RESEARCH ARTICLE

Transposon mutagenesis identifies cooperating genetic drivers during keratinocyte transformation and cutaneous squamous cell carcinoma progression

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Abstract

The systematic identification of genetic events driving cellular transformation and tumor progression in the absence of a highly recurrent oncogenic driver mutation is a challenge in cutaneous oncology. In cutaneous squamous cell carcinoma (cuSCC), the high UV-induced mutational burden poses a hurdle to achieve a complete molecular landscape of this disease. Here, we utilized the Sleeping Beauty transposon mutagenesis system to statistically define drivers of keratinocyte transformation and cuSCC progression in vivo in the absence of UV-IR, and identified both known tumor suppressor genes and novel oncogenic drivers of cuSCC. Functional analysis confirms an oncogenic role for the ZMIZ genes, and tumor...
Author summary

Non-melanoma skin cancers, the most common cancers in the US, are caused by UV skin exposure. Nearly 1 million cases of cutaneous squamous cell carcinoma (cuSCC) are diagnosed in the US each year. While most cuSCCs are highly treatable, more than twice as many individuals die from this disease as from melanoma. The high burden of UV-induced DNA damage in human skin poses a challenge for identifying initiating and cooperating mutations that promote cuSCC development and for defining potential therapeutic targets. Here, we describe a genetic screen in mice using a DNA transposon system to mutagenize the genome of keratinocytes and drive squamous cell carcinoma in the absence of UV. By sequencing where the transposons selectively integrated in the genomes of normal skin, skin with pre-cancerous lesions and skin with fully developed cuSCCs from our mouse model, we were able to identify frequently mutated genes likely important for this disease. Our analysis also defined cooperation between sets of genes not previously appreciated in cuSCC. Our mouse model and ensuing data provide a framework for understanding the genetics of cuSCC and for defining the molecular changes that may lead to the future therapies for patients.

Introduction

Cutaneous squamous cell carcinoma (cuSCC) is the second most common cancer in man, with approximately one million cases diagnosed annually in the United States. Although the majority of cuSCC are considered a low-risk neoplasm, up to 5% of high-risk cuSCCs are locally or distantly invasive and carry a poor prognosis due to a lack of biomarkers, therapeutic targets, or FDA-approved molecularly targeted therapies. This represents a substantial unmet need for approximately 50,000 patients per year with high-risk cuSCC, and an opportunity to identify new therapeutic modalities that could improve disease outcomes. All non-viral associated skin cancers are thought to require multiple cooperating mutations that deregulate distinct signaling pathways to initiate and progress the multi-step transformation of normal cells into a clinically significant neoplasm. Indeed, identifying cooperating mutations that drive malignant transformation is a prerequisite for developing better combinatorial therapies for managing and treating skin cancers. Most skin cancers, including cuSCC [1, 2], have the highest mutation rates among human cancers due to ultraviolet irradiation (UV-IR) induced damage from chronic, intermittent sun exposure. Thus, using human cancer sequencing data alone, with some of the highest mutational burdens of any cancer, poses challenges to identify cooperating, low-penetrant mutations that lead to cancer progression. This presents a need to develop in vivo model systems to help identify and prioritize novel cooperating candidate cancer drivers for keratinocyte transformation and subsequent progression to invasive cuSCC.

Sleeping Beauty (SB) insertional mutagenesis [3] is a powerful tool used to perform genome-wide forward genetic screens in laboratory mice for cancer gene discovery [4–14] in animal models of both hematopoietic and solid tumors [12, 15]. The SB system employs a DNA transposon that is mobilized throughout the genome by a conditional Cre-induced suppressive roles for KMT2C, CREBBP and NCOA2, in the initiation or progression of human cuSCC. Taken together, our in vivo screen demonstrates an extremely heterogeneous genetic landscape of cuSCC initiation and progression, which can be harnessed to better understand skin oncogenic etiology and prioritize therapeutic candidates.
transposase expressed in trans that binds the inverted repeats at the end of the transposon, initiates DNA double-strand breaks and reintegrates the transposon randomly at TA-dinucleotides (reviewed in [5]). The transposon itself contains a minimal internal promoter and splice-donor (SD) and bidirectional splice-acceptor (SA) sequences, such that the SB transposons can either activate proto-oncogenes or inactivate tumor suppressor genes, and identify early cancer progression drivers that cooperate to initiate tumors [11, 14] and potentially drive metastasis [16, 17]. Importantly, SB insertions induce changes in gene expression, thus providing epigenetic information not easily obtained from carcinogenesis mouse models using chemical [18] or chronic UV irradiation [19, 20] or from limited patient samples. We demonstrate that SB mobilization of a low-copy T2/Onc3 transposon allele is sufficient to induce and progress a variety of cancers in vivo. Here, we report our efforts for cancer gene discovery in skin tumors. Using high-throughput sequencing approaches [9, 11] and our SB Driver Analysis [21] statistical framework, we profiled genome-wide SB mutations from cuKA and cuSCC and defined recurrently mutated, statistically significant candidate cancer drivers (CCDs) from bulk cuSCC tumors and from normal keratinocytes and early stage tumors, identifying both known tumor suppressor genes and novel oncogenic drivers. We further prioritized oncogene and tumor suppressor candidates and provide in vitro and in vivo functional evidence for the roles of these genes in the initiation and progression of cuSCC. Taken together, our efforts provide a systematic evolutionary landscape of cuSCC genesis, and highlights key pathways promoting disease that are potential therapeutic targets.

**Results**

**Candidate cancer driver discovery in SB-driven keratinocyte cancer models**

We performed a forward genetic screen using SB insertional mutagenesis to define the genes that cooperate with *Trp53* mutant alleles to drive tumorigenesis. Mice carrying either a conditional *Trp53* null (*Trp53 KO/−*) [22] or recurrent point mutant allele (*Trp53 R172H/−*) [23] and *Actb-Cre* transgene [24] were crossed to Sleeping Beauty mice double homozygous for a low-copy, bi-functional T2/Onc3 transposon (12740) and an inducible SB transposase [7] (S1 Fig). The resulting progeny, referred to as SB|Onc3 mice, exhibited whole-body mutagenesis and succumbed to a wide variety of solid and hematopoietic tumor types with variable penetrance (S1 Table). The presence of a *Trp53* mutant allele accelerated tumor progression and significantly reduced tumor-free survival relative to *Trp53* wild-type littermate controls (*P* < 0.0001, Fig 1A). The SB|Onc3 mice with wildtype *Trp53* exhibited a disease penetrance and latency similar to T2/Onc3 cohorts produced with constitutively active SBase [7]. SB|*Trp53 KO/+* mice were the earliest to reach endpoint, but had the lowest overall tumor burden, presumably because these mice had the highest proportion of hematopoietic disease. SB|*Trp53 KO/+* mice reached endpoint earlier than SB|*Trp53 KO/+* mice and had the widest spectrum of solid tumors. The most prevalent of the more than 20 solid tumor types observed were early-stage cutaneous keratoacanthoma (cuKA) (Fig 1B) and well-differentiated cutaneous squamous cell carcinoma (cuSCC) (Fig 1C and 1F). Importantly, these cuSCC/cuKAs were absent in mice with mutant *Trp53* lacking SB, strongly suggesting SB mutagenesis drives initiation of cellular transformation and subsequent cancer progression of cuSCC in vivo.

**SBCapSeq defines recurrent transposon events in normal, premalignant, and tumor skin cell genomes**

To investigate the genetic events underlying the SB-driven keratinocyte transformation and cuSCC progression in wildtype and *Trp53* mutant mice, we sequenced SB insertions sites from...
60 histologically confirmed cuSCCs, 11 cuKA and 32 normal skin genomes (S2 Table) using an enhanced version of SB Capture Sequencing (SBCapSeq) from our previously published workflow [9] that is optimized for liquid-phase capture of transposons from both normal and tumor specimen genomes. Using the SBCapSeq bioinformatics workflow [9] and gene-guided SB Driver Analysis statistical framework [21], we identified 11,113,694 reads at 59,337 non-redundant transposon insertion sites from cuSCCs which were used to statistically define 1,333 candidate cancer driver genes (S3 Table). Nearly all CCDs were identified across all cohorts, suggesting that their role in driving cuSCC is independent of \( \text{Trp53} \) mutation. Stringent filtering of the SB insertions to include only events with the highest sequencing read depths defines trunk driver alterations predicted to be under positive-selection within the clonally expanding tumors, where they participate in either initiation or early progression of cellular transformation [9, 11]. To identify CCDs that drive keratinocyte transformation, we selected the SB insertion sites with the highest number of sequencing reads to statistically define trunk drivers as previously described [21, 25], reasoning that tumor initiating insertions would show the highest frequency in tumor cells (see Methods). We identified 86 trunk drivers (S4 Table). The top two most significant trunk drivers in cuSCC genomes were the paralogous transcriptional co-factors \( \text{Zmiz1} \) and \( \text{Zmiz2} \), which are collectively and mutually exclusively mutated in 71% of end-stage SB\textit{cuSCC} genomes (S2 Fig). Using the same methodology, we next analyzed cuKA specimens harvested either from mice with cuSCC or from an independent age- and genotype-matched cohort and identified 2,085,280 reads at 6,727 non-redundant transposon insertion sites, defining 192 candidate cancer driver genes (S5 Table) and 10 trunk drivers (S6 Table). Finally, we identified 2,910,945 reads at 47,537 non-redundant...
transposon insertion sites from normal skin specimens with statistically defined enrichment in 547 genes (S7 Table). No trunk drivers were identified. The histologically normal skin genomes from mice with SB mobilization had a mean of 1,320 insertion events (range 1–3,857). Notably, fewer than 1% of the insertions had read depths >300 reads per site (range 1–2,297), which is one order of magnitude less than 23.7% of cuKA genomes or 8.5% of cuSCC genomes with detected insertion events at >300 reads per specimen (cuKA range 1–19,798 and cuSCC range 1–35,809). An oncoprint illustrates the read-depth and insertion patterns that define the trunk drivers in cuSCC and cuKA (Fig 2), comparing the incidence of selected insertions in these same drivers in normal skin. Together, these data strongly suggest that
clonal selection of recurrent, high-read depth SB insertion sites is featured prominently within the cuSCC and cuKA genomes and almost entirely missing within the normal skin genomes.

Further evidence for clonal selection associated with disease progression was provided by technical replication of SBCapSeq libraries, demonstrating excellent biological reproducibility (sequencing of two independent library preparations from single isolation of genomic DNA) of cuSCC genomes (S3A and S3B Fig) and no reproducibility in normal skin genomes (S3C and S3D Fig), illustrating the absence of clonal expansion in the normal samples despite the high insertion burden. Importantly, replicate sequencing of a single library preparation from two normal skin samples demonstrated excellent technical reproducibility (S3E and S3F Fig), further supporting the robustness of our methodology to detect clonal selection. These observations are consistent with the large numbers of mutations observed during chronic UV-IR exposure in human eyelid [26] and SKH hairless mouse [19, 20] skin and, together with our SB data, indicates that keratinocytes within phenotypically normal skin can tolerate high background mutagenic insults that may act as a mutational reservoir prior to clonal selection of cooperating driver mutations that permit frank outgrown of cuSCC in vivo. This is likely the reason that no trunk drivers could be identified from the normal skin specimen genomes. The quantitative, stage-dependent insertion mutation burdens derived from SBCapSeq in cutaneous specimens reveal incredible levels of tumor heterogeneity in SB-driven keratinocyte transformation and progression to cuSCC.

cuSCC demonstrate clonal and sub-clonal intratumor heterogeneity

We identified four skin masses with histologically continuous but distinct regions of cuKA and cuSCC, suggesting for the first time that cuKA may progress directly to invasive cuSCC in vivo. Examples of gross and histological sections are shown in Fig 3A, 3B, 3C, and 3D. We performed SBCapSeq on representative, histologically distinct regions to investigate intratumor heterogeneity and to gain insight into the drivers likely to mediate cuKA to cuSCC progression. Regional sequencing gave an overall lower sequencing depth than our bulk tumor sequencing. Therefore, we adjusted the minimum read-depth threshold to 200 reads for defining clonal insertions. For several loci, we identified insertions at the same nucleotide address across tumor regions from both cuKA and cuSCC in the same mouse, indicating clonal origins of these lesions (S8 Table). Hierarchical two-dimensional clustering of recurrent genic SBCapSeq insertion data from four skin masses containing distinct cuKA and cuSCC regions is shown in Fig 3E. Invariably, clonal insertions into Zmiz1 were identified across each section of all 4 masses at three different TA-dinucleotide addresses, suggesting that Zmiz1 was under positive selection within the nascent transformed cells that gave rise to the distinct cuKA and cuSCC features. We also found evidence for sub-clonal insertion events identified at the same nucleotide address in some but not all regions within the same tumor, demonstrating the evolution of the tumor. Some genes with sub-clonal insertion events with read depths greater than 200 (see Methods) were defined as drivers in the bulk population, including Tcf4 and Pak2, while other genes, such as Egr2 and Rab3gap1, were not (Figs 3F and S4). These data provide evidence for a common keratinocyte progenitor that acquires SB insertions in genes that promote cuKA and are maintained in cuSCC. This analysis also highlights the intra- and inter-tumor heterogeneity of insertions in SB cuSCC tumors and importantly demonstrates that SB insertion profiles can be used to trace the evolution of cuKA to cuSCC using both the insertion nucleotide address and read depth to define clonal relationships.

Comparative oncogenomic meta-analysis

One hundred-seven of the SB candidate Trunk driver genes are direct orthologs of human cancer genes found in the Cancer Gene Census (GRCh38 COSMICv86) [27], a growing catalogue
of mutations causally implicated in cancer, which is an enrichment greater than expected by chance ($\chi^2 = 60.75$ with Yates correction, $P<0.0001$; S1 Text). Another 64 SB candidate driver genes, including ZMIZ1, ZMIZ2, and KMT2C, have human orthologs with recurrent non-silent mutations identified by exome sequencing [1, 2, 28] of 68 cuSCC genomes ($\chi^2 = 23.70$ with Yates correction, $P<0.0001$; S1 Text). To investigate whether the human homologs of the mouse SB events are gained or lost in human tumors, we interrogated the limited data for CNAs in cuSCC tumors published for 40 tumors [29] which identified new and confirmed known recurrent chromosomal gains and losses in cuSCC. We found that KDM4C lies within the region of Chr9p frequently lost, while NOTCH1 (Chr9q), RASA1 (Chr5q) and KMT2C (Chr7q) lie within gained chromosomal regions. KMT2C is a known tumor suppressor in
several solid tumors. Interestingly, copy-number analysis of cuSCC lymph node metastases [30] showed that NOTCH2 (Chr1p12) and CREBBP (Chr16p13) lie within chromosomal regions lost, while the chromosomal region containing TGFBR2 (Chr3q22) can be either lost or gained, and genomic regions containing NOTCH1 (Chr9q), TCF7L2 (Chr10q25) and AKT3 (Chr1q) are gained. 13 genes, including PTEN and TGFBR2, two well-documented tumor suppressor genes, contain mutations in cuSCC lymph node metastases, while PTEN is also lost in cuSCC [29]. It is worth pointing out that missense and nonsense mutations in NOTCH1 are frequent, early events in cuSCC [2], suggesting that amplified genomic regions may contain genes that are deregulated through other mechanisms. Further, NOTCH2 is frequently mutated in cuSCC. From these data, we conclude that, while the human homologs of the SB events can be deleted in cuSCC, it appears that missense and nonsense mutations are much more frequent events. We conclude from these findings that SB identifies the same genes mutated by the diversity of mutational processes operating within end-stage human cuSCC genomes. Our insertion data implies that some of these mutations may be inactivating in human SCCs and warrant further investigation.

To gain insight into the biological functions of the candidate discovery drivers, we used a curated subset of the Enrichr analysis platform [31, 32] gene-set libraries for pathways and gene ontologies (see Methods) to perform biological pathway and process enrichment analysis from the cutaneous specimen cohorts. For this analysis, we first derived an integrated cuSCC driver list from the 60 SB|Onc3 cuSCCs sequenced by SBCapSeq with drivers defined from an additional 23 SB|Onc3 cuSCCs sequenced using Roche 454 and analyzed with our SBDriver Analysis pipeline (S1 Text). We also derived a composite list of cuKA drivers from two different cuKA datasets, one from cuKAs derived from SB|Onc3 mice sequenced by SBCapSeq and the second from cuKAs isolated from a previously published mouse model with K5-Cre; PtenCKO/+ and SB|Onc2 or SB|Onc3 mice [13, 33] sequenced using Roche 454. Using these integrated driver lists, we discovered statistically significant enrichment of recurrent SB candidate drivers in many signaling pathways and biological processes previously associated with human SCC, including EGFR, MAPK, NOTCH, and WNT-TGFβ (S5 Fig and S9 Table). Notably, we also observed that all SB|cuSCC tumor genomes have at least one defined driver involved in chromatin modification, consisting of chromatin and histone modification enzymes, as well as Hippo pathway genes. Further, the SB|cuSCC Trunk drivers encode proteins with significantly more known protein-protein interactions (PPI) than expected by chance ($P = 8.55 \times 10^{-6}$, STRING enrichment analysis; number of nodes: 144, number of edges: 124, expected number of edges: 82) as do the SB|cuKA trunk drivers ($P = 6.93 \times 10^{-5}$, STRING enrichment analysis; number of nodes: 37, number of edges: 14, expected number of edges: 4; S6 Fig). PPIs among drivers identified in SB-driven cuKAs suggests that perturbation of NOTCH signaling may be an initiating or acquired event in cuKA.

Several of the top 30 most frequent drivers in SB cuSCC are key players in multiple pathways, suggesting that SB perturbation of multi-functional genes drives cuSCC. An oncoprint aligning the conserved drivers between SB cuSCC, SB cuKA and mutated human homologs in cuSCC that map to enriched pathways is shown in Fig 4 Comparative genomic integration of the SB defined cuSCC and cuKA drivers with human cuSCC genomes reveals that while differences in the incidence of individual genes across cohorts can vary substantially, the biological pathways are conserved (S5 Fig and S9 Table). These differences may be due in part to the types of alterations produced by SB transposition (gene activation/inactivation) versus UV-IR (silent/non-silent mutations) or, alternatively, may reflect the complexity of the systems of genetic networks [34] that exists between driver genes and keratinocyte transformation phenotypes across species.
Clinical relevance of SB|cuSCC driver orthologs in human SCC

The mammalian paralogous genes Zmiz1 and Zmiz2 are mutually exclusive trunk drivers activated in three quarters of SB-driven cuSCC genomes in an SB-driven cuSCCs, suggesting that Zmiz proteins are proto-oncogenes in cutaneous squamous keratinocyte transformation in vivo. Zmiz1 and Zmiz2 encode transcription factor proteins that share core C-terminal features, including androgen receptor (AR) binding domain, proline-rich binding domain, and MIZ-type zinc finger PIAS (protein inhibitor of activated STAT) domains, that coordinately act to both interact with and regulate the activity of various cancer-associated proteins [35–43]. ZMIZ1 (RAI17/ZIMP10) and ZMIZ2 (ZIMP7) are collectively mutated or amplified in one third of human cuSCC [1, 2] and 5% of hnSCC [44] genomes.

We also sought to investigate the oncogenic potential of ZMIZ1 and ZMIZ2 in SCCs using gene expression analysis from patient samples. Because there are no publicly available expression sets for cuSCC, we utilized human TCGA head and neck SCC (hnSCC) datasets, for which gene expression (RNA-seq) and patient survival outcomes are available, as a proxy for SCCs. We performed a metagene expression analysis of ZMIZ1, identifying a 23-gene signature that links high expression with poor outcomes in human hnSCC patients with a correlation threshold of 0.65 (Cox Proportional Hazards Regression, $P = 0.0195$; Log-rank Test, $P = 0.028$ at 50% quintile; S7 Fig).

From our SB-driven cuSCC model, we observed that all Zmiz-mutated cuSCC genomes contained one or more inactivating SB mutations in chromatin-remodeling genes (Fig 4), suggesting this class of mutated tumor suppressor genes may contribute significantly in driving cuSCC progression. Four of the top-ranked chromatin remodeling trunk driver orthologs were also found to be significantly mutated in >15% of human cuSCC genomes [1, 2].
Importantly, as observed in mouse tumors (Fig 2), mutations in the orthologs KMT2C (MLL3), ARID1B (BAF250B), NCOA2 (KAT13C), and CREBBP (CBP/KAT3A) were also found in the same cuSCC genomes harboring ZMIZ mutations (Fig 4), suggesting they might cooperate during human keratinocyte transformation or tumor progression. Previously, KMT2C (MLL3) mutations were found to be associated with poor outcomes in aggressive cuSCC [1], suggesting mutations in this gene may be a biomarker for advanced stage disease.

**Gene expression alterations associated with Zmiz1ΔN185 in cuSCC genomes**

SB activating trunk insertions in Zmiz1 or Zmiz2 are unique to SB-induced cuSCC tumors and are predicted to activate gene expression by driving expression of specific N-terminal truncated forms of Zmiz1ΔN185 [7] and Zmiz2ΔN184, a closely related paralog (S2A and S2B Fig). The likelihood of achieving unique, sense-strand oriented SB insertion events into two related genes at the same protein coding region on two different chromosomes in dozens of tumors from independent mice by chance is incredibly small ($P < 0.0001$), suggesting both trunk driver ZMIZ oncoproteins contribute directly to keratinocyte initiation and tumor progression *in vivo*. The ZMIZ1ΔN185 and ZMIZ2ΔN184 onco-proteins share 62% sequence identity and retain both the zinc finger PIAS, proline-rich transactivation and AR binding domains. Importantly, both oncoproteins lack the N-terminus in the full-length protein, which is thought to inhibit the intrinsic transcriptional activity of a C-terminal proline-rich transactivation domain [42].

To analyze the expression of SB insertion-driven transcripts for Zmiz1, Zmiz2 and other drivers (SBfusion reads [9]) in cuSCC genomes identified by high-insertion read depth from our genomic sequencing results, we sequenced ribo-depleted, whole-transcriptome RNA-seq (wtRNA-seq) libraries created from 7 cuSCC tumors with activating SB insertions into either Zmiz1 (n = 6) or Zmiz2 (n = 1) (S10 Table). The wtRNA-seq reads contained spliced chimeric transposon splice-donor or splice-acceptor sequences fused to adjacent known exons in the mouse genome. In total, we identified 2,906 SBfusion wtRNA-seq reads from 889 unique RefSeq genes (S11 Table), including 154 progression and 27 trunk cuSCC drivers, respectively (S12 and S13 Tables), which is substantially more than expected by chance (Chi-square test with Yates correction, $\chi^2 = 159$, $P < 0.0001$). Thirty-three genes with SBfusion events were supported by ≥5 SBfusion reads, including our top trunk drivers Zmiz1 and Zmiz2 and several keratins (Fig 5A). We observed a positive correlation between clonal activating insertions with high read-depth SB insertions (Fig 5B) and high expression of SBfusion reads in Zmiz1 (Fig 5C). Importantly, in each tumor, SBfusion reads containing either SB-Zmiz1 or SB-Zmiz2 spliced exons were represented in the top 2% of SBfusion transcripts and were supported by multiple independent SBfusion reads (S14 and S15 Tables). Further, Zmiz1 or Zmiz2 gene expression was significantly increased within each of the 7 cuSCC genomes (≥100 and up to 300 times normal expression levels) (Figs 5D and S9A, S9B, S9C, S9D, S9E, and S9F). This strongly suggests that the SB insertion events driving either Zmiz1 or Zmiz2 have been clonally selected to increase oncogenic expression of the N-terminal truncated forms during keratinocyte transformation and cuSCC progression *in vivo*. We interrogated the expression of SBfusion reads from 8 additional genes in this tumor subset with high read depth insertion sites (Fig 5C). Epha1 and Wnt7b had activating SB insertions with high SBCapSeq read depths and wtRNA-seq SBfusion reads resulting in significantly higher gene expression (≥100 times) compared to basal expression levels observed in tumors without insertions or SBfusion reads (Figs 5D, S8E, and S8F). Since SB insertions in Epha1 and Wnt7b were not recurrent at the population level, these genes were not identified as statistically significant drivers in cuSCC. However, the integrated genomic and transcriptomic analysis strongly suggests that each of
Novel genetic drivers of skin SCC identified by transposon mutagenesis
Fig 5. Clonally selected SB insertion events affect trunk driver gene expression in SB-induced cuSCC genomes. Seven cuSCC genomes with mutually exclusive high-read depth Zmiz1 or Zmiz2 SB insertions were selected for RNA-seq analysis of SB-fusion transcripts and global gene expression levels. (A) Genes with the most abundant SBfusions reads detected by wtRNA-seq analysis of the 7 cuSCC tumors. (B) Predicted gene expression effects of SB insertions in 10 drivers with high read-depth SB insertions. (C) High positive correlation between SBCapSeq read depth and Zmiz1 expression by wtRNA-seq analysis. (D) Multifold induction of gene expression in cuSCC masses with (+) high read depth activating SB insertions among 4 candidate oncogenic drivers compared with normal gene expression levels in cuSCC tumors without (−) SB insertions. Individual gene plots may be found in (S14 and S15 Tables). (E) Reduced gene expression in cuSCC masses with (+) high read depth inactivating SB insertion events among 6 candidate tumor suppressor drivers compared with normal gene expression levels in cuSCC tumors without (−) SB insertions. Individual gene plots may be found in (S14 and S15 Tables).

https://doi.org/10.1371/journal.pgen.1009094.g005

these genes may have a ‘private’ role in driving tumor progression in the genomes in which they occur with elevated expression levels. Both genes encode proteins that are linked to cancer hallmarks—Epha1 is an ephrin receptor subfamily of the protein-tyrosine kinase family [45] and Wnt7b is a WNT family member [46]. For the 6 predicted inactivated genes, Nf1, Adam10, Oxsr1, Map2k4, Cul5, and Cep350 we observed reduced gene expression that correlated with high SBCapSeq read depth of inactivating SB insertions and truncated SBfusions reads (Figs 5E, and S9A, S9B, S9C, S9D, S9E, and S9F). Adam10, Cul5, and Nf1 are Trunk Drivers and Map2k4 is a Progression Driver in cuSCC cohorts. Oxsr1 and Cep350 were private selected events in individual tumors that may operate in cuSCC progression within the tumors that harbor clonal insertion events. Taken together, these data demonstrate that integrated genomic and transcriptome analyses may provide a means to discover significant cooperating drivers in cuSCC genomes from bulk specimens.

Next, to investigate the transcriptional signatures associated with SB-driven Zmiz1\(^{AN185}\) in end-stage cuSCC, we selected 13 cuSCCs with activating Zmiz1 insertions and 9 end-stage cuSCCs lacking insertions in Zmiz1 to perform microarray analysis. We confirmed that SB insertions drive increased expression of truncated Zmiz1 by RT-PCR and by normalized microarray data analysis, where cuSCC tumors with Zmiz1 insertions were found to have significantly higher mRNA abundance than tumors without Zmiz1 insertions (Fig 6A and S16 and S17 Tables). We also observed a positive correlation between high-read-depth SB insertions among the 13 tumors harboring trunk Zmiz1 insertions and increased normalized Zmiz1 gene expression (Fig 6B). Differential expression analysis of the microarray data using limma from tumors with and without SB insertions in Zmiz1 identified 432 probe sets from 355 genes significantly differentially expressed genes with FDR adjusted \(P<0.2\) (S18 Table). A heat map of normalized expression plots for the top 25 most differentially expressed genes is shown in Fig 6C. Differential expression plots for the top 8 most differentially expressed genes (\(P_{\text{FDR}}<0.05\)) is shown in Fig 6D. Mtdh1 encodes metadherin, also known as AEG-1, a downstream target of Ras and c-Myc, and MTHD1 is upregulated in many cancers [47]. Set encodes the SET nuclear proto-oncogene, which interacts with several histone modifying proteins, including CREBBP, a driver identified in our SB cuSCC tumors and KMT2A, a family member of KMT2C, also identified as a driver in our screen. Taken together, our microarray gene expression analysis demonstrates that the ZMIZ1\(^{AN185}\) isoform may be a critical mediator of gene expression changes essential for early keratinocyte transformation and sustained during cuSCC progression to end-stage tumors.

ZMIZ1\(^{AN185}\) expression transforms human cutaneous keratinocytes in vitro

The selection of SB insertions promoting up-regulation of a truncated Zmiz1 transcript suggests that ZMIZ1\(^{AN185}\) may be an early transforming event during cuSCC development. To test this hypothesis, we first generated stable overexpression of either ZMIZ1 or the truncated ZMIZ1\(^{AN185}\) in the spontaneously immortalized human cutaneous keratinocyte HaCaT cell line
Interestingly, we noted that ZMIZ1ΔN185 was more highly expressed at the protein level compared to full-length ZMIZ1, despite similar transduction efficiencies (Fig 7A). This observation is in line with published data by Rogers et al. 2013, where they expressed ZMIZ1ΔN185 with greater stability than full-length ZMIZ1 in various cancer cell lines. Next, we assessed anchorage-independent growth as an in vitro surrogate for the earliest stages of tumor development in HaCaT cells stably transduced with empty vector, full-length ZMIZ1 or ZMIZ1ΔN185. Soft agar assays showed that ZMIZ1ΔN185 expression resulted in significantly increased colony formation compared to empty vector or full-length ZMIZ1 (P = 0.0004, one-factor ANOVA, Figs 7B and 7C). These data suggest that ZMIZ1ΔN185 has a distinct function from full-length ZMIZ1 and that it promotes transformation in an oncogenic manner. Next, we investigated whether ZMIZ1 functions as an oncosgene in established cuSCCs. We stably knocked-down total ZMIZ1 and showed a significant decrease in cell proliferation with ZMIZ1 depletion compared to a controlled
scrambled shRNA at 96 hours for cuSCC cell lines A431 (Fig 7D and 7E, 2-factor ANOVA overall \( P = 0.0045 \); FDR-corrected \( q \)-values for multiple testing \( q < 0.0001 \)) and COLO16 (Fig 7F and 7G, 2-factor ANOVA overall \( P = 0.0005 \); FDR-corrected \( q \)-values for multiple testing \( q < 0.0001 \)). We showed similar results when we knocked down the ZMIZ1 paralog ZMIZ2 in each cell line [49]. Taken together, these data suggest that ZMIZ1 \( \Delta N185 \) is an oncogene that drives transformation of human cutaneous keratinocytes \textit{in vitro}, and that ZMIZ1 functions as an oncogene in established cuSCCs. Further experiments are needed to address the direct role of ZMIZ1 \( \Delta N185 \) in promoting cuSCC initiation.

**Knockdown of chromatin modifiers drive human cutaneous keratinocyte transformation**

Chromatin remodeling is one of the most frequently inactivated biological processes in SCC, and has been demonstrated to be involved in either tumor initiation or progression in multiple
cancer types [50–55]. In our SB cuSCC screen, chromatin remodeling was the top biological process that was significantly enriched (S5 Fig and S9 Table). However, it remains unclear how expression alteration of chromatin remodeling genes function in the cuSCC disease trajectory. We prioritized three tumor suppressor genes with high concordance between SB-cuSCC and human cuSCC involved in histone modification (KMT2C and CREBBP) or transcriptional regulation (NCOA2) to study their roles in keratinocyte transformation using immortalized and minimally transformed HaCaT keratinocyte cell line [48]. First, we confirmed expression of each gene within HaCaT cells [49]. Next, we used shRNAs to assess whether loss of these tumor suppressors can promote cellular transformation in vitro as a surrogate for tumor initiation. Forced depletion of CREBBP significantly increased cellular proliferation relative to control (Fig 8A, 1-factor ANOVA, P < 0.05) and a similar trend was observed with KMT2C knockdown. No significant difference in proliferation was observed between NCOA2 knockdown and control (Fig 8A). To more stringently investigate transformation, we assessed anchorage-independent growth, a hallmark of cellular transformation, using a soft agar colony formation assay (Fig 8B). Knockdown of CREBBP (Fig 8C) or KMT2C (Fig 8D) expression resulted in larger and significantly more colonies relative to control while no significant difference was observed between NCOA2 knockdown and control (Fig 8E). Taken together, these data suggest that knockdown of CREBBP or KMT2C can transform the immortalized HaCaT cell system in vitro.

Knockdown of chromatin modifiers promote cuSCC tumor growth in vivo

Next, we investigated whether down-regulation of CREBBP, KMT2C or NCOA2 could promote tumor progression in cuSCCs. First, we used shRNAs to stably knockdown these genes in established cuSCC cell lines A431, COLO16 and cuSCC13. We showed down regulation of KMT2C and NCOA2 for A431 (Fig 9A), COLO16 (Fig 9B) and cuSCC13 (Fig 9C) by qPCR relative to cells transduced with a non-targeting control shRNA. Stable knockdown of CREBBP was achieved in A431 and COLO16 cells (S10A and S10B Fig). Next, we performed proliferation assays and demonstrated a significant increase in A431 cell proliferation with down regulation of KMT2C and NCOA2 at 96 hours compared to control (Fig 9D, 2-factor ANOVA overall P = 0.0005; FDR-corrected q-values for multiple testing q < 0.0001 for KMT2C; q = 0.0019 for NCOA2). Knockdown of NCOA2 in COLO16 cells trended towards increased proliferation (Fig 9E) but it was not significant. However, proliferation was significantly increased with KMT2C knockdown at 96 hours compared to control (Fig 9F, FDR-corrected q-values for multiple testing q = 0.01). No change in proliferation was observed in A431 or COLO16 cells with CREBBP knockdown (S10C and S10D Fig). This is in contrast to the increased colony formation with CREBBP depletion in HaCaT cells. Finally, proliferation in SCC13 cells was significantly increased with KMT2C or NCOA2 depletion compared to control (Fig 9F, 2-factor ANOVA overall P = 0.008; FDR-corrected q-values for multiple testing q = 0.06 for KMT2C; q < 0.0001 for NCOA2). We then determined whether CREBBP, KMT2C and NCOA2 function as cuSCC TSGs in vivo and performed xenograft assays by subcutaneous injection of cells for each of the three cuSCC cell lines into NSG immunodeficient mice. CREBBP depletion in A431 cells did not result in a change in tumor volume (S10E Fig) and this gene was not assayed in vivo using the other two cuSCC cell lines. All three cuSCC cell lines demonstrated enhanced in vivo tumor growth with KMT2C or NCOA2 depletion compared to cells with a non-targeting control shRNA (Fig 9G and 9H and 9I, 1-factor ANOVA, q < 0.05). Pictures of cuSCCs harvested at necropsy at defined time points of 25 days for A431 (S11A Fig), COLO16 at 18 days (S11B Fig) and SCC13 at 49 days (S11C Fig) show the individual tumor volumes for the control and targeted knockdown cohorts. Together, these data
provide functional evidence that KMT2C and NCOA2, but not CREBBP, are human TSGs that promote in vivo cuSCC progression.

**Discussion**

In this study, we sought to define the cooperating genetic events required for keratinocyte transformation and progression to frank cuSCC using the SB transposon mutagenesis system...
in vivo. Using an Actb-Cre transgene to activate systemic expression of the transposon, we observed tumor lesions in multiple organs. The kinetics of tumorigenesis were further accelerated in mice harboring either null (Trp53KO/+ [22]) or recurrent point mutant (Trp53R172H/+ [23]) alleles. This observation is consistent with previous studies showing that missense mutations in Trp53 promote malignant transformation by abrogating both the tumor suppressive and promoting the oncogenic (gain-of-function and/or dominant negative) functions of Trp53 in vivo [23, 56–59]. We and others have also documented that in wild type and heterozygous sensitizing mutant mice that present with solid tumors, a much greater proportion of inactivating tumor suppressor SB insertion patterns predominate [21, 25].

Despite the high curative rate for cuSCC, there is a significant population of patients who develop advanced disease for which there are no defined therapeutic targets. Due to the high UV-induced mutation burden, combined with the scarcity of clinical samples, it is challenging...
to define the recurrent and druggable targets within these tumors. To address this need, we isolated early and advanced lesions from a SB mouse model of cuSCC to define recurrent drivers and potential new avenues for therapeutic intervention. From our in-depth sequencing analysis of SB insertions in advanced cuSCC genomes, we identified clonal and sub clonal drivers involved in the progression and maintenance of cuSCC. 64 and 17 drivers have human orthologs with non-silent mutations or copy number loss in human cuSCC genomes, respectively. Intriguingly, the majority of drivers in SB-driven cuSCC lesions were predicted tumor suppressor genes, suggesting that cumulative loss of these genes is a central feature of cuSCC. We showed that a few key TSGs, including *KMT2C*, are differentially expressed in cuSCC with prognostic implications for patients with hnSCC.

We subsequently focused on understanding the genetic events required for keratinocyte transformation. Our SBCapSeq and SB Driver analysis identified 349 CCDs, regardless of *Trp53* status, of which 107 have been implicated in human cancers. Intriguingly, we observed that the SB-driven cuSCC lesions were driven almost exclusively by SB inactivating insertions in tumor suppressor genes, suggesting that cumulative loss of these TSG drivers is a central feature of cuSCC. Indeed, recent analysis from our lab, using a complimentary promoterless SB transposon forward genetic screen, revealed that tumor suppressor drivers alone can lead to keratinocyte initiation and cuSCC progression in vivo [49]. We defined two mutually-exclusive paralogous oncogenic drivers, Zmiz1 and Zmiz2 among the most recurrent drivers and our report is the first to identify Zmiz2 as a significant candidate cancer gene in any SB study and the first to discover mutual exclusivity among activated Zmiz1, Zmiz2, and Mamld1 in early keratinocyte transformation and cuSCC progression. Insertions into Zmiz1 [60–62] and Mamld1 [63] have been previously observed in skin tumors induced by transposon insertional mutagenesis. The possible redundancy of ZMIZ oncoproteins in the keratinocyte transformation program, by mutual exclusivity of trunk insertions within cuSCC genomes, strongly suggests that the Zmiz1\(^{\Delta N185}\) and Zmiz2\(^{\Delta N184}\) oncoproteins may have common neomorphic properties that contribute to the hallmarks of skin cancer. The work of Rogers et al. [62] and more recently Mathios et al. [60] on the epigenetic regulation of ZMIZ1 expression in human cancer, together with our finding that ZMIZ1 and ZMIZ2 are mutated with at least one non-silent alteration in at least one third of published human cuSCC genomes [1, 2] suggest that these genes may be an important initiating trunk mutation in human keratinocyte initiation and cuSCC progression. We confirmed this hypothesis, showing transformation of the immortalized keratinocyte HaCaT cells stably transduced with the N-terminally truncated ZMIZ1\(^{\Delta N185}\) isoform and decreased cellular proliferation in established cuSCC lines with ZMIZ1 or ZMIZ2 depletion.

Intriguingly, all SB cuSCC tumors with Zmiz1/2 insertions had inactivating insertions in at least one gene involved in chromatin remodeling, which suggests that changes in epigenetic regulation are necessary to progress a benign lesion to frank cuSCC. Our functional data suggests that *CREBBP* is involved in the earlier stages of keratinocyte transformation, while decreased *NCOA2* expression can drive disease progression. COMPASS complex members *Kmt2c* and *Kdm6a* and the switch/sucrose non-fermenting (SWI/SNF) genes *Arid1b*, *Arid1a* and *Pbrm1* were recurrently altered in our SB-cuSCC tumors. Although the COMPASS and SWI/SNF pathway members have been reported to be mutated in various cancers [64–67], whether alterations in one or cooperation of both pathways is required for the etiology of the disease remains unclear. We found that knockdown of *KMT2C*, a key molecule in the COMPASS complex, in human cuSCC cell lines accelerated *in vitro* proliferation and *in vivo* xenograft growth, supporting a role for this gene as a tumor suppressor in advanced disease. Of clinical significance, several recent studies have highlighted potential dependencies amenable to pharmacologic inhibition in cancer cells with aberrations in COMPASS members [67].
Our experimental approach for driving high mutational burdens by SB mutagenesis recapitulates the sporadic, stepwise evolutionary selection of cooperating drivers in non-melanoma skin cancers in humans exposed to solar UV. Despite these divergent mutagens, we observed remarkable overlap of driver genes, pathways, and networks between the mouse and human cutaneous SCCs, suggesting a deep mutual biology between these shared drivers and cutaneous oncogenesis. We hope that by defining the cooperative oncogenic and tumor suppressor networks that operate during keratinocyte transformation and subsequent cuSCC progression, the results of our screen will provide a foundation for exploring new therapeutic strategies.

**Methods**

**Ethics statement**

Mice were bred and maintained in accordance with approved procedures granted by the respective Institutional Animal Care and Use Committees (IACUC) at the National Cancer Institute Frederick National Lab, A*STAR Biological Resource Centre, Houston Methodist Research Institute, and Moffitt Cancer Center.

**Mice used for SB screens**

The following alleles were used to construct the SB-driven mouse model of multiple solid tumor histologies: Actb-Cre (FVB/N-Tg(ACTB-cr e)2Mrt/J) [24]; Trp53\textsuperscript{flox/+} (FVB.129P2-Trp53\textsuperscript{tm1Brn}/Nci) [22]; Trp53\textsuperscript{LSL-R172H}/+ (129S4-Trp53\textsuperscript{m2Tyj}/Nci) [23]; T2/Onc2(TG.12740)(TgTn(sb-T2/Onc3)12740Njen) [7]); and Rosa26-LSL SBase or SBase\textsuperscript{LSL}; (Gt(ROSA)26Sor\textsuperscript{tm1}(sb1)Njen) [68]). The resulting cohorts of mice were on mixed genetic backgrounds consisting of C57BL/6J, 129, C3H and FVB. Genotyping by PCR assays with primers specific to the alleles was performed. No sample size estimate was used to determine the number of mice for aging. Animals were co-housed for the duration the experiment, except in rare cases where single mice were separated based on vet recommendation due to fighting or near tumor burden endpoint. Gross necropsies were performed and all masses were documented and prepared for subsequent analysis. See S1 Fig for details of mouse crosses. Both sexes were used for experiments. No randomization was performed; mice were assigned to groups based on genotype. No blinding was performed.

**Histological analysis**

Histological analysis of spleen was performed on 5-μm sections of formalin-fixed, paraffin-embedded (FFPE) specimens stained with hematoxylin and eosin. As expected, robust nuclear staining of SB transposase (SBase) was confirmed by immunohistochemistry on FFPE tissues after antigen retrieval (pH 9) and endogenous peroxidase inhibition followed by overnight incubation with mouse antibody to SBase (anti-SBase; R&D Systems; pH 9; 1:200 dilution). After incubation with primary antibody, chromogen detection (with HRP polymer, anti-rabbit or anti-mouse, with Envision System from Dako) and hematoxylin counterstaining were performed per manufacturer’s instructions. Genomic DNA (gDNA) was isolated from flash frozen necropsy specimens using Qiagen Gentra Puregene DNA isolation kit protocol for tissue.

**Mapping transposon insertion sites using the splink\_454 method**

SB insertion reads were generated by 454 GS Titanium sequencing (Roche) of pooled splinkerette PCR reactions with nested, barcoded primers was performed [10, 11, 69]. Pre- and post-processing of 454 reads to assign sample DNA barcodes, filter out local hopping events from donor chromosomes, and map and orient the SB insertion sites across the entire nuclear
genome of the mouse was performed. All SB insertions from donor chromosomes were filtered out prior to identification of common insertion sites using the Gaussian kernel convolution (GKC) [69, 70] and SB driver analysis [21] methods.

**Mapping transposon insertion sites using the SBCapSeq method**

Full details for the SBCapSeq protocol (9) optimized for sequencing from solid tumors, will be published elsewhere (Mann et al., in review; for general protocol and concept, see [9, 71, 72]). Briefly, for selective SB insertion site sequencing by liquid hybridization capture, gDNA (0.5 μg per sample) of either bulk tumor specimens or single cell WGA genomes was used for library construction using the AB Library Builder System, including random fragmentation and ligation of barcoded Ion Xpress sequencing adapters. Adapter-ligated templates were purified by Agencourt AMPure beads and fragments with insert size of 200±30bp were excised, purified by Agencourt AMPure beads, amplified by 8 cycles of adapter-ligation-mediated polymerase chain reaction (aLM-PCR), and purified by Agencourt AMPure beads with elution in 50 μl of TE (1X Tris-EDTA [Ethylenediaminetetraacetic acid], pH8). Capture hybridization of single or multiplexed up to 12 barcoded libraries (60 ng per sample) was performed using custom xGen Lockdown Probes (IDT; full details available at https://doi.org/10.35092/yhjc.11441001.v1 [71]). All 120-mer capture and blocking oligonucleotide probes were purchased from IDT as Ultramer DNA Oligos with Standard Desalting. Bar-coded single and multiplex captured library fragments were further amplified by 12 cycles of LM-PCR and run using an Agilent 2100 Bioanalyzer or TapeStation to estimate enrichment. Bar-coded single and multiplex captured libraries were quantified by Qubit Fluorometer and quantitative Real Time-PCR (qRT-PCR) were used to dilute libraries for template preparation and Ion Sphere Particle (ISP) loading using Ion Chef System and sequencing on the Ion Proton platform with PIv3 semiconductor wafer chips per manufacturers recommended instructions. High-throughput sequencing of up to 39 multiplex captured libraries was carried out per PIv3 chip to achieve at least 1.5 million reads per barcode. Reads containing the transposon IRDR element were processed using the SBCapSeq bioinformatic workflow as described [9].

**SB driver analysis**

BED formatted files containing SB insertions from each of the histologically verified SB-cuSCC, SB-cuKA, and SB-cuSK cohort specimens (S1 and S2 and S3 Data) were used to perform SB Driver Analysis to identify statistically significant discovery, progression, and trunk driver genes that contain more SB insertions than expected by chance and were recurrently altered in three or more tumors [21]. Discovery and progression SB Driver Analysis considered all SB insertion events; trunk SB Driver Analysis considered only insertions represented by 5 or more reads. Statistically significant discovery drivers were defined as adjusted p-values using false discovery rate (FDR) multiple testing correction (q-value<0.5). Statistically significant progression and trunk drivers were defined as adjusted p-values using family-wise error rate (FWER) multiple testing correction (FWER adjusted-P<0.05). Statistically significant drivers on donor and non-donor chromosomes analyses were performed separately and combined into a single driver lists (S1 and S2 and S3 Data). Due to a local-hopping phenomenon known to occur with SB, where transposition events are biased to occur in cis along the donor chromosome more frequently than in trans to non-donor chromosomes in the genome, insertions from the donor chromosome are typically filtered away computationally before candidate cancer genes are identified. However, doing so in cohorts derived from mice with only a single SB donor allele (like our cuSCC cohorts) means that genome-wide driver analysis is not possible, because all genes on the donor chromosome are censored and are not reported even
though they may contribute to the overall tumor burden. Since we obtained quantitative SBCapSeq datasets in this study, we evaluated whether all genes on the donor chromosome should be censored or if perhaps, as observed in PiggyBac transposon genomes, only a relatively small portion of the donor chromosome locus, occurring in close proximity to the donor concatemer insertion site, demonstrate significant bias compared with the frequency of SB insertions into non-donor chromosomes. In the cuSCC cohorts sequenced with SBCapSeq, we found that donor chromosome SB insertion site frequencies are significantly higher between chr9:79,000,000 and chr9:95,000,000 and drop to the genome wide average for non-donor chromosomes over the rest of chromosome 9 (S12 Fig), suggesting that filtering all SB donor chromosome insertion events may not be warranted. To confirm and extend this observation, we used all tumors within the SBCDDB with chromosome 9 donor alleles and identified that the same region between chr9:79,000,000 and chr9:95,000,000 had a higher frequency of SB insertions than the rest of chromosome 9 loci, which again matched non-donor chromosome insertion frequencies (S12 Fig). Thus, we ran SB Driver Analysis on tumors derived from mouse cohorts harboring an SB T2/Onc3 TG.12740 donor allele chromosome 9 (cuSCC83_454, cuSCC60_SBC, and cuKA11_SBC), by excluding the 81 genes that map between *Filip1* at chr9:79663368–79825689 and *Tfdp2* at 96096693–96224065 from the driver gene output files. To define trunk drivers, we used a normalized number as a percent of the total SB insertions, 300 reads is the cut off for the top 5% of reads, based on the assumption that the distribution of SB insertions from normal skin represented a normal distribution of background events not exceeding 5% of reads. We selected a driver cut-off as a whole number that was within the 5% of the highest insertion sites from the normal skin samples recognizing that a random number of insertions may have higher read depths from technical (e.g., amplicon amplification bias or known limitations of the Ion Torrent sequencing and alignment platform of homopolymer regions) and/or biological (e.g., clonal keratinocyte distributions within the sampled histologically normal skin or insertion events linked to loci overrepresented within the sampled histologically normal skin genomic DNA) sources of variation that may permit insertion overabundance compared with the majority (~95%) of reads within the range of sequencing depths represented within the specimen pools. In sequencing the histologically distinct regions from cuSCC masses, we used a normalized number as a percent of the total SB insertions, 200 reads representing read depths in the top 30%, so as to provide more power to detect recurrent gene insertions, given that our sequencing depth was lower for this experiment. Despite lowering this threshold, we note that *Zmiz1* was the only gene that consistently had activating transposon insertions in all four independent lesions sequenced, suggesting that *Zmiz1* proto-oncogenic activation is selected early and maintained throughout clonal evolution of these histologically heterogeneous masses.

**Microarray gene expression analysis**

Gene expression profiling of histologically confirmed SB-cuSCC masses selected by presence or absence of *Zmiz1* insertion by 454Splink sequencing was performed using Affymetrix microarrays, as described previously [9]. Briefly, 100ng of total RNA for each sample was extracted using a Norgen Biotek Animal Tissue RNA Purification kit (Cat #25700) followed by labeled with an Affymetrix 3’ IVT Express kit (Cat # 901229) using the manufacturer’s instructions. Labeled samples were hybridized to Affymetrix GeneChip Mouse Genome 430 2.0 Arrays, and scanned at the University of Otago Genomics & Bioinformatics Facility. Raw data processing used R (version 3.6.1) [73] with the “rma” function of the “affy” package [74], including quantile normalization but no background correction. Quality assessment of the microarray data was performed in R using the “affyQCReport” package [75]. Differential
expression was performed using limma [76], and gene set analysis of Reactome pathways was performed using ReactomePA [77]. All additional data analysis and visualization were performed using Python and R.

**Transcriptome sequencing**

Total RNA was isolated from flash frozen necropsy specimens using mirVana miRNA Isolation Kit (Ambion by Life Technologies, AM1560) from a shaved portion of each pathology verified cuSCC mass from 7 tumors for which sufficient tissues were available and the presence of a high read depth Zmiz1 or Zmiz2 trunk driver insertion was identified by SBCapSeq. Whole transcriptome RNA-Seq (wtRNA-Seq) libraries were prepared from total RNA (5 μg per sample) plus ERCC RNA Spike-In (Ambion by Life Technologies, 4456739) followed by selective ribosomal RNA (rRNA) depletion using the RiboMinus Eukaryote System v2 (Ambion by Life Technologies, A15026). rRNA-depleted total RNA (500 ng per sample) was used for whole transcriptome library construction according to the Ion Total RNA-Seq Kit for the AB Library Builder System (Life Technologies, 4482416) protocol for barcoded libraries, Ion Sphere Particle (ISP) loading using Ion Chef System and sequencing on the Ion Proton platform with PI v3 semiconductor wafer chips per manufacturers recommended instructions. Up to 4 RNA-Seq libraries were multiplex sequenced on a PI v3 chip twice to achieve >3 GB (~40 million reads) per specimen. The Life Technologies Torrent Suite software was used to perform checking of the raw sequence data before the generation of sequencing read files in FASTQ format and minimally processed mRNA-Seq and wtRNA-Seq reads were processed using the Bowtie2 [78] and Tophat [79] algorithms to align reads to a custom version of the mouse mm9+pT2/Onc genome, as described [9]. The detection of SB fusion events, defining novel SB fusion transcripts, and measuring transcript abundance in SB cuSCC tumor specimens from RNA-Seq data were performed as described [9].

**Biological pathway and process enrichment analysis**

Enrichr [31, 32], an online analysis tool for human and mouse gene-set enrichment, was used to identify specific signaling pathways and processes enrichment using the various cohort candidate cancer driver genes and/or their human orthologs (S9 Table and S1 Text). The online STRING network enrichment analysis tool [80] was also used to determine functional connections between trunk drivers from the cuSCC and cuKA cohorts.

**Human cancer cell lines**

The cutaneous squamous cell carcinoma A431(CRL–1555) and grade IV lung squamous cell carcinoma SW900 (HTB–59) cell lines, were purchased from American Type Culture Collection (ATCC) and grown according to the manufacturer’s suggested conditions in complete medium (1× DMEM (ATCC, 30–2002) for A431 cells or 1× ATCC-formulated Leibovitz’s L–15 Medium, Catalog No. 30–2008 medium (ATCC, 30–2001) for SW900 cells) supplemented with 10% FBS and 1× penicillin-streptomycin grown at 37˚C in 5% CO2. The cutaneous squamous cell carcinoma SCC13 cell line [81, 82], were obtained from Harvard Skin Disease Research Center (HSDRC) and grown using the human keratinocyte culture methods provided by the HSDRC. The cutaneous squamous cell carcinoma COLO16 cell line [83] were grown in DMEM/Ham’s F12 50/50 media supplemented with cholera toxin (Sigma, C0852), insulin (Sigma, I5500-50MG), epidermal growth factor (Serotec/Bio-Rad, EGF-1), hydrocortisone (Sigma, H4881-1G), liothyronine (Sigma, T6397-100MG) and apo-transferrin (Sigma, T2252-100MG). All cell lines were verified human origin and free from pathogens and
mycoplasma. Mycoplasma infection monitoring was performed using MycoAlert Detection Kit (Lonza) and only mycoplasma-free cultures were used.

**Generation of stable shRNA-expressing cell lines**

High titer lentiviral particles for pGIPZ-shRNA constructs targeting each of the 12 cutaneous candidate driver genes, and one non-targeting control, were purchased from Thermo Scientific Open Biosystems (S19 Table). Cells were plated at a density of $5 \times 10^4$ cells per well in a 24-well plate in complete media 24 hours prior to infection. The following day, cells were plated with serum-free culture medium containing 8 µg/mL polybrene (Millipore), and transduced as pools of three independent shRNAs to different exons of each target gene, or individually for the control, at multiplicity of infection of 6. Puromycin selection was added the following day at concentrations of 1.0 to 3.0 µg/mL puromycin (Thermo Fisher Scientific) in complete media, and replaced every three until stable lines were achieved.

**Proliferation assay**

To assess cuSCC progression, cuSCC cell lines were seeded in 6 well plates at a density of $5 \times 10^4$ cells per well in complete media. Cell numbers for each condition were counted after 48- and 96-hour post-seeding. To assess transformation by proliferation, HaCaT cells were seeded at a density of $5 \times 10^5$ cells per well and cell counts performed after seven days.

**Soft agar assay**

To assess anchorage-independent growth, HaCaT cells were seeded at a density of $1 \times 10^4$ cells per well in a six well plate in 0.3% agar, over a layer 0.6% agar. Complete media was added to each well the following day, and replaced every two days for four weeks. At end point, cells were fixed in 20% methanol and 0.0025% crystal violet. Washes were done with milliQ water until low background was achieved, and plates were scanned, and colonies counted for each condition. Zmiz1 full length, Zmiz1\textsubscript{FL} (pLV[Exp]-EF1A\textgreater{mZmiz1[NM_183208.4](ns})\textbackslash{Myc/FLAG}: P2A:TurboRFP(ns):P2A:Puro), N-terminally truncated, Zmiz1\textsubscript{ΔN185} (pLV[Exp]-EF1A\textgreater{mZmiz1[deltaN185 -NM_183208.4](ns})\textbackslash{Myc/FLAG}: P2A:TurboRFP(ns):P2A:Puro), or empty vector control, EV (pLV[Exp]-EF1A\textgreater{TurboRFP(ns):P2A:Puro}), expression vectors were custom synthesized using the VectorBuilder platform (Cyagen). Cells were transfected with the expression plasmids and selected with 800µg/mL G418 to obtain stable cell lines.

**qRT-PCR**

Total RNA was purified and DNase treated using the RNeasy Mini Kit (Qiagen). Synthesis of cDNA was performed using SuperScript VILO Master Mix (Life Technologies). Quantitative PCR analysis was performed on the QuantStudio 12K Flex System (Life Technologies) or 7900HT Sequence Detection System (Applied Biosystem). All signals were normalized to the levels of GAPDH TaqMan probes. TaqMan probes were obtained from Life Technologies (S20 Table).

**Xenografts**

One million cells were prepared for injection into the left flank of randomized selected male and female immunodeficient NSG (NOD.Cg-Prkdcsidc;Il2rgtm1Wjl/SzJ; JAX, 005557; 6–15 weeks old) mice. Allocation to study groups was random. Xenograft measurements were taken twice weekly using digital calipers while mice were conscious but restrained by an experimenter familiar with collecting caliper measurements of xenografts and blinded to the
experimental group designations. GFP fluorescence was visualized at the time of caliper measurement. Ellipsoid tumor volumes were calculated as volume (mm$^3$) = 0.52(length (mm) × width$^2$ (mm$^2$)), where the two longest axes, length and width, were the major and minor diameter measurements, respectively; width$^2$ represents an assumption that the xenograft depth was equivalent to the diameter of the minor axis. Statistical significance of tumor volumes at defined time points was determined by either one-way t test for all cohorts relative to shNTC or by two-way, repeated-measures ANOVA with Bonferroni correction for multiple comparisons, as indicated.

**Western blot**

Whole cell or tissue lysates were prepared in RIPA buffer supplemented with protease and phosphatase inhibitors, and samples sonicated to achieve optimum lysis. Protein concentrations were quantitated using the BCA assay, and 20 μg of lysates were loaded into 8% SDS-polyacrylamide gels. Gels were resolved at 80V for 120 mins, transferred onto nitrocellulose membrane for 70 mins at 0.35A with transfer buffer containing 20% methanol, and blocked for one hour at room temperature with 5% w/v BSA in 0.1% TBST. Membranes were incubated with primary antibodies in 5% w/v BSA in 0.1% TBST. Antibodies used in this study are: Total ERK (1:1000, Cell Signaling #9102), phospho ERK (1:1000, Cell Signaling #4695), GAPDH (1:5000, Santa Cruz sc-69778), anti-mouse and anti-goat secondary antibodies (1:10000, LiCOR IRDye 680/800 Cat. 925-68070/926-32211). Blots were imaged using the LiCOR Odyssey Fc system.

**Software**

Unless otherwise noted, bioinformatic analysis pipelines, report generation, and figure visualization were performed using bash, R (version 3.6.1), Python scripts, and GraphPad Prism 8 software (Version 8.1.1). Hierarchical clustering was performed in Python 2.7.10 with the scipy 0.13.0b1 toolbox using the Hamming distance metric with Ward’s linkage method. The R packages ggplot2 [84] and Gviz [85] were used to generate the graphics in various figure panels.

**URLs**

TgTn(sb-T2/Onc3)12740Njen; Gt(ROSA)26Sortm2(sb11)Njen double mutant mice, https://ncifrederick.cancer.gov/Lasp/MouseRepository/MouseModels/StrainDetails.aspx?StrainNum=01XGB&g=ROSA26; FVB/N-Tg(ACTB-cre)2Mrt/J mice, http://jaxmice.jax.org/strain/003376.html; Cancer Gene Census, http://cancer.sanger.ac.uk/census/; Mouse Genome Informatics (MGI) database, ftp://ftp.informatics.jax.org/pub/reports/index.html; Enrichr, http://amp.pharm.mssm.edu/Enrichr/, STRING, https://string-db.org/, The Cancer Genome Atlas (TCGA), http://cancergenome.nih.gov/; FVB.129P2-Trp53tm1Brn/Nci mice, https://ncifrederick.cancer.gov/Lasp/MouseRepository/MouseModels/StrainDetails.aspx?StrainNum=01XC2&g=Trp53; 129S4-Trp53tm2Tyj/Nci mice, https://ncifrederick.cancer.gov/Lasp/MouseRepository/MouseModels/StrainDetails.aspx?StrainNum=01XM2&g=Trp53; immunodeficient NSG (NOD.Cg-PrkdcscidIl2rgtm1Wjl/SzJ) mice, https://www.jax.org/strain/005557.

**Accession codes**

NCBI BioProject accessions: PRJNA580460 for whole-transcriptome RNA-Seq data; PRJNA580462 for microarray expression data.
Supporting information

S1 Fig. Overview of genetic crosses to generate SB|Trp53|Onc3 mouse model. Schematic of genetic crosses to generate the genetic cohorts aged for tumor development in this study. (TIFF)

S2 Fig. SB insertion patterns in activated and inactivated drivers. Recurrent TA-dinucleotide SB insertion events from individual cuSCC genomes reveal activating sense strand SB insertions in the locus of the paralogs Zmiz1 (A) or Zmiz2 (B). Chimeric fusion between the SB transposon splice donor (SD) and splice acceptor sites at exon 9 of Zmiz1 or exon 6 of Zmiz2 results in CAG-promoter driven transcription of a truncated mRNA predicted to encode N-terminally truncated Zmiz1ΔN-185 aa or Zmiz2ΔN-184 aa containing functional MIZ-type zinc finger domains and a Siz/PIAS RING finger (SP-RING). Representative SB insertion maps in cuSCC trunk drivers showing the locations of mapped SB insertions (triangles) predicted to inactivate tumor suppressors, shown in detail for Pten (C). Inactivating insertion maps for chromatin remodelers Ncoa2 (D) and Kmt2c (E). (TIFF)

S3 Fig. Evaluating the reproducibility of SBCapSeq results from bulk cuSCC and normal skin specimens. Representative plots of individual SB insertion sites based on read depth from two individual cuSCC genomes (A-B) and two unselected skin cell genomes from histologically normal skin (C-D) using genomic DNAs isolated from bulk tumor specimens from biological replicate libraries (independent library workflows applied to the same biological specimen isolate) comparing library 1 (x-axis) to library 2 (y-axis). Biological reproducibility of cuSCC specimen libraries is indicated by both Pearson’s and Spearman’s correlation metrics (A-B). No biological reproducibility of normal skin specimen libraries is observed (C-D) indicated by negative values for Pearson’s and Spearman’s correlation metrics and a failure to identify the same SB insertion sites. Individual SB insertion sites plotted by read depth from two technical replicates (the same library preparation) from two separate sequencing runs (E-F) confirms high reproducibility between individual libraries at high-to-moderate read depths, supported by both Pearson’s (r = 0.48) and Spearman’s (ρ = 0.49 or undetermined) correlation metrics. (TIFF)

S4 Fig. Hierarchical two-dimensional clustering of recurrent events in cuKA and cuSCC. Hierarchical two-dimensional clustering (Hamming distance with the Ward method of agglomeration) of recurrent genic SBCapSeq insertion events from four skin masses containing distinct cuKA and cuSCC regions. Four distinct specimen (x-axis) and gene (y-axis) clades, each pertaining to a single bi-lesional mass, demonstrate clonal identities. Zmiz1 was the only gene recurrently mutated across all samples with high read depths, highlighted in blue text. (TIFF)

S5 Fig. Curated biological pathways and processes enriched within SB-induced cuSCC. Table of significant pathways collated from pathway enrichment categories. (TIFF)

S6 Fig. Trunk driver STRING network analysis. (A) The SB|cuSCC Trunk Driver (n = 84) network has significantly more known protein-protein interactions than expected by chance (P = 8.55 × 10−6, STRING enrichment analysis; number of nodes: 144, number of edges: 124, expected number of edges: 82 (B) The SB|cuKA Trunk Driver (n = 62) network also has significantly more known protein-protein interactions than expected by chance (P = 6.93 × 10−5, STRING enrichment analysis; number of nodes: 37, number of edges: 14,
expected number of edges: 4.

(TIFF)

S7 Fig. ZMIZ1 metagene within the TCGA Head & Neck Squamous Cell Carcinoma (hnSCC) RNA-seq dataset. (A) ZMIZ1 metagene heatmap constructed using singular value decomposition of human hnSCC RNA-seq dataset from TCGA consisting of a 23-gene signature. Multivariate analysis of a gene signatures that correlates with the expression of ZMIZ1 in HNSCC. (B) Survival plots for patients with head and neck type SCC (hnSCC) based on the ZMIZ1--centric metagene.

(TIFF)

S8 Fig. Clonally selected SB insertions affect trunk driver proto-oncogene expression in SB-cuSCC genomes. (A) Summary of Zmiz1 and Zmiz2 transcripts identified from bulk analysis of cuSCC cells by wtRNA-seq are represented as the log_{10} ratio of fragments per kilobase of transcript per million (FPKM) from the average of 6 genomes with SB insertions into Zmiz1 compared to the transcripts from the single genome with Zmiz2 SB insertion, defined by FPKM mapped reads from ribo-depleted RNA. (B) Transcripts identified from bulk analysis of cuSCC cells by wtRNA-seq are represented as the log_{10} ratio of SB insertion containing compared to wild-type transcripts, defined by fragments per kilobase of transcript per million mapped reads from ribo-depleted. (C-F) Individual panels from Fig 5E: multifold induction of gene expression in cuSCC masses with (+) high read depth activating SB insertion events among 4 candidate oncogenic drivers compared with normal gene expression levels in cuSCC tumors without (−) SB insertions.

(TIFF)

S9 Fig. Clonally selected SB insertions affect trunk driver genes by inactivating expression in SB-cuSCC genomes. (A-F) Individual panels from Fig 5F: reduced gene expression in cuSCC masses with (+) high read depth inactivating SB insertion events among 6 candidate tumor suppressor drivers compared with normal gene expression levels in cuSCC tumors without (−) SB insertions.

(TIFF)

S10 Fig. CREBBP knockdown does not alter proliferation rate in cuSCC cell lines. (A) 96-hour proliferation assay (n = 3, error bars SEM). (B) In vivo xenograft assay of COLO16 shNTC or shCREBBP into NSG immunodeficient mice (n = 9 per condition).

(TIFF)

S11 Fig. Gross photographs of cuSCC xenograft masses collected at necropsy showing robust TurboGFP expression. Xenografts of cuSCC human cell lines A431, COLO16 and SCC13 with stable lentiviral constructs expressing shRNAs directed knockdown of target drivers KMT2C (shKMT2C) or NCOA2 (shNCOA2) relative to a non-targeting control (shNTC) grown in NSG immunodeficient mice.

(PDF)

S12 Fig. SB T2/Onc3 TG.12740 allele donor position mapping and exclusion for SB Driver Analysis. Cumulative SB insertion sites at TA-dinucleotides from (A) 454_Splink and (B) Ion_SBCapSeq datasets showing the likely TG.12740 donor site at 87,000,000 bp and the exclusion region (yellow box) between 79,000,000 and 95,000,000 bp applied when SB Driver Analysis was run on chromosome 9. The 81 genes between Filip1 at chr9:79663368–79825689 and Tfdp2 at chr9:96096693–96224065 were excluded from SB Driver Analysis. The complete list of 81 genes excluded from SB Driver Analysis from tumors harboring the SB T2/Onc3 TG.12740 donor allele: 1190002N15Rik, 1700034K08Rik, 1700065D16Rik, 1700065D16Rik,
4930524O08Rik, 4930554C24Rik, 4933400C23Rik, 9330159M07Rik, 9430037G07Rik, A330041J22Rik, Adams57, AF529169,ANKRD34c, ATR, B430319G15Rik, Bckdhb, Bcl2a1a, Bcl2a1b, Bcl2a1d, Cep162, Chst2, Cish, Cyb5r4, D430036f16Rik, Dopey1, Elov14, Fam46a, Gk5, Hmgn3, Htr1b, Ibtk, Impg1, Irak1bp1, Lca5, Me1, Mei4, Mir184, Mir6386, Mir7656, Morf4l1, Mrap2, Mthfs, Mthfs1, Myo6, Nt5e, Paqr9, Pcolce2, Pgm3, Phip, Plod2, Pls1, Plscr1, Plscr2, Plscr4, Plscr5, Prss35, Rasgrf1, Ripply2, Rwdd2a, Senp6, Sh3bgrl2, Slc9a9, Snap91, Snhg5, Snx14, Syncrip, Tbc1d2b, Tbx18, Tmed3, Tpbg, Trim43a, Trim43b, Trim43c, Trpc1, Ttk, U2surp, Ube2cbp, Xrn1, Zfp949, Zic1, Zic4.

S1 Table. Tumor incidence and subgroup classifications by cohort.
(XLSX)

S2 Table. Specimen metafile data for projects sequenced using SBCapSeq protocol with Ion Torrent Proton sequencer.
(XLSX)

S3 Table. Discovery and progression SB Driver Analysis for cuSCC60_SBC.
(XLSX)

S4 Table. Trunk SB Driver Analysis for cuSCC60_SBC.
(XLSX)

S5 Table. Discovery and progression SB Driver Analysis for cuKA11_SBC.
(XLSX)

S6 Table. Trunk SB Driver Analysis for cuKA11_SBC.
(XLSX)

S7 Table. Discovery and progression SB Driver Analysis for cuSK32_SBC.
(XLSX)

S8 Table. SBCapSeq read depth and analysis for 4 cuSCC genomes selected for multi-region resequencing because they had intermixing of cuSCC and cuKA histologies.
(XLSX)

S9 Table. Enrichr gene set pathway enrichment analysis of cuSCC drivers.
(XLSX)

S10 Table. Summary of 7 cuSCC transcriptomes selected for whole transcriptome RNAseq analysis.
(XLSX)

S11 Table. BED file of SBfusion insertions in 7 cuSCC genomes by whole transcriptome RNAseq analysis.
(XLSX)

S12 Table. Venn diagram for overlap of genes with SBfusion reads detected by whole transcriptome RNAseq analysis and cuSCC60_SBC discovery driver.
(XLSX)

S13 Table. Venn diagram for overlap of genes with SBfusion reads detected by whole transcriptome RNAseq analysis and all cuSCC drivers.
(XLSX)
S14 Table. Transcripts per million (TPM) normalized whole transcriptome RNAseq values per gene from RNA isolated from cuSCC genomes with and without Zmiz1 insertions.

(XLSX)

S15 Table Fragments Per Kilobase of Transcripts per Million (FPKM) normalized whole transcriptome RNAseq values per gene transcript from RNA isolated from cuSCC genomes with and without Zmiz1 insertions.

(XLSX)

S16 Table. Normalized microarray values per gene from RNA isolated from cuSCC genomes with and without Zmiz1 insertions.

(XLSX)

S17 Table. Normalized microarray values per probe from RNA isolated from cuSCC genomes with and without Zmiz1 insertions.

(XLSX)

S18 Table. All 289 genes with differential expression analysis from microarray data from RNA isolated from cuSCC genomes with and without Zmiz1 insertions with P < 0.0001 and q < 0.05.

(XLSX)

S19 Table. Lentiviral vectors containing shRNAs used in this study.

(XLSX)

S20 Table. TaqMan probes used in this study.

(XLSX)

S1 Text. Oncogenomic comparisons between SB candidate Trunk driver genes and their direct orthologs in human Cancer Gene Census.

(DOCX)

S1 Data. BED file of SB insertions for cuSCC60_SBC.

(XLSX)

S2 Data. BED file of SB insertions for cuKA11_SBC.

(XLSX)

S3 Data. BED file of SB insertions for cuSK32_SBC.

(XLSX)

S4 Data. BED file of SB insertions for 4 cuSCC genomes selected for multi-region resequencing because they had intermixing of cuSCC and cuKA histologies.

(XLSX)

S5 Data. Numerical data for graphs.

(XLSX)

Acknowledgments

The authors wish to thank the Copeland–Jenkins labs in Singapore and Houston for helpful discussions; we thank Tyler Jacks (MIT) for P53 LSL R172H mice; Anton Berns (NKI) for P53 floxed mice; Gail Martin (UCSF) for Actin-beta-Cre mice; James G. Rheinwald and the Harvard Skin Disease Research Center for the SCC13 cell line; David Adams, Theo Whipp, Richard Rance, Alistair Rust, and the Wellcome Trust Sanger Institute sequencing and informatics
teams for 454 sequencing; the Elsa Flores Lab for use of the LiCOR imaging system; Sean Yoder (Moffitt Cancer Center) and Dr. Chaomei Zhang (Moffitt Cancer Center) for providing sequencing services; Pearlyn Cheok, Nicole Lim, Dorothy Chen and Cherylin Wee for assistance with tumor monitoring and animal husbandry at IMCB (Singapore, Republic of Singapore), Hubert Lee and Erik Freiter for assistance with animal husbandry at HMRI (Houston, TX, USA), and Crystal Reed and Bethanie Gore for assistance with animal husbandry at MCC/USF (Tampa, FL, USA). Necropsy and histology were performed by the Advanced Molecular Pathology Laboratory, Institute of Molecular and Cell Biology (A’STAR), Singapore. Necropsy and histology were also performed by the Tissue Core Facility and Ion Torrent sequencing was performed by the Molecular Genomics Core at the H. Lee Moffitt Cancer Center & Research Institute.

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References

1. Pickering CR, Zhou JH, Lee JJ, Drummond JA, Peng SA, Saade RE, et al. Mutational landscape of aggressive cutaneous squamous cell carcinoma. Clin Cancer Res. 2014; 20(24):6582–92. https://doi.org/10.1158/1078-0432.CCR-14-1768 PMID: 25303977

2. South AP, Purdie KJ, Watt SA, Haldenby S, den Breems NY, Dimon M, et al. NOTCH1 mutations occur early during cutaneous squamous cell carcinogenesis. J Invest Dermatol. 2014; 134(10):2630–8. https://doi.org/10.1038/jid.2014.154 PMID: 24662767

3. Ivica Z, Hackett PB, Plasterek RH, Izsak Z. Molecular reconstruction of Sleeping Beauty, a Tcp1-like transposon from fish, and its transposition in human cells. Cell. 1997; 92(6):867–74. https://doi.org/10.1016/s0092-8674(00)80436-5 PMID: 9390559

4. Collier LS, Largaespada DA. Transposons for cancer gene discovery: Sleeping Beauty and beyond. Genome Biol. 2007; 8 Suppl 1:S15. https://doi.org/10.1186/gb-2007-8-s1-s15 PMID: 18047692

5. Copeland NG, Jenkins NA. Harnessing transposons for cancer gene discovery. Nat Rev Cancer. 2010; 10(10):696–706. https://doi.org/10.1038/nrc2916 PMID: 20844533

6. Dupuy AJ, Akagi K, Largaespada DA, Copeland NG, Jenkins NA. Mammalian mutagenesis using a highly mobile somatic Sleeping Beauty transposon system. Nature. 2005; 436(7048):221–6. https://doi.org/10.1038/nature03691 PMID: 16015321

7. Dupuy AJ, Rogers LM, Kim J, Nannapaneni K, Starr TK, Liu P, et al. A modified sleeping beauty transposon system that can be used to model a wide variety of human cancers in mice. Cancer Res. 2009; 69(20):8150–6. https://doi.org/10.1158/0008-5472.CAN-09-1135 PMID: 19808965

8. Mann KM, Jenkins NA, Copeland NG, Mann MB. Transposon Insertional Mutagenesis Models of Cancer. Cold Spring Harbor Protoc. 2013.

9. Mann KM, Newberg JY, Black MA, Jones DJ, Araya-Manzanares F, Guzman-Rojas L, et al. Analyzing tumor heterogeneity and driver genes in single myeloid leukemia cells with SBCapSeq. Nat Biotechnol. 2016; 34(9):962–72. https://doi.org/10.1038/nbt.3637 PMID: 27479497

10. Mann KM, Ward JM, Yew CC, Kovochich A, Dawson DW, Black MA, et al. Sleeping Beauty mutagenesis reveals cooperating mutations and pathways in pancreatic adenocarcinoma. Proc Natl Acad Sci U S A. 2012; 109(18):7394–41. https://doi.org/10.1073/pnas.1202491109 PMID: 22421440

11. Mann MB, Black MA, Jones DJ, Ward JM, Yew CC, Newberg JY, et al. Transposon mutagenesis identifies genetic drivers of Braf(V600E) melanoma. Nat Genet. 2015; 47(5):486–95. https://doi.org/10.1038/ng.3275 PMID: 25848750

12. Mann MB, Jenkins NA, Copeland NG, Mann KM. Sleeping Beauty mutagenesis: exploiting forward genetic screens for cancer gene discovery. Curr Opin Genet Dev. 2014; 24:16–22. https://doi.org/10.1016/j.gde.2013.11.004 PMID: 24575332

13. Rangel R, Lee SC, Hon-Kim Ban K, Guzman-Rojas L, Mann MB, Newberg JY, et al. Transposon mutagenesis identifies genes that cooperate with mutant Pten in breast cancer progression. Proc Natl Acad Sci U S A. 2016; 113(48):E7497–E505. https://doi.org/10.1073/pnas.1613859113 PMID: 27849608

14. Takeda H, Wei Z, Kosoh R, Rust AG, Yew CC, Mann MB, et al. Transposon mutagenesis identifies genes and evolutionary forces driving gastrointestinal tract tumor progression. Nat Genet. 2015; 47(2):142–50. https://doi.org/10.1038/ng.3175 PMID: 25559195

15. Mann KM, Jenkins NA, Copeland NG, Mann MB. Transposon insertional mutagenesis models of cancer. Cold Spring Harbor Protoc. 2014; 2014(3):235–47. https://doi.org/10.1101/pdb.top069849 PMID: 24591685

16. Genovesi LA, Ng CG, Davis MJ, Remke M, Taylor MD, Adams DJ, et al. Sleeping Beauty mutagenesis in a mouse medulloblastoma model defines networks that discriminate between human molecular subgroups. Proc Natl Acad Sci U S A. 2013; 110(46):E4325–34. https://doi.org/10.1073/pnas.1318691110 PMID: 24167280

17. Perez-Mancera PA, Rust AG, van der Weyden L, Kristiansen G, Li A, Sarver AL, et al. The deubiquitinating enzyme USP9X suppresses pancreatic ductal adenocarcinoma. Nature. 2012; 486(7402):266–70. https://doi.org/10.1038/nature11114 PMID: 22899621

18. Nassar D, Latil M, Boecck B, Lambrechts D, Blanpain C. Genomic landscape of carcinogen-induced and genetically induced mouse skin squamous cell carcinoma. Nat Med. 2015; 21(8):948–54. https://doi.org/10.1038/nm.3878 PMID: 26169291

19. Chitsazadeh V, Coster C, Drummond JA, Nguyen T, Joseph A, Chilukuri S, et al. Cross-species identification of genomic drivers of squamous cell carcinoma development across preneoplastic intermediates. Nat Commun. 2016; 7:12601. https://doi.org/10.1038/ncomms12601 PMID: 27574101

20. Knatko EV, Praslicka B, Higgins M, Evans A, Purdie KJ, Harwood CA, et al. Whole-Exome Sequencing Validates a Preclinical Mouse Model for the Prevention and Treatment of Cutaneous Squamous Cell
21. Newberg JY, Black MA, Jenkins NA, Copeland NG, Mann KM, Mann MB. SB Driver Analysis: a Sleeping Beauty cancer driver analysis framework for identifying and prioritizing experimentally actionable oncogenes and tumor suppressors. Nucleic Acids Res. 2018; 46(16):e94. https://doi.org/10.1093/nar/gky450 PMID: 29846651

22. Jonkers J, Meuwissen R, van der Gulden H, Peterse H, van der Valk M, Berns A. Synergistic tumor suppressor activity of BRCA2 and p53 in a conditional mouse model for breast cancer. Nat Genet. 2001; 29(4):418–25. https://doi.org/10.1038/ng477 PMID: 11694875

23. Olive KP, Tuveson DA, Ruhe ZC, Yin B, Willis NA, Bronson RT, et al. Mutant p53 gain of function in two mouse models of Li-Fraumeni syndrome. Cell. 2004; 119(6):847–60. https://doi.org/10.1016/j.cell.2004.11.004 PMID: 15607980

24. Lewandoski M, Meyers EN, Martin GR. Analysis of Fgf8 gene function in vertebrate development. Cold Spring Harb Symp Quant Biol. 1997; 62:159–68. PMID: 9598348

25. Newberg JY, Mann KM, Mann MB, Jenkins NA, Copeland NG. SBCDDB: Sleeping Beauty Cancer Driver Database for gene discovery in mouse models of human cancers. Nucleic Acids Res. 2018; 46(D1):D1011–D7. https://doi.org/10.1093/nar/gkx956 PMID: 29059366

26. Martincorena I, Rosahan A, Gerstung M, Ellis P, Van Loo P, McLaren S, et al. Tumor evolution. High burden and pervasive positive selection of somatic mutations in normal human skin. Science. 2015; 348(6237):880–6. https://doi.org/10.1126/science.aaa6806 PMID: 25999502

27. Futreal PA, Coin L, Marshall M, Down T, Hubbard T, Wooster R, et al. A census of human cancer genes. Nat Rev Cancer. 2004; 4(3):177–83. https://doi.org/10.1038/nrc1299 PMID: 14993899

28. Lee CS, Bhaduri A, Mah A, Johnson WL, Ungewickell A, Aros CJ, et al. Recurrent point mutations in the kinetochore gene KNSTRN in cutaneous squamous cell carcinoma. Nat Genet. 2014; 46(10):1060–2. https://doi.org/10.1038/ng.3091 PMID: 25194279

29. Inman GJ, Wang J, Nagano A, Alexandrov LB, Purdie KJ, Taylor RG, et al. The genomic landscape of cutaneous SCC reveals drivers and a novel azathioprine associated mutational signature. Nat Commun. 2018; 9(1):3667. https://doi.org/10.1038/s41467-018-06027-1 PMID: 30202019

30. Li C, Gao Z, Li F, Li X, Sun Y, Wang M, et al. Whole Exome Sequencing Identifies Frequent Somatic Mutations in Cell-Cell Adhesion Genes in Chinese Patients with Lung Squamous Cell Carcinoma. Sci Rep. 2015; 5:14237. https://doi.org/10.1038/srep14237 PMID: 26503331

31. Chen EY, Tan CM, Kou Y, Duan Q, Wang Z, Meirelles GV, et al. Enrichr: interactive and collaborative HTML5 gene list enrichment analysis tool. BMC Bioinformatics. 2013; 14:128. https://doi.org/10.1186/1471-2105-14-128 PMID: 23586463

32. Kuleshov MV, Jones MR, Rouillard AD, Fernandez NF, Duan Q, Wang Z, et al. Enrichr: a comprehensive gene set enrichment analysis web server 2016 update. Nucleic Acids Res. 2016; 44(W1): W90–7. https://doi.org/10.1093/nar/gkw377 PMID: 27141961

33. Stransky N, Egloff AM, Tward AD, Kostic AD, Cibulskis K, Sivachenko A, et al. The mutational landscape of head and neck squamous cell carcinoma. Science. 2011; 333(6046):1157–60. https://doi.org/10.1126/science.1208130 PMID: 21798893

34. Nadeau JH, Dudley AM. Genetics. Systems genetics. Science. 2011; 331(6020):1015–6. https://doi.org/10.1126/science.1203869 PMID: 21350153

35. Beliakov S, Zimp7 and Zimp10, two novel PIAS-like proteins, function in androgen receptor-mediated transcription and interacts with SWI/SNF-like BAF complexes. Mol Endocrinol. 2005; 19(12):2915–29. https://doi.org/10.1205/me.2005-0097 PMID: 16051670

36. Kahyo T, Nishida T, Yasuda H. Involvement of PIAS1 in the sumoylation of tumor suppressor p53. Mol Cells. 2001; 8(3):713–8. https://doi.org/10.1016/s1097-2765(01)00349-5 PMID: 11583632

37. Lee SH, Zhu C, Peng Y, Johnson DT, Lehmann L, Sun Z. Identification of a novel role of ZMIZ2 protein in regulating the activity of the Wnt/beta-catenin signaling pathway. J Biol Chem. 2013; 288(50):35913–24. https://doi.org/10.1074/jbc.M113.529727 PMID: 24174533

38. Megidish T, Xu JH, Xu CW. Activation of p53 by protein inhibitor of activated Stat1 (PIAS1). J Biol Chem. 2002; 277(10):8255–9. https://doi.org/10.1074/jbc.C200001200 PMID: 11788578

39. Schmidt D, Muller S. Members of the PIAS family act as SUMO ligases for c-Jun and p53 and repress p53 activity. Proc Natl Acad Sci U S A. 2002; 99(5):2872–7. https://doi.org/10.1073/pnas.022559499 PMID: 11867732

40. Schmidt D, Muller S. PIAS/SUMO: new partners in transcriptional regulation. Cell Mol Life Sci. 2003; 60(12):2561–74. https://doi.org/10.1007/s00018-003-3129-1 PMID: 14685683
Sharma M, Li X, Wang Y, Zarnegar M, Huang CY, Palvimo JJ, et al. hZmp10 is an androgen receptor co-activator and forms a complex with SUMO-1 at replication foci. EMBO J. 2003; 22(22):6101–14. https://doi.org/10.1093/embojd/ced585 PMID: 14609956

43. Soler G, Radford-Weiss I, Ben-Abdelali R, Mahlaoui N, Ponceau JF, Macintyre EA, et al. Fusion of ZMIZ1 to ABL1 in a B-cell acute lymphoblastic leukaemia with a t(9;10)(q34;q22.3) translocation. Leukemia. 2008; 22(6):1278–80. https://doi.org/10.1038/sj.leu.2405033 PMID: 18007576

44. Cancer Genome Atlas N. Comprehensive genomic characterization of head and neck squamous cell carcinomas. Nature. 2015; 517(7536):576–82. https://doi.org/10.1038/nature14129 PMID: 25631445

45. Potla L, Boghaert ER, Armelillo D, Frost P, Damle NK. Reduced expression of EphrinA1 (EFNA1) inhibits three-dimensional growth of HT29 colon carcinoma cells. Cancer Lett. 2002; 175(2):187–95. https://doi.org/10.1016/s0304-3835(01)00613-9 PMID: 11741747

46. Huguet EL, McMahon JA, McMahon AP, Bicknell R, Harris AL. Differential expression of human Wnt genes 2, 3, 4, and 7B in human breast cell lines and normal and disease states of human breast tissue. Cancer Res. 1994; 54(10):2615–21. PMID: 8168088

47. Huang Y, Li LP. Progress of cancer research on astrocyte elevated gene-1/Metadherin (Review). Oncol Lett. 2014; 8(2):493–501. https://doi.org/10.3892/ol.2014.2231 PMID: 25009642

48. Boukamp P, Petrussevska RT, Breitkreutz D, Hornung J, Markham A, Fusenig NE. Normal keratinization in a spontaneously immortalized aneuploid human keratinocyte cell line. J Cell Biol. 1988; 106(3):761–71. https://doi.org/10.1083/jcb.106.3.761 PMID: 2450098

49. Aiderus A, Contreras-Sandoval AM, Meshey AL, Newberg JY, Ward JM, Swing D, et al. Promoterless Transposon Mutagenesis Drives Solid Cancers via Tumor Suppressor Inactivation. Cancers 2021, 13(2), 225. https://doi.org/10.3390/cancers13020225 PMID: 33435458

50. Yu J, Wu WK, Liang Q, Zhang N, He J, Li X, et al. Disruption of NCOA2 by recurrent fusion with LACTB2 in colorectal cancer. Oncogene. 2016; 35(2):187–95. https://doi.org/10.1038/onc.2015.72 PMID: 25830207

51. Mathur R, Alver BH, San Roman AK, Wilson BG, Wang X, Agoston AT, et al. ARID1A loss impairs enhancer-mediated gene regulation and drives colon cancer in mice. Nat Genet. 2017; 49(2):296–302. https://doi.org/10.1038/ng.3744 PMID: 27941798

52. Jia D, Augert A, Kim DW, Eastwood E, Wu N, Ibrahim AH, et al. Crebbp Loss Drives Small Cell Lung Cancer and Increases Sensitivity to HDAC Inhibition. Cancer Discov. 2018; 8(11):1422–37. https://doi.org/10.1038/cdd.2018.124 PMID: 30181244

53. Dhar SS, Zhao D, Lin T, Gu B, Pal K, Wu SJ, et al. MLL4 Is Required to Maintain Broad H3K4me3 Peaks and Super-Enhancers at Tumor Suppressor Genes. Mol Cell. 2018; 70(5):825–41 e6. https://doi.org/10.1016/j.molcel.2018.04.028 PMID: 29861161

54. Cho SJ, Yoon C, Lee JH, Chang KK, Lin JX, Kim YH, et al. KMT2C Mutations in Diffuse-Type Gastric Adenocarcinoma Promote Epithelial-to-Mesenchymal Transition. Clin Cancer Res. 2018; 24(24):6556–69. https://doi.org/10.1158/1078-0432.CCR-17-1679 PMID: 30108106

55. Andricovich J, Perkail S, Kai Y, Casasanta N, Peng W, Tzatsos A. Loss of KDM6A Activates Super-Enhancers to Induce Gender-Specific Squamous-like Pancreatic Cancer and Confers Sensitivity to BET Inhibitors. Cancer Cell. 2018; 33(3):512–26 e8. https://doi.org/10.1016/j.ccell.2018.02.003 PMID: 29533787

56. Doyle B, Morton JP, Delaney DW, Ridgway RA, Wilkins JA, Sansom OJ. p53 mutation and loss have different effects on tumourigenesis in a novel mouse model of pleomorphic rhabdomyosarcoma. J Pathol. 2010; 222(2):129–37. https://doi.org/10.1002/path.2748 PMID: 20662002

57. Hanel W, Marchenko N, Xu S, Yu SX, Weng W, Moll U. Two hot spot mutant p53 mouse models display differential gain of function in tumorigenesis. Cell Death Differ. 2013; 20(7):898–909. https://doi.org/10.1038/cdd.2013.17 PMID: 23538418

58. Lang GA, Ikawatuma T, Suh YA, Liu G, Rao VA, Parant JM, et al. Gain of function of a p53 hot spot mutation in a mouse model of Li-Fraumeni syndrome. Cell. 2004; 119(6):225–37. https://doi.org/10.1016/j.cell.2004.11.006 PMID: 15607861

59. Morton JP, Timpson P, Karim SA, Ridgway RA, Athineos D, Doyle B, et al. Mutant p53 drives metastasis and overcomes growth arrest/senescence in pancreatic cancer. Proc Natl Acad Sci U S A. 2010; 107(1):246–51. https://doi.org/10.1073/pnas.0908428107 PMID: 20018721

60. Mathios D, Hwang T, Xia Y, Phallen J, Rui Y, See AP, et al. Genome-wide investigation of intragenic DNA methylation identifies ZMIZ1 gene as a prognostic marker in glioblastoma and multiple cancer types. Int J Cancer. 2019. https://doi.org/10.1002/ijc.32587 PMID: 31373686

61. Quintana RM, Dupuy AJ, Bravo A, Casanova ML, Alameda JP, Page A, et al. A transposon-based analysis of gene mutations related to skin cancer development. J Invest Dermatol. 2013; 133(1):239–48. https://doi.org/10.1038/jid.2012.245 PMID: 22832494
62. Rogers LM, Riordan JD, Swick BL, Meyerholz DK, Dupuy AJ. Ectopic expression of Zmiz1 induces cutaneous squamous cell malignancies in a mouse model of cancer. J Investig Dermatol. 2013; 133(7):1863–9. https://doi.org/10.1038/jid.2013.77 PMID: 23426136

63. Friedel RH, Friedel CC, Bonfert T, Shi R, Rad R, Soriano P. Clonal expansion analysis of transposon insertions by high-throughput sequencing identifies candidate cancer genes in a PiggyBac mutagenesis screen. PLoS One. 2013; 8(8):e72338. https://doi.org/10.1371/journal.pone.0072338 PMID: 23940809

64. Shain AH, Pollack JR. The spectrum of SWI/SNF mutations, ubiquitous in human cancers. PLoS One. 2013; 8(1):e55119. https://doi.org/10.1371/journal.pone.0055119 PMID: 23355908

65. Gozdecka M, Meduri E, Mazan M, Tzelepis K, Dudek M, Knights AJ, et al. UTX-mediated enhancer and chromatin remodeling suppresses myeloid leukemogenesis through noncatalytic inverse regulation of ETS and GATA programs. Nat Genet. 2018; 50(6):883–94. https://doi.org/10.1038/s41588-018-0114-z PMID: 29736013

66. Kadoch C, Hargreaves DC, Hodges C, Elias L, Ho L, Ranish J, et al. Proteomic and bioinformatic analysis of mammalian SWI/SNF complexes identifies extensive roles in human malignancy. Nat Genet. 2013; 45(6):592–601. https://doi.org/10.1038/ng.2628 PMID: 23644491

67. Wang L, Zhao Z, Ozark PA, Fantini D, Marshall SA, Rendleman EJ, et al. Resetting the epigenetic balance of Polycomb and COMPASS function at enhancers for cancer therapy. Nat Med. 2018; 24(6):758–69. https://doi.org/10.1038/s41591-018-0034-6 PMID: 29785026

68. Starr TK, Allaire R, Silverstein KA, Staggs RA, Sarver AL, Bergemann TL, et al. A transposon-based genetic screen in mice identifies genes altered in colorectal cancer. Science. 2009; 323(5922):1747–50. https://doi.org/10.1126/science.1163040 PMID: 19251594

69. March HN, Rust AG, Wright NA, ten Hoeve J, de Ridder J, Eldridge M, et al. Insertional mutagenesis identifies multiple networks of cooperating genes driving intestinal tumorigenesis. Nature genetics. 2011; 43(12):1202–9. https://doi.org/10.1038/ng.990 PMID: 22057237

70. Mann MB, Mann KM, Contreras-Sandoval AM, Guzman-Rojas L, Newberg JY, Meshey AL, et al. SBCapSeq Protocol manuscript files for ‘Quantifying tumor heterogeneity, clonal dynamics, and cancer driver gene evolution from Sleeping Beauty transposon mutagenesis models using SBCapSeq’ 2019.

71. Mann KM, Guzman-Rojas L, Amaya-Manzanares F, Jones DJ, Newberg JY, Jenkins NA, et al. SBCap-Seq Protocol: a method for selective cloning of _Sleeping Beauty_ transposon insertions using liquid capture hybridization and Ion Torrent semiconductor sequencing. 2016.

72. R Core Team. R: A language and environment for statistical computing. Computing RFfS, editor. Vienna, Austria: R Foundation for Statistical Computing; 2019.

73. Gautier L, Cope L, Bolstad BM, Irizarry RA. affy—analysis of Affymetrix GeneChip data at the probe level. Bioinformatics. 2004; 20(3):307–15. https://doi.org/10.1093/bioinformatics/btg405 PMID: 14960456

74. Parman C, Halling C, Gentleman R. affyQCReport: QC Report Generation for affyBatch objects. R package version. 2005; 1(0):1.

75. Ritchie ME, Phipson B, Wu D, Hu Y, Law CW, Shi W, et al. limma powers differential expression analyses for RNA-sequencing and microarray studies. Nucleic Acids Res. 2015; 43(7):e47. https://doi.org/10.1093/nar/gkv007 PMID: 25605792

76. Yu G, He QY. ReactomePA: an R/Bioconductor package for reactome pathway analysis and visualization. Mol Biosyst. 2016; 12(2):477–9. https://doi.org/10.1039/c5mb00663e PMID: 26661513

77. Langmead B, Salzberg SL. Fast gapped-read alignment with Bowtie 2. Nat Methods. 2012; 9(4):357–9. https://doi.org/10.1038/nmeth.1923 PMID: 22388286

78. Trapnell C, Pachter L, Salzberg SL. TopHat: discovering splice junctions with RNA-Seq. Bioinformatics. 2009; 25(9):1105–11. https://doi.org/10.1093/bioinformatics/btp120 PMID: 19289445

79. Szklarczyk D, Gable AL, Lyon D, Junge A, Wyder S, Huerta-Cepas J, et al. STRING v11: protein-protein association networks with increased coverage, supporting functional discovery in genome-wide experimental datasets. Nucleic Acids Res. 2019; 47(D1):D607–D13. https://doi.org/10.1093/nar/gky1131 PMID: 30476243

80. Rheinwald JG, Green H. Epidermal growth factor and the multiplication of cultured human epidermal keratinocytes. Nature. 1977; 265(5593):421–4. https://doi.org/10.1038/265421a0 PMID: 299924

81. Rheinwald JG, Beckett MA. Tumorigenic keratinocyte lines requiring anchorage and fibroblast support cultured from human squamous cell carcinomas. Cancer Res. 1981; 41(5):1657–63. PMID: 7214336
83. Vin H, Ojeda SS, Ching G, Leung ML, Chitsazzadeh V, Dwyer DW, et al. BRAF inhibitors suppress apoptosis through off-target inhibition of JNK signaling. Elife. 2013; 2:e00969. https://doi.org/10.7554/eLife.00969 PMID: 24192036

84. Wickham H. Ggplot2: elegant graphics for data analysis. New York: Springer; 2009. viii, 212 p.

85. Henare K, Wang L, Wang LC, Thomsen L, Tjono S, Chen CJ, et al. Dissection of stromal and cancer cell-derived signals in melanoma xenografts before and after treatment with DMXAA. Br J Cancer. 2012; 106(6):1134–47. https://doi.org/10.1038/bjc.2012.63 PMID: 22415295