** *Tet2***-deficient CD8$^+$ TILs to terminal exhaustion in CRC liver metastasis.**

(A) Experiment schema of *Tet2KO:Tet2WT* chimeric mouse model with CRC liver metastasis. Lethally irradiated CD45.1$^+$ WT-recipients were reconstituted with mixtures of *Vav1^{Cre+}Tet2$^{fl/fl}$* CD45.2$^+$ and *Tet2WT* CD45.1$^+$/CD45.2$^+$ BM cells at ratios of 1:2 or 1:3 (KO : WT competitor) to generate *Tet2KO:Tet2WT* mixed chimeras. Control *Tet2$^{fl/fl}$:Tet2WT* chimeras received a 1:1 mixture of *Tet2$^{fl/fl}$* CD45.2$^+$ and *Tet2WT* CD45.1$^+$/CD45.2$^+$ BM cells. Four weeks post-transplantation, Venus-*AKTP$^{M/LOH}$* colorectal cancer cells were intrasplenically injected. Chimerism was assessed in peripheral blood 8 weeks post-injection, followed by analysis of CD8$^+$ TILs and LMTB.

(B) Representative liver metastasis samples and box plot comparing LMTB between *Tet2KO:Tet2WT* chimeric mice ($n = 10$) and *Tet2$^{fl/fl}$:Tet2WT* controls ($n = 12$) ($29.1 \pm 24.1\%$ vs. $10.6 \pm 16.7\%$; $p = 0.0076$). Scale bars: 1 cm.

(C) Representative flow cytometry plots showing the proportion of PD-1$^+$Tim3$^+$ CD8 T$_{TEX}$ in *Tet2$^{fl/fl}$* CD45.2$^+$, host-derived *Tet2WT* CD45.1$^+$, and *Tet2KO* CD45.2$^+$ CD8$^+$ TILs.

(D) Box plots comparing frequencies of PD-1$^+$Tim3$^+$ CD8 T$_{TEX}$ across donor and host CD8$^+$ TIL populations ($n = 13$ per group). See also Table S25D for detailed quantification and statistical comparisons.

CH, clonal hematopoiesis; TILs, tumor-infiltrating lymphocytes; CRC, colorectal cancer; BM, bone marrow; LMTB, liver metastatic tumor burden; T$_{TEX}$, terminally exhausted T cells. Data in (B) and (D) are compiled of at least 2 independent experiments and presented as mean $\pm$ SD. Statistical significance was determined by Mann-Whitney U test. *$p < 0.05$, **$p < 0.01$, ***$p < 0.001$, ****$p < 0.0001$, ns = not significant.

Anti-PD-1 treatment suppresses CRC liver metastasis in Tet2^fl/fl mice

Given the enrichment of exhaustion markers in *Tet2^fl/fl* CD8$^+$ TILs, we evaluated the

therapeutic efficacy of PD-1 blockade (Figure S10A). *Tet2^fl/fl* mice treated with anti-PD-1

antibody exhibited complete tumor eradication, whereas IgG-treated controls developed

profound liver metastatic growth (Figures S10B and S10C). Although IgG-treated

Vav1^Cre Tet2KO mice displayed reduced LMTB compared to *Tet2^fl/fl* controls (IgG-treated

Vav1^Cre Tet2KO vs. IgG-treated *Tet2^fl/fl,* 0.54 ± 0.39 mm2 vs. 23.53 ± 12.64 mm2, *p* = 0.008),

minimal residual tumors persisted (Figures S10B and S10C). This indicates that targeting the

PD-1 axis is sufficient to reverse the metastasis-permissive environment associated with

Tet2-intact hematopoiesis.

DISCUSSION

While CH is highly prevalent among patients with solid tumors, its impact on fundamental

aspects of cancer biology, including tumor growth, treatment response, and metastasis, has

remained poorly defined[21,26,50,51]. In this study, we demonstrate that *TET2*-mutant CH confers

protection against cancer metastasis, extending prior observations that *TET2* inactivation can

enhance anti-tumor immunity in settings such as chimeric antigen receptor T-cell therapy[52,53]

and immune checkpoint blockade[15,33]. By integrating large-scale clinical data with genetic

and functional mouse models, we identify *TET2*-mutant CH as a previously unrecognized

tumor-suppressive state that limits metastatic outgrowth.

In a cohort of over 16,000 solid tumor patients, we show that individuals with *TET2*-

driver clonal hematopoiesis have significantly lower odds of presenting with metastatic

disease and exhibit reduced metastatic burden. Notably, although *TET2*-mutant CH is

generally considered an adverse condition due to its association with increased morbidity and

mortality from atherosclerotic and fibrotic diseases[54-57], we did not observe a difference in

median survival between solid tumor patients without CH and those with *TET2*-CH. This

finding stands in contrast to patients harboring CH driven by mutations in DNA damage

response, cohesin, or spliceosome genes, who experienced significantly worse overall

survival, consistent with prior studies[22,23,27,58]. Together, these data suggest that the biological

consequences of CH are highly driver dependent, and that *TET2*-mutant CH differs

fundamentally from other high-risk CH genotypes in the context of solid tumor

microenvironments.

Our liver metastasis models provide functional evidence supporting a causal

relationship between *TET2*-mutant CH and protection from metastatic disease. Both

complete and chimeric pan-hematopoietic *Tet2* inactivation were sufficient to attenuate

metastatic outgrowth, demonstrating that this association is not merely correlative.

Importantly, we show that *Tet2*-deficient CD8 T cells are necessary and sufficient for the

anti-metastatic effect of *TET2*-mutant CH, linking this phenotype directly to adaptive

immune function. This finding helps reconcile prior heterogeneous reports on the impact of

Tet2-mutant CH in syngeneic tumor models, where tumor growth has been reported as

promoted, suppressed, or unaffected depending on tumor type, implantation site, and

experimental context[29,30,59]. Indeed, our patient-level analyses revealed tumor-type–specific

variation, with the strongest protective association observed for liver and brain metastasis.

Mechanistically, our data established that *Tet2* deficiency restrains metastatic

progression by limiting terminal CD8 T cell exhaustion. Tet2 is transcriptionally induced

following T cell receptor engagement[53,60] and has been shown to act as a negative regulator

of adaptive immunity by catalyzing DNA demethylation at key regulatory loci, thereby

favoring differentiation toward exhausted or regulatory states at the expense of durable

memory[60,61]. In the context of CH, this negative feedback is diminished. Within metastatic

liver lesions, *Tet2*-deficient CD8$^+$ TILs exhibit sustained methylation at regulatory elements

of the *Tox* locus, resulting in transcriptional repression of TOX, a master regulator of terminal

exhaustion[62,63]. This epigenetic constraint preserves progenitor and stem-like CD8 T cell

states and enables sustained effector generation under chronic antigen stimulation.

These findings define a TET2–DNA methylation–TOX axis as a central epigenetic

checkpoint governing exhaustion fate commitment in CD8 T cells. Similar mechanisms have

been implicated in enhanced control of chronic infection and cancer by *Tet2*-mutant CD8 T

cells and *Tet2*-deficient CAR-T cells[53], underscoring the generalizability of this regulatory

pathway. However, unrestrained T-cell activation associated with *TET2*-mutant CH may also

be pathological in certain contexts, including autoimmunity[64,65], immune-related toxicities

during immunotherapy[66], or malignant transformation following acquisition of additional

mutations[67,68]. Indeed, *TET2*-mutant CH is a recognized risk factor for T cell lymphoma[69] ,

highlighting the dual-edged nature of sustained immune activation.

In addition to its effects on T cells, *Tet2* loss has been shown to enhance antigen-

presenting capacity in macrophages[15]. While myeloid-restricted *Tet2* inactivation was

insufficient to suppress liver metastasis in our models, it may still contribute to improved

tumor control in concert with CD8 T cells. Consistent with this notion, we and others have

previously observed transcriptional signatures of heightened T-cell activation in colorectal

cancer and melanoma samples from patients with *TET2*-mutant CH[15,70]. Defining the relative

contributions of distinct immune lineages to other CH-associated clinical phenotypes remains

an important area for future investigation.

Several limitations of this study should be acknowledged. We were unable to directly

assess whether tumor-infiltrating *TET2*-mutant T lymphocytes correlate with metastatic

status in patients, as such data were not available. Although *TET2* is a well-established

negative regulator of CD8 T cell activity[53,60,71], the frequency and functional relevance of

TET2-mutant CD8 T cells within human tumors remain incompletely characterized.

Moreover, our experimental models focused on experimental liver metastasis and therefore

do not address potential effects of *TET2*-mutant CH on primary tumor growth, metastatic

dissemination, or circulation through blood and lymphatic compartments. While the *Cd4Cre*

model demonstrates that T cells are necessary and sufficient for the observed phenotype, the

potential contribution of other *Tet2*-deficient immune cells warrants further study. Finally,

the causes of death in solid tumor patients with high-risk CH are not known and may be

unrelated to metastatic disease.

In summary, we show that *TET2*-mutant CH suppresses cancer metastasis in both

patients and animal models by enforcing an epigenetic barrier to terminal CD8 T-cell

exhaustion. These findings reposition *TET2*-mutant CH as a protective modifier of tumor

progression and suggest that CH status may have utility as a prognostic biomarker for

metastatic risk, at least in certain cancers. Notably, the magnitude of metastatic suppression

conferred by *TET2*-mutant CH was substantial in our experimental models, approaching the

level of tumor control observed with immune checkpoint blockade, suggesting that

pharmacologic strategies designed to mimic hematopoietic *TET2* inactivation—or to target

downstream components of the TET2–TOX axis—could represent novel therapeutic

approaches for the prevention or treatment of metastatic disease.

METHODS

Experimental model and subject details

Human study cohort

Patients who were tested for clonal hematopoiesis using an MSK-IMPACT targeted

sequencing panel with somatic mutations from peripheral blood sequencing results curated

as putative drivers or non-drivers of CH in the Bolton *et al*. 2020 cohort who were also

evaluated for metastatic disease by Nyugen *et al*. 2022 were included in our sub-cohort

analysis. Study data containing anonymous patient identifiers were downloaded from

cBioPortal on September 4th, 2025. Using variants reported by Bolton *et al.* 2020, clonal

hematopoiesis (CH) was called as a putative driver mutation with variant allele fraction

(VAF) $\geq$ 2%. Each mutation was defined as either putative (PD) or non-putative based on its

role in cancer pathogenesis and recurrence in an MSKCC in-house dataset of myeloid

neoplasms as in Bolton *et al.* 2020. It was further was defined as either a non-myeloid PD or

myeloid PD based on relevance to myeloid neoplasms specifically.

Mice

Vav1[Cre36], *Cd4[Cre37]*, *Cd19[Cre38]*, and *LysM[Cre39]* transgenic mice were crossed with *Tet2[fl/fl]*

mice[35] to generate homozygous *Tet2*-knockout mice (*Vav1[Cre]Tet2KO, Cd4[Cre]Tet2KO*, and

LysM[Cre]Tet2KO). *Tet2[fl/fl]* mice served as *Tet2* wild-type control. All mice were maintained in

a specific pathogen-free animal facility in the Laboratory of Animal Center, University of

Tsukuba. All animal procedures adhered to the The Regulations for Animal Experiments in

University of Tsukuba. Mouse genotyping was performed using genomic DNA extracted

either from tails by KAPA kit method (KAPA Biosystems) or from blood by QIAamp DNA

Blood Mini Kit (Qiagen). The primers used for genotyping are listed in Table S26.

Organoid cell line

Venus-labeled mouse intestinal tumor-derived organoid-originated cells simultaneously

carrying *Apc[Δ716]*, *Kras[+/G12D]*, *Tgfbr2[-/-]*, and *Trp53[R270H/LOH]* mutations (Venus-

AKTP[M/LOH] cells) were established as previously described[40]. Venus-*AKTP[M/LOH]* cells were

cultured in Advanced DMEM/F-12 medium (Gibco, #2508872) supplemented with 10%

Fetal Bovine Serum (FBS, Hyclone, #AE25787263), 1% Penicillin-Streptomycin (PS,

Fujifilm, #168-23191), 0.3 μg/ml Rock inhibitor (Y27632, Fujifilm, #235-00513), 0.9 μg/ml

GSK-3 inhibitor (CHIR99021, R&D system, #4423), and 21 μg/ml ALK inhibitor (A8301,

R&D systems, #5640). For the passage, Venus-*AKTP[M/LOH]* cells were enzymatically

dissociated with 0.25% Trypsin at 37°C for 10 minutes and mechanically dissociated by

pipetting.

Method details

Clinical cohort data processing and statistical analysis

(1) Clonal hematopoiesis mutation grouping

Patients were categorized as having CH if they had a variant annotated as CH PD in Bolton

et al. 2020[22]. Patients were then categorized as having a *TET2* or *DNMT3A* PD only if they

had no other CH-defining PD variants. Using a mutually exclusive hierarchical system,

patients were first categorized as having a *TET2* driver, followed by a *DNMT3A*, cohesin

(*STAG2, RAD21, SMC1, SMC3, PDS5A, PDS5B*), spliceosome (*SF3B1, SRSF2, ZRSR2,*

U2AF1), or DNA-damage-response driver (*PPM1D, CHEK2, TP53, ATM*). For example, if

a patient had both a cohesin (e.g. *STAG2*) and spliceosome (*SF3B1*) PD variant, they would

be categorized in the cohesin group. A patient with a *TET2* and *STAG2* PD variant would

also be categorized in the cohesin group.

(2) Quantification of cancer metastatic burden

Patients were labelled as metastatic patient or non-metastatic patient by Nguyen *et al.*. For

each patient with metastatic disease, the number of metastases and metastatic sites was

quantified. Patients were assigned as positive or negative for cancer metastases to various

organs including lung, liver, draining lymph node (DLN), brain, and bone.

(3) Clinical cohort statistical analysis

The odds of metastatic disease stratified by each PD variant group was assessed using

multivariate logistic models to account for covariates that have previously been linked either

to CH prevalence or cancer outcome, including chemotherapy and radiation treatment status,

age, sex, smoking status, race, and cancer type. Similar models were used to assess odds of

metastasis within cancer types for *TET2*-driver CH. However, due to few occurrences of non-

metastatic cancer in *TET2*-driver CRC patients and prostate cancer patients and limitations

of multivariate analysis, Fisher's exact test for count data was applied in addition to the

logistic model to ensure robust statistical testing. (Table S1). For odds of metastasis to

specific sites, one model for each metastatic site allowed for patients with metastases at

multiple sites (such as brain and bone) to be classified as positive for all metastatic sites.

To assess metastatic burden, Wilcoxon rank-sum tests were used to quantify

differences in distribution of both number of metastases, and number of metastatic sites.

Further, multivariate negative binomial generalized linear models were used to account for

the same covariates as above.

Patients were categorized in a mutually exclusive manner according to the criteria

outlined above to investigate associations between PD variant groups and survival. Kaplan-

Meier survival curves and a corresponding Cox proportional hazards model were used to

account for covariates previously linked to survival, including age, sex, race, cancer type,

smoking status, chemotherapy and radiation treatment status, and metastatic status.

Difference in distribution of age between CH driver patients and non-CH driver

patients was quantified using a multivariate linear model to account for covariates associated

with age, including age, sex, race, cancer type, smoking status, chemotherapy and radiation

treatment status, and metastatic status. The association between *ASXL1* CH driver mutation

prevalence and smoking status was assessed using a multivariate logistic model, as were the

associations between DNA-damage-response CH driver mutation prevalence with treatment

status, and metastatic disease with treatment status. Odds of metastatic disease in patients

with a *TET2* CH driver mutation as opposed to a *TET2* CH non-driver mutation was also

assessed using a multivariate logistic model.

All statistical tests are two-tailed, and analyses in which multiple comparisons were

made were corrected for multiple hypothesis testing in a false discovery rate (FDR)

controlled fashion. An FDR-controlled *P* value threshold of 0.05 was used to denote

significance.

Venus-AKTP$^{M/LOH}$ cell injection, liver metastasis tumor burden (LMTB) evaluation

Venus-*AKTP$^{M/LOH}$* cells were used for the mouse spleen injection after at least four times of

passage. One million (1x10^6) of thoroughly dissociated Venus-*AKTP$^{M/LOH}$* cells resolved in

30 μl of Matrigel Matrix (Corning) were injected into mouse spleen. Mouse livers were

harvested 30 days post-injection and imaged as whole organs using the fluorescence stereo

microscope equipped with the digital color microscope camera (M165 FC/DFC450C, Leica

Microsystems) at 4.6x magnification to macroscopically assess the presence of Venus-

AKTP$^{M/LOH}$ cell metastasis. LMTB was evaluated as previously described[72]. Briefly,

formalin-fixed paraffin-embedded liver sections were stained with hematoxylin and eosin

(H&E). The tumor burden was quantified by calculating the ratio of the total tumor area to

the total liver section area using a Keyence BZ-X710 microscope and BZ-X analyzer

software The two-tailed Mann-Whitney U test with α = 0.05 was used to compare the size

ratio between the two groups.

Antibody-based depletion of CD8 T cells

Anti-CD8α antibody treatment was initiated two days prior to intrasplenic injection of Venus-

AKTP^M/LOH cells. InVivoMAb anti-mouse CD8α antibody (clone 2.43, BioXCell, #BE0061)

was diluted with Phosphate-Buffered Saline (PBS) to the concentration of 1μg/μL and

intraperitoneally injected at an initial dose of 200 μg/mouse, followed by 6 booster injections

of 100 μg/mouse every 4 days throughout the tumor progression course. Control mice

received PBS in volumes matching those used in the anti-CD8α antibody treatment schedule.

Mice were sacrificed 30 days after the injection of Venus-*AKTP^M/LOH* cells to collect the

whole liver for LMTB evaluation.

Enzymatic digestion of liver metastases and isolation of tumor-infiltrating immune cells

for flow cytometry

Mouse livers were harvested 30 days post-injection. Liver metastatic tumors and surrounding

liver tissues 0.5 cm from the edge of tumors were minced into small pieces < 1mm^3. The

minced tumor tissue was then digested by 15 ml of digestion buffer containing 0.75 mg/ml

collagenase A (Sigma-Aldrich, #10103578001) and 50 μg/ml DNase I (Worthington

Biochemical, #LS002139) in DMEM (Gibco, #10567-014), and incubated at 37 °C with

shaking at 170 rpm for 30 mins. After tissue digestion, the cell suspension was first filtered

through a 70-μm cell strainer (Falcon) to remove larger debris. The resulting homogenate

was then further filtered through a 30-μm cell strainer (Miltenyi Biotec) to obtain a single-

cell suspension. Any remaining undigested lumps were further dissociated by gentle pipetting.

DMEM supplemented with 10% FBS (Hyclone, #AE25787263) and 1% Penicillin-

Streptomycin (PS, Fujifilm, #168-23191) was added to terminate the digestion. The cell

suspension was centrifuged at 163× *g* for 7 mins at 4 °C, and then the cell pellet was

resuspended in 5 ml of 40% Percoll (Cytiva, #17089102) in PBS. The cell suspension was

then slowly and carefully dropped on top of 3 ml of Histopaque-1083 (Sigma-Aldrich,

#10831) without mixing and centrifuged at 1000× *g* for 20 mins in brake off mode. The cells

concentrated at the interface layer were carefully aspirated and added into 30 ml of FACS

buffer (PBS containing 2% Bovine Serum Albumin (BSA) and 2 mM EDTA) and centrifuged

at 800× *g* for 5 mins at 4 °C. To detect

exhausted T cells, the yielded cell pellet was resuspended in FACS buffer and

incubated with a panel of fluorophore-conjugated anti-mouse antibodies consisting of anti-

CD45, anti-CD3e, anti-CD4, anti-CD8α, anti-PD-1, and anti-Tim3 (Havcr2/CD366)

antibodies (Table S27A) before being analyzed by BD FACSAria III (BD Biosciences).

Bulk-RNA sequencing

Mouse livers were harvested 30 days post-injection. Tumor-infiltrating leukocytes yielded

by the same procedure as that used for flow cytometry were utilized for the CD45-positive

leukocyte selection. 10 μL of CD45 microbeads (Miltenyi Biotec, #130-052-301) was mixed

well and incubated with each $1x10^7$ cells in 90 μL of FACS buffer for 30mins at 4°C. The

CD45-positive selection was then conducted using the MACS LS column and the

MidiMACS separator (Miltenyi Biotec), following the column-based magnetic cell-

separation strategy established by the manufacturer. CD4$^+$ TILs and CD8$^+$ TILs were

identified using a panel of fluorophore-conjugated anti-mouse antibodies consisting of CD45,

CD3e, CD4, and CD8α, and isolated by BD FACSAria III Cell Sorter (BD Biosciences).

Total RNA was isolated using RNeasy Micro Kit (Qiagen). RNA integrity number

(RIN) values were measured using the Agilent 2100 Bioanalyzer (Agilent Technologies).

Samples with RIN > 7 were subsequently used for library preparation using the SMARTer

Stranded Total RNA-seq Kit v2 - Pico Input Mammalian (Takara Inc.) according to the

manufacturer's protocol. The library size distribution and concentration were assessed with

the High Sensitivity DNA assay using the Agilent 2100 Bioanalyzer (Agilent Technologies).

Sequencing was carried out using the Illumina HiSeq X platform (Macrogen) with a standard

150 bp paired-end read protocol. Subsequently, the gene expression was analyzed from the

sequencing data using the CLC workbench version 11.0.1 (Qiagen) according the procedure

described previously[73]. DEGs were identified by the Differential Expression for RNA-Seq

tool of CLC workbench using the Wald test applying for the Against-control-group

comparison with the *Tet2^{fl/fl}* group serving as the control. The gene expression datasets were

then used to identify significantly enriched gene sets for each group using GSEA[42] (GSEA

4.3.0, https://www.gsea-msigdb.org/gsea/index.jsp) with the ImmuneSigDB subset of C7

(4872 gene sets, https://www.gsea-msigdb.org/gsea/msigdb/human/collections.jsp) for

annotation.

Upstream regulator analysis

Overlapping downregulated DEGs within CD8$^+$ TILs from *Vav1^{Cre}Tet2KO* and

Cd4^{Cre}Tet2KO groups compared to the *Tet2^{fl/fl}* group (FDR< 0.05, Log2 FC < -1; Table S8)

were used to predict the upstream transcription factors by the Upstream Regulator Analysis

function of QIAGEN's IPA[74] (QIAGEN Redwood City, www.qiagen.com/ingenuity). The

pathway analysis was performed using the core analysis based on gene expression fold

change[74]. The upstream transcription regulators were called with overlap p-value < 0.01. The

extracted upstream regulators were then used to generate the molecular interaction network

by the Network and Pathway tools of IPA.

Indirect immunofluorescence staining, co-expression cell evaluation

Liver metastasis tissue collected from $Vav1^{Cre}Tet2KO$, $Cd4^{Cre}Tet2KO$ and $Tet2^{fl/fl}$ groups

were mounted in O.C.T embeding compound (Sakura Finetek Japan Co.), frozen rapidly in

hexane-dry ice and stored at -80°C. The frozen tissue blocks were then sectioned at 4 μm-

thickness in a cryostat at -20°C. Sections were dried up in air for 60 mins at room temperature

(RT), fixed in 4% paraformaldehyde for 15 mins and permeabilized in 0.1% Triton X100 for

10 mins and incubated with protein blocking buffer (Protein block, Abcam, #ab64226).

Sections were then incubated with primary antibodies at 4°C for 16 hours, washed 3 times in

PBS for 5 mins each and subsequently incubated with secondary antibodies for 30 mins at

RT. Antibodies and the dilution used for immunofluorescence staining were provided in the

Table S27D. After three washes, tissue sections were incubated with True VIEW

Autofluorescence Quenching Kit (Vector Laboratories, #SP-8400-15) for 5 mins at RT and

mounted with 4',6-diamidino-2- 129 phenylindole (DAPI, Vectashield Mounting Medium;

Vector Laboratories, #H-1200).

The sections were imaged using the confocal laser scanning microscope (Leica TCS

SP8, Leica Microsystems). To quantify exhaustion phenotypes, CD8$^+$ TILs co-expressing

PD-1, Tim3, or Tox were manually counted in 10 to 20 random intratumoral fields per sample

at 40× magnification. Data were expressed as the percentage of each subset relative to the

total CD8$^+$ TIL population. The one-way ANOVA test with $\alpha = 0.05$ was used to compare

the percentage of co-expressing cells between groups.

Single-cell RNA sequencing (scRNA-seq) library preparation, data acquisition and pre-

processing

Liver samples from *Vav1^{Cre}Tet2KO* mice and *Tet2$^{fl/fl}$* mice were harvested at day 30 post-

injection, and tumor-infiltrating leukocytes were harvested using the cell processing

procedure that used for flow cytometry. The yielded cells were washed and resuspended in

PBS supplemented with 1% BSA. Cells were then incubated with Brilliant Violet 421

(BV421) anti-mouse CD45.2 antibody (clone 104, Biolegend, #109832) for 20 minutes at

4°C, followed by staining with 7-AAD viability solution (ThermoFisher Scientific, 00-6993-

50; 1:20) for 10 minutes on ice. CD45-positive live cells were subsequently sorted using the

BD FACSAria III (BD Biosciences). Isolated CD45+ live cells were processed to generate

barcoded 5′ gene expression (5′ GEX) libraries, together with T-cell receptor (TCR)-enriched

libraries, using the Chromium Next GEM Single Cell V(D)J Reagent Kits v1.1 (10x

Genomics, Pleasanton, CA, USA). The procedure followed the manufacturer's instructions

(CG000208 Rev G) and aimed to capture 10,000 cells per library. All libraries were subjected

to quality 150 control checks and quantified using the 2100 Bioanalyzer High Sensitivity

DNA Kit (Agilent 151 Technologies) and KAPA Library Quantification kit for Illumina

platforms (Kapa Biosystems, KK4873). The library were sequenced on the Illumina

NovaSeqX Plus Sequencing System (Macrogen), then mapped to the mouse genome

(GRCm39) and demultiplexed using CellRanger (10X Genomics, v.8.0.1).

scRNA-seq data processing of individual samples

Matrix data of each samle was processed using Seurat[75] (v.5.1.0) in RStudio (v.4.4.1).

Doublets were detected and removed using the "scDblFinder" function from the scDblFinder

package[76] (v.1.18.0) with the default estimated doublet rate. Cells with fewer than 200 or

more than 2,500 unique genes and those with over 5% of mitochondrial gene reads were

excluded. After removing ribosomal and mitochondrial genes, gene expression count data

was normalized using the "NormalizeData" function. Normalized data were linearly

transformed (scaled) with the "ScaleData" function and subjected to principal component

analysis (PCA) via the "RunPCA" function. Graph-based clustering was performed based on

gene expression profiles using the "FindNeighbors" and "FindClusters" functions. Clusters

were visualized using Uniform Manifold Approximation and Projection (UMAP) technique

via the "RunUMAP" and "DimPlot" functions.

Cell clusters were annotated as T lymphocytes, B lymphocytes, NK cells, and

myeloid cells based on marker genes identified by the "FindAllMarkers" and

"FindConservedMarkers" functions. For a gene to be considered a marker, a minimum of

25% gene-expressing cells per cluster was required, and the minimum Log2 FC in gene

expression of each cluster, compared to all other clusters, was set to 0.25.

Data integration for scRNA-seq and batch correction

Canonical correlation analysis (CCA) was performed to identify shared features across

multiple gene expression datasets using the FindIntegrationAnchors function of Seurat. Four

datasets (data from tumor-infiltrating leukocytes of $Tet2^{fl/fl}$ ($n = 2$) and $Vav1^{Cre}Tet2KO$ ($n =$

2) mice) were integrated based on the identified anchors using the IntegrateData function

with 30 canonical correlation dimensions. The integrated data were scaled, subjected to PCA,

and used for graph-based clustering for cell annotation, as described above.

scRNA-seq subclustering and cell-type annotation

CD8$^+$ TIL clusters were identified and extracted from the integrated data, followed by scaling,

PCA-based dimensionality reduction, and unsupervised graph-based subclustering for cell

annotation. Subclusters that highly expressed marker genes characteristic of other immune

cell subsets, distinct from CD8$^+$ TILs, were excluded. Following in silico sorting, the CD8

T-cell dataset comprised 2,251 $Vav1^{Cre}Tet2KO$ CD8$^+$ TILs and 1,101 $Tet2^{fl/fl}$ CD8$^+$ TILs. To

minimize discrepancies in cell numbers between the groups, downsampling was performed,

resulting in a balanced dataset of 1,101 CD8$^+$ TILs per group. The downsampling process

was repeated multiple times to ensure consistency with the original data. Cell subclusters

were annotated by analyzing the expression levels of canonical gene markers using Seurat's

"FindAllMarkers" and "FindConservedMarkers" functions.

DEG analysis of scRNA-seq

DEG analysis between $Tet2^{fl/fl}$ and $Vav1^{Cre}Tet2KO$ conditions was performed using

"FindMarkers" function of Seurat. Genes were considered differentially expressed if they

met the following criteria: (i) expressed in a minimum of 10% of cells per cluster within at

least one condition, and (ii) exhibited a Benjamini–Hochberg adjusted p-value (adjusted p-

value) less than 0.05.

Analysis of single-cell gene set activity

We used the gene set comprising of 42 consistently exhausted T cell associated marker genes

identified by Zhang[45] to calculate the T-cell exhaustion score for single cell $CD8^+$ TIL dataset

using "AddModuleScore" function. The number of bins of aggregate expression levels for

all analyzed features was set to 24 and number of control features selected from the same bin

per analyzed feature was set to 100. The differential enrichment of T-cell exhaustion scores

between *Tet2$^{fl/fl}$* and *Vav1^{Cre}Tet2KO* groups was visualized using boxplots, and statistical

significance was assessed via the Mann-Whitney U test.

We used the AUCell package[44] v.1.26.0 to analyze the Memory versus Exhausted

CD8 T-cell signature enrichment, and Effector versus Exhausted CD8 T-cell signature

enrichment   for $CD8^+$ TIL   dataset.   The   gene   set

"*GSE9650_EXHAUSTED_VS_MEMORY_CD8_TCELL_DN*"   and   "*GSE9650*

_EXHAUSTED_VS_MEMORY_CD8_TCELL_UP" were obtained from the Human MSigDB

(v2024.1.Hs) and converted to mouse orthologs using the 'orthologs" function from the

babelgene package (v.22.9). Gene expression rankings were calculated for each cell using

the "AUCell_buildRankings" function, followed by computation of the area under the curve

(AUC) scores of every cells for each gene set using the "AUCell_calcAUC" function. To

create an aggregate Memory versus Exhausted CD8 T-cell signature, the AUC values of the

upregulated gene set were subtracted from those of the downregulated gene set. Similarly, to

calculate the aggregate Effector versus Exhausted CD8 T-cell signature enrichment, we

subtracted the AUC values of the downregulated gene set from those of the upregulated gene

set using the gene sets "GSE9650_EFFECTOR_VS_EXHAUSTED_CD8_TCELL_DN"

and "GSE9650_EFFECTOR_VS_EXHAUSTED_CD8_TCELL_UP" obtained from the

Human MSigDB (v2024.1.Hs). The differential enrichments of these gene signatures

between $Tet2^{fl/fl}$ and $Vav1^{Cre}Tet2KO$ groups were visualized using boxplots, and statistical

significance was assessed via the Mann-Whitney U test.

Single-cell trajectory inference

We used *Monocle 3 toolkit*[46] (v.1.3.7) to construct the cellular pseudotime trajectories for

CD8 T-cell subsets. Single cell datasets were preprocessed using PCA, retaining 50

dimensions. Dimensionality reduction was then performed using UMAP technique. Cells

were clustered using the "cluster_cells" function with the Leiden algorithm, setting the

resolution parameter to 0.008, the number of nearest neighbors (k) to 20, and performing two

iterations. A principal graph representing cellular transitions was learned using the

"learn_graph" function with default heuristics and without partitioning the data. According

to previous knowledge, we selected the root cell cluster to order single cells along the

trajectory using the "order_cells" function. Single cell pseudotime and trajectory were

visualized using the "plot_cells" function.

TCR-seq data analysis

Single cell TCR-seq contig data from all samples were combined and annotated into

clonotypes based on shared nucleotide sequences, amino acid sequences, and V(D)J gene

segment usage using the "combineTCR" function of scRepertoire package[47] (v.2.3.0).

Clonotypes were defined using the combination of V(D)J genes (TCR/Ig) and the nucleotide

sequence of the CDR3 region (strict clonotype calling) throughout the analysis.   The

clonotype data was integrated with gene expression datasets for clonal frequency calculation

using the "combineExpression" function. TCR clonotype similarity across cell clusters was

analyzed using the "clonalOverlap" function with the Morisita overlap index[47].

Clonal diversity was quantified using Shannon entropy and the inverse Simpson

index[77,78]. Clonal inequality was assessed using the Gini coefficient and Lorenz curve

analysis[79]. All metrics were calculated from TCR clonotype frequencies using R (v.4.4.1) in

RStudio.

Whole genome bisulfite sequencing (WGBS)

PD-1$^+$CD8$^+$ TILs isolated from the liver metastatic tumors of *Vav1^{Cre}Tet2KO* ($n = 3$) and

Tet2$^{fl/fl}$ ($n = 3$) mice were used for library preparation. Genomic DNA for WGBS was

extracted from sorted cells using the QIAamp DNA Blood Mini Kit (Qiagen). The library for

WGBS was prepared using the postbisulfite adaptor tagging (PBAT) protocol as described

previously[80,81]. After library preparation, library yields were quantified using the library

quantification kit (Takara Bio Inc.) and amplified using PrimeStar Max DNA polymerase

(Takara Bio Inc.). The amplified libraries were sequenced on the Illumina HiSeq X platform

at Macrogen.

WGBS data were aligned to the *Mus musculus* reference genome (GRCm39) using

the Bismark pipeline[82]. Differentially methylated region (DMR) analysis was performed

using *methylKit* package (v.1.30.0). For the genome-wide analysis, DMRs of *Tet2KO* PD-

1^+CD8^+ TILs in comparison to *Tet2^{fl/fl}* PD-1^+CD8^+ TILs were detected by scanning non-

overlapping 1000bp-windows across the whole genome locus using the calculateDiffMeth

and getMethylDiff functions, applying a q-value < 0.01 and |meth.diff| ≥ 20%. For DMR

analysis focusing *Tox* gene, DMRs were detected by screening 500bp-windows at 250bp-

step, applying the same statistical thresholds as in the global analysis. Differentially

methylated cytosines (DMCs) within the *Tox* locus were identified at single-CpG resolution

with the same q-value and methylation difference cutoffs. DMRs were annotated using the

annotatePeakInBatch function from the ChIPpeakAnno package (v3.38), with mouse gene

annotations from the Gencode vM35 primary assembly and the ENCODE Registry of cCREs

(v3, 2021) serving as reference. DMRs overlapping enhancer, promoter, CTCF, or TSS

regions were classified as regulatory DMRs (rDMRs).

To quantify the combined regulatory impact of differential methylation on gene

expression, we developed a composite methylation impact score that accounts for both the

breadth (number of affected regulatory elements) and depth (magnitude of methylation

changes) of regulatory disruption. The impact score for each gene was calculated as:

Impact Score = (Total regulatory DMRs) × (Mean absolute methylation difference)

Where:

874   Total regulatory DMRs: the number of differentially methylated regions

875   (qvalue < 0.01, |methylation difference| ≥ 20%) overlapping with any regulatory

876   element (promoter, TSS, enhancer, or CTCF binding site) associated with the gene.

877   Mean absolute methylation difference: the average magnitude of methylation

878   change (%) across all regulatory DMRs for that gene.

879   This scoring approach identifies genes with extensive regulatory methylation

changes that are likely to have substantial effects on transcriptional control. Genes with

higher scores have both multiple affected regulatory elements and large-magnitude

methylation changes.

***In vivo* anti-PD-1 blockade**

Anti-PD-1 antibody treatment was started seven days after the intrasplenic injection of

Venus-*AKTP$^{M/LOH}$* cells. InVivoMAb anti-mouse PD-1 (RMP1-14, Bio X Cell, #BE0146)

was diluted with PBS to a final concentration of 1 μg/μL and intraperitoneally injected at a

dose of 10 μg/g of mouse body weight every 3 days for 6 doses in total. Control mice received

InVivoMAb rat IgG2a isotype control (2A3, Bio X Cell, #815022F1) following the same

dilution and dosage as the anti-PD-1 antibody treatment protocol. Mice were sacrificed 30

days after the injection of Venus-*AKTP$^{M/LOH}$* cells to collect the whole liver for LMTB

evaluation.

Competitive bone marrow transplantation (BMT) mouse model with CRC liver

metastasis

Bone marrow isolation and preparation: Bone marrow cells were harvested from femurs and

tibias of 8-10 week-old donor mice. Bones were flushed with ice-cold PBS containing 2%

FBS using a 23-gauge needle. Cell suspensions were filtered through 70-μm cell strainers.

Viable cell counts were determined by trypan blue exclusion.

Lethal irradiation and competitive transplantation: Recipient CD45.1$^+$ C57BL/6J

mice (8-12 weeks old) were lethally irradiated using an X-ray irradiator (Hitachi Power

Solutions MBR-1520R-3) with a split-dose protocol of 4.5 Gy twice, administered 4 hours

apart (total dose of 9.0 Gy) while housed in pie cages to ensure uniform radiation exposure.

For *Tet2KO:Tet2WT* chimeras, *Vav1Cre$^+$Tet2$^{fl/fl}$* CD45.2$^+$ bone marrow cells were mixed

with *Tet2WT* CD45.1$^+$/CD45.2$^+$ competitor bone marrow cells at ratios of 1:2 or 1:3. Control

Tet2$^{fl/fl}$:Tet2WT chimeras received a 1:1 mixture of *Tet2$^{fl/fl}$* CD45.2$^+$ and *Tet2WT*

CD45.1$^+$/CD45.2$^+$ bone marrow cells. A total of 2×10^6 bone marrow cells in 200 μL PBS

were injected via retro-orbital venous plexus within 4-6 hours post-irradiation. Mice were

maintained in sterilized caging, food, and water were provided during the first 14 days after

transplantation.

Chimerism was assessed in peripheral blood collected via facial vein bleeding at 4

weeks and re-evaluated at 12 weeks post-BMT to confirm the proliferative advantage of

Tet2KO hematopoietic and lymphoid compartments. Donor cell engraftment was quantified

by flow cytometry analysis of CD45.1 and CD45.2 expression.

After confirming adequate hematopoietic reconstitution at four weeks post-BMT,

Venus-*AKTP^{M/LOH}* cells were injected intrasplenically to establish liver metastases. Eight

weeks after cancer cell injection, mice were euthanized and livers were harvested and

processed for histological analysis. LMTB was quantified by microscopic measurement as

described above. Tumor-infiltrating immune cells were isolated from liver metastases by

enzymatic digestion as described above and subsequently analyzed by flow cytometry to

identify exhausted T-cell populations using a panel of fluorophore-conjugated anti-mouse

antibodies, consisting of anti-CD3ε, anti-CD4, anti-CD8α, anti-PD-1, anti-Tim3

(Havcr2/CD366), anti-CD45.1, and anti-CD45.2 antibodies (see Table S27B).

***In vitro* modeling of CD8 T-cell exhaustion**

CD8 T cells were isolated from spleens of *CD4^{Cre}Tet2KO* mice, and *Tet2^{fl/fl}* mice using the

Mouse CD8 T Cell Isolation Kit (Miltenyi Biotec, # 130-095-236) according to the

manufacturer's instructions. Purity (>90% CD45$^+$CD3$^+$CD8$^+$ cells) was confirmed by flow

cytometry.

Purified CD8 T cells were seeded at 2×10^7 cells per well in 6-well plates and activated

with plate-bound anti-CD3ε antibody (2 µg/mL, clone 145-2C11, BioLegend, #100340) and

soluble anti-CD28 antibody (2 µg/mL, clone 37.51, BioLegend, #102116) in complete RPMI

1640 medium supplemented with 10% FBS, 1% penicillin-streptomycin, 50 µM β-

mercaptoethanol, 10 mM HEPES, and recombinant mouse IL-2 (125 U/mL, PeproTech,

#212-12). After 4 days, T cells were harvested and transferred to fresh wells pre-coated with

anti-CD3 antibody alone (2 µg/mL) for chronic re-stimulation. This re-stimulation was

repeated on day 8. T cells were cultured for a total of 12-14 days at 37°C with 5% CO_2, with

medium containing IL-2 (125 U/mL) refreshed every 2-3 days.

At indicated timepoints (days 7 and 12), T cells were harvested and assessed for

exhaustion markers by flow cytometry using panels of fluorophore-conjugated anti-mouse

antibodies, consisting of anti-CD3ε, anti-CD8α, anti-PD-1, anti-Tim3 (Havcr2/CD366), and

anti-Tcf1 antibodies (see Table S27C). For intracellular staining, cells were first stained with

Fixable Viability Dye eFluor 780 (eBioscience, #65-0865-14), followed by cell-surface

markers. Cells were then fixed and permeabilized using the Foxp3/Transcription Factor

Staining Buffer Set (eBioscience, #00-5523-00) according to the manufacturer's instructions,

and stained with intracellular antibodies for 30 minutes at room temperature. Data were

acquired on BD FACSAria III and analyzed using FlowJo software (BD Biosciences).

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Declaration of Interests

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competing interests.

Declaration of generative AI and AI-assisted technologies

During manuscript preparation, generative AI tools were used solely for language editing and

clarity improvement. The authors reviewed and validated all content and take full

responsibility for the manuscript.

RESOURCE AVAILABILITY

Lead contact

Further information and requests for resources and reagents should be directed to and will be

fulfilled by the lead contact, Mamiko Sakata-Yanagimoto (sakatama@md.tsukuba.ac.jp).

Materials availability

This study did not generate new unique reagents.

Data and code availability

Statistical source data and detailed results for human cohort analyses (Figure 1 and Figure

S1) are provided in Table S1 and Table S2. All original code will be deposited at GitHub and

made publicly available as of the date of publication. The code for scRNA-seq and WGBS

analysis will be available at: [Link will be provided upon publication]. The code for clonal

hematopoiesis   correlation   analysis   is   available   at:

https://github.com/GiacomoCampo/msk_ch_correlations.   Any   additional   information

required to reanalyze the data reported in this paper is available from the lead contact upon

request.

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