An attractor state zone precedes neural crest fate in melanoma initiation

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The field cancerization theory suggests that a group of cells containing oncogenic mutations are predisposed to transformation[1, 2]. We previously identified single cells in $BRAF^{V600E};p53^{-/-}$ zebrafish that reactivate an embryonic neural crest state before initiating melanoma[3-5]. Here we show that single cells reactivate the neural crest fate from within large fields of adjacent abnormal melanocytes, which we term the "cancer precursor zone." These cancer precursor zone melanocytes have an aberrant morphology, dysplastic nuclei, and altered gene expression. Using single cell RNA-seq and ATAC-seq, we defined a distinct transcriptional cell attractor state for cancer precursor zones and validated the stage-specific gene expression initiation signatures in human melanoma. We identify the cancer precursor zone driver, ID1, which binds to TCF12 and inhibits downstream targets important for the maintenance of melanocyte morphology and cell cycle control. Examination of patient samples revealed precursor melanocytes expressing ID1, often surrounding invasive melanoma, indicating a role for ID1 in early melanomagenesis. This work reveals a surprising field effect of melanoma initiation *in vivo* in which tumors arise from within a zone of morphologically distinct, but clinically covert, precursors with altered transcriptional fate. Our studies identify novel targets that could improve early diagnosis and prevention of melanoma.

Melanoma incidence has risen sharply over the past 30 years; however multiple studies show high disagreement among pathologists in the diagnoses of early potential precursor melanocyte lesions[6-9]. The cancer attractor state theory presented by Stuart Kauffman 50 years ago hypothesizes that stable gene expression profiles arise from complex gene expression networks to reinforce malignant cell phenotypes[10-12]. However, due to the difficulty of visualization and

detection, the earliest stages in melanoma initiation have not been extensively studied in either

human or animal models on a morphological or transcriptional level.

Over 80% of melanomas contain activating mutations in the MAPK pathway, such as

BRAF^{V600E}, which are often found alongside tumor suppressor loss-of-function mutations[13, 14]. We

developed a zebrafish melanoma model in which melanocytes express BRAF^{V600E} and are

deficient for *p53*. Despite the massive melanoma-prone cancerized field generated,

BRAF^{V600E};p53$^{-/-}$ fish only develop one to three melanomas in their lifetime, with almost no

additional genetic mutations[3, 15]. In contrast to conventional dogma that melanomas arise from

melanocytic nevi, there is growing recognition that most melanomas arise de novo, despite the fact

that 70-90% of nevi carry the BRAF^{V600E} mutation[13]. Thus, many potential melanoma precursors

may be clinically covert and are molecularly undefined. Moreover, the disconnect between the

acquisition of oncogenic mutations and the initiation of tumorigenesis indicates that additional

pathways are required to drive tumor initiation.

Melanoma arises due to malignant transformation of melanocytes, which are

embryonically derived from the neural crest. We previously developed a neural crest reporter,

crestin:EGFP, and showed that individual *BRAF^{V600E};p53$^{-/-}$* melanocytes reactivate the neural crest

progenitor state to form a tumor[3, 4]. Here we identify a novel intermediate step in melanoma

initiation and visualize melanoma arising from within morphologically and transcriptionally

aberrant cancer precursor zone melanocytes, which are driven by ID1 inhibition of TCF12. These

cancer precursor zones define cells that are transitioning to the cancer attractor state, which drives

and reinforces the epigenetic change of normal melanocytes into stable neural crest reactivated

melanoma cells.

Melanoma forms in cancer precursor zones

In order to characterize the earliest stages of melanoma formation, we generated *p53* deficient, *crestin:EGFP* zebrafish in which melanocytes express BRAF^{V600E}, are marked red by *mitfa:mCherry,* and are devoid of pigment due to a *tyrosinase* CRISPR knockout (*mitfa:BRAF^{V600E};p53$^{-/-}$;mitfa$^{-/-}$;crestin:EGFP;mcr:Empty;mitfa:mCherry;tyr$^{-/-}$*, referred to as *BRAF;p53*)[3, 16, 17]. We performed time course imaging and found that prior to neural crest reactivation, regions of hundreds of morphologically aberrant melanocytes express high levels of *mitfa:mCherry*, which we term the cancer precursor zone (Fig. 1a, Extended Data Fig. 1a-c). We closely followed a cohort of 13 fish that exhibit varying kinetics of CPZ formation and neural crest reactivation with an average onset of 11- and 15-weeks post-fertilization (wpf), respectively (Fig. 1b, Extended Data Fig. 1d). Individual *crestin* positive cells subsequently arose from within a cancer precursor zone to form a tumor (Fig. 1a, Extended Data Fig 1b,c, Extended Movie 1). While not all cancer precursor zones reactivate *crestin:EGFP* and form a tumor, once the neural crest program is reactivated, every fish will inevitably go on to form a melanoma (Fig. 1b). We correlated our imaging with histologic analysis of mitfa and crestin across melanoma initiation stages, confirming that individual *crestin* positive cells give rise to tumors (Fig. 1c). We confirmed cancer precursor zone formation around *crestin* patches in two additional zebrafish melanoma models without p53 inactivation, *casper;mcr:BRAF^{V600E}* fish and *casper;mcr:NRAS^{Q61R}* zebrafish suggesting CPZs form in multiple melanoma genotypes and are not dependent on p53 loss (Extended Data Fig. 1e, 1f).

Cancer precursor zones have aberrant morphology

To characterize how cancer precursor zone cells lead to neural crest reactivation, we

evaluated the anatomical location of CPZs. In line with previous studies, CPZs arise across the

entire fish body with more frequent localizations in the head, dorsal fin, and tail base[21] (Extended

Data Fig. 2a). As anatomic position did not play major a role in CPZ formation, we next analyzed

if CPZ size or if *mitfa:mCherry* intensity drove neural crest reactivation. CPZs exhibited a range

of sizes ($0.54\pm0.27mm^2$) and *mitfa:mCherry* intensities (105.6±27.4 a.u.) prior to neural crest

reactivation (Extended Data Fig. 2b,c). These data indicate neural crest reactivation is a

consequence of cell morphology and/or cell state rather than anatomic locations or physical

features of the zone. To address this, we imaged *BRAF;p53* cancerized field melanocytes and CPZ

melanocytes for morphology. Despite the presence of oncogenic mutations, the cancerized field

melanocytes appear morphologically normal, with proper dendrite formation and spacing

throughout the skin (Extended Data Fig. 3a, left). However, upon CPZ formation, the melanocytes

exhibit aberrant morphology with loss of dendrites (Extended Data Fig. 3a, right). Histologic

analysis shows that cancer precursor zone melanocytes encompassing *crestin:EGFP+* cells are

highly proliferative with dysplastic nuclei (Extended Data Fig. 3b,c). Our zebrafish model reveals

a novel intermediate step between cancerized field melanocytes and tumor initiation, where

morphologically abnormal cancer precursor zone melanocytes encompass the cell undergoing

neural crest progenitor state activation as it transforms.

Neural crest reactivation upon transplant

A hallmark of malignancy is the ability of a tumor cell to engraft upon transplantation. To

determine if cancer precursor zones have malignant potential, we FACS isolated normal

melanocytes, cancerized field melanocytes, cancer precursor zone cells, small *crestin+* patch cells,

108 or tumors (Extended Fig. 4a). These cells are enrobed in Matrigel and injected under the skin of

109 irradiated recipient fish; a limit dilution assay was performed at 14 days post-transplant. Cancer

110 precursor zone cells engrafted at about half the rate of *crestin+* cells from small patches or tumors

111 while normal or cancerized field melanocytes did not engraft (Fig. 1d, Extended Data Fig. 4b).

112 Small *crestin* patch cells engrafted at a similar rate to tumor cells, suggesting that neural crest

113 reactivated cells have similar malignant potential. Intriguingly, we found that engrafted cells from

114 cancer precursor zones subsequently expressed *crestin:EGFP*, even though FACS isolated cancer

115 precursor zone cells were *crestin:EGFP-* upon transplantation (Extended Data Fig. 4c). These data

116 reveal that cancer precursor zone cells have malignant potential due to their ability to reactivate

117 the neural crest state and engraft but have a significantly decreased tumor initiating potential.

Melanoma initiation atlas

120 *BRAF;p53* zebrafish only develop one to three melanomas over their lifetime with minimal

121 additional genetic mutations[15], indicating that transcriptional or epigenetic changes are key drivers

122 of melanoma initiation. To explore this, we performed single-cell RNA-seq (scRNA-seq). Cells

123 isolated from *BRAF;p53* zebrafish skin containing either morphologically normal control

124 cancerized field melanocytes or small *crestin* patches with surrounding cancer precursor zones

125 were run through the inDrop scRNA-seq platform, doublets removed, then partitioned into 18

126 clusters and annotated using known markers (Fig 2a, Extended Data Fig. 5a-c). Epithelial and

127 immune gene signatures largely overlap between the control and *crestin+* sample; however, the

128 melanocyte clusters differentially segregate into cancerized field melanocytes, cancer precursor

129 zone, and *crestin+* melanoma populations. Pseudotime analysis highlights cellular reprogramming

130 from cancerized field melanocytes to crestin+ melanoma with CPZ melanocytes as a transitory

population (Fig. 2b, Extended Fig. 5d). We then re-clustered selected melanocytes and assessed

differential gene expression. Despite the oncogenic mutations, control cancerized field

melanocytes (mel) highly express expected terminally differentiated melanocyte markers, *dct,*

pmela, and *tryp1b* (Fig. 2c). Cancer precursor zone (CPZ) cells begin to express neural crest

markers and *crestin+* cells no longer express differentiated melanocyte markers and instead

resemble embryonic neural crest cells (Fig. 2c, 2d). These analyses reveal that the transformation

of a terminally differentiated melanocyte occurs through a transcriptional shift through a cancer

precursor zone and towards a neural crest progenitor state.

To verify these signatures in melanoma initiation, we performed RNA-seq on FACS

isolated *mitfa:mCherry+* control cancerized field melanocytes and compared them to

mitfa:mCherry+ cancer precursor zone melanocytes and *mitfa:mCherry+/crestin:EGFP+* cells

isolated from small, medium, and large *crestin* patches in *BRAF;p53* zebrafish. CPZ induction did

not cause a global impact on gene expression relative to cancerized field with only 502 genes

(2.8% [n=17,483], $log_2FC\pm2$, p-value <0.001) altering in expression, indicating this transcriptional

switch is specific to progression in melanomagenesis (Extended Data Fig. 6a). Similarly, the

transition from cancerized field to patch and tumor led to differential expression of 1,438 (8.2%)

and 1,670 (9.6%) genes, respectively (Extended Data Fig. 6b, 6c). Due to these changes in gene

expression, PCA analysis shows that the sorted melanocytes cluster according to stage, with the

cancer precursor zone melanocytes overlapping both the cancerized field and *crestin* patch clusters

(Extended Data Fig. 6d). This suggests that the cancer precursor zone cells represent a transition

state between the cancerized field and neural crest reactivated cells.

Stage-specific chromatin states define initiation

The distinct gene expression signatures separating melanocytes, cancer precursor zones, and

crestin positive cells indicates that melanoma initiation occurs through unique transcriptional cell

states. Differential expression analysis of both the scRNA-seq and RNA-seq data revealed stage-

specific expression signatures (Fig. 2e, Extended Data Fig. 7a). Cancer precursor zone cells have

increased expression of BMP signaling (red), as well as neural plate border and pre-migratory

neural crest transcription factors (green). These expression profiles were correlated with gene

expression at various stages of human melanoma formation from two published datasets (Extended

Data Fig. 7b, 7c)[22, 23].

To dissect the chromatin landscape, we performed ATAC-seq on sorted melanocytes from

various stages of initiation. We filtered called peaks for direct overlap or within a 5000 bp distance

of annotated genes for differential accessibility analysis (multiple tests corrected with FDR <0.05).

Like RNA-seq, ATAC-seq showed cellular reprogramming of specific loci. Only 999 genes (2.8%

[n=35,592], $log_2FC\pm2.0$, p-value <0.001) had differentially accessible regions (DAR) in cancer

precursor zone melanocytes relative to cancerized field. While neural crest reactivation in patches

and tumors led to more chromatin reprogramming (1,117 DAR in patch; 8,612 DAR tumor), all

melanoma initiation stages showed focal chromatin changes specific to cancer progression

(Extended Data Fig. 8a-c). Promoter and enhancer regions associated with differentially expressed

genes contained peaks that opened in a stage-specific manner (Fig. 2f, Extended Data Fig. 9a-g).

The chromatin accessibility findings correspond with the gene expression data and further indicate

that a transcriptionally driven cancer attractor state is present at the earliest stages of melanoma

initiation.

Cancerized field lacks MAPK activation

Most human melanomas arise de novo, with only a small fraction of nevi progressing to

malignancy[24]. However, 70-90% of benign nevi carry the BRAF[V600E] mutation and activation of

the MAPK pathway does not always directly correlate with BRAF mutations in many

melanomas[24-27]. Despite the presence of constitutively active BRAF in our zebrafish model, the

expression levels of MAPK pathway members and downstream targets are upregulated exclusively

in cancer precursor zone and *crestin+* cells (Extended Data Fig. 10a). Negative regulators of

MAPK, *spry2*, *spry4*, *dusp2*, and *dusp6*, show little to no expression and closed chromatin in the

cancerized field melanocytes (Extended Data Fig. 10a, 10b). ATAC-seq analysis on isolated

melanocytes revealed that the enhancer regions around MAPK target genes are not open in control

cancerized fields but become open in cancer precursor zones (Extended Data Fig. 10b).

Furthermore, phospho-ERK immunohistochemistry showed no staining in cancerized field

melanocytes, low staining in *mitfa*[med] cancer precursor zones, and high staining in *mitfa*[high] cancer

precursor zones (Extended Data Fig. 10c). We considered the possibility that the *mitfa*[high] state

increases the amount of BRAF[V600E] expression due to the *mitfa:BRAF[V600E]* transgene; however,

RNA-seq on isolated melanocytes shows no change in BRAF[V600E] expression between the

cancerized field and cancer precursor zone stages, only increasing at the patch stage (Extended

Data Fig. 10d). Together these data indicate that the MAPK pathway is not active until the cancer

precursor zone stage and is further activated during tumor initiation.

ID1 identifies and drives precursor zones

Given that cancer precursor zones form prior to and have a distinct transcriptional state from

crestin patches, we hypothesized these states are independently regulated. Since cancer precursor

zones had high expression of BMP pathway members and downstream targets (Fig. 2e), we used

the MiniCoopR system to overexpress ID1 in melanocytes[16]. ID1 overexpression significantly

increased cancer precursor zones with no increase in *crestin* patches at 6 weeks of age (Fig. 3a,

3b). We confirmed upregulation of ID1 expression in cancer precursor zones by RNAscope in situ

hybridization[28]. ID1 is highly expressed in CPZ melanocytes but not in adjacent cancerized field

cells (Extended Data Fig. 11a, 11b). These data indicate ID1 and BMP signaling are strong drivers

of cancer precursor zone formation. Unlike CPZs, *crestin+* patches exhibit high expression of IRF3

pathway members and downstream targets (Extended Data Fig. 11c). Indeed, IRF3 overexpression

caused a significant increase in the number of *crestin* patches without a change in cancer precursor

zone formation at 6 weeks post-fertilization (Fig. 3a, 3b). While ID1 or IRF3 overexpression did

not cause an earlier incidence of tumor formation, both led to a significant increase in the number

of tumors per fish (Fig. 3c, 3d). Overall, these results provide evidence that the stages of melanoma

initiation can be independently regulated, and perturbation of these stages lead to an increase in

tumor formation.

To verify the role of ID1 and IRF3 in human melanoma, we performed

immunofluorescence staining on patient samples (Fig. 3e). The percent of melanocytes expressing

ID1 and the intensity of ID1 staining was significantly increased in patient samples with atypical

melanocyte proliferative (potentially precursor) zones, melanoma in situ, or invasive melanoma

compared to those in normal skin (Fig. 3f, Extended Data Fig. 11d). Interestingly, ID1+ cells were

frequently found in the surrounding epithelial borders of invasive melanoma. In addition to the

ability of ID1 to initiate cancer precursor zones in the zebrafish, the appearance of ID1 in over

80% of atypical melanocytic hyperplasia fields indicates that ID1 is a major driver of cancer

precursor zones. Additionally, the percent of melanocytes with nuclear IRF3 staining increased at

both early and late stages of initiation, with a striking increase in staining in invasive melanoma

samples (Fig. 4e, 4f). These data validate the role for ID1 and IRF3 in melanomagenesis and

identify clinically covert ID1+ precursor melanocyte zones in humans.

ID1 inhibits TCF12 to drive initiation

ID1 is overexpressed in over 20 types of cancer, including melanoma, where strong ID1 expression

was significantly associated with increased tumor thickness and decreased patient survival[29, 30].

ID1 is a transcriptional repressor that prevents basic helix-loop-helix (bHLH) transcription factors

from binding to DNA[31]. We performed IP-MS on A375 melanoma cells overexpressing V5-tagged

ID1 or Clover control. The bHLH factor TCF12 was the most significantly pulled down protein

with ID1 (Fig. 4a, Extended Data Fig. 12a). When performing ATAC-seq on sorted melanocytes

isolated from ID1-overexpressing zebrafish, we found significant enrichment of the TCF12 motif

(ACAGCTG) under chromatin peaks that decreased with ID1 overexpression, indicating ID1 is a

repressor of TCF12 signaling (HOMER, p=1e-20). Predicted downstream targets of TCF12 are

significantly decreased in cancer precursor zones where ID1 is most highly expressed (Extended

Data Fig. 12b)[32]. Next, we performed RNA-seq on sorted melanocytes from zebrafish

overexpressing ID1 and identified potential downstream targets in the context of melanoma

initiation that were downregulated either with ID1-overexpression or in cancer precursor zones

(Extended Data Fig. 12c-d). TCF12 motifs were found under open chromatin peaks in the

promoter/enhancer regions associated with these genes (Fig. 4b). Strikingly, many of these genes

are important in the maintenance of dendritic processes and cell adhesion, such as *rhogb*, *parvg*,

lamc1, and *enah*. During cancer precursor zone formation, the zebrafish melanocytes attain an

aberrant morphology, including a loss of dendrites, which is likely achieved via a decrease in these

targets. When performing gene ontology analysis on downregulated genes containing TCF12

motifs, pathways arose such as neural crest cell migration (-6.002 LogP) and pigment cell

differentiation (-4.547 LogP), including genes such as *ednrba*, *slc45a2*, and *mlpha*.

To confirm these findings and identify targets of TCF12, we performed chromatin

immunoprecipitation following V5-tagged ID1 or Clover overexpression in immortalized human

primary melanocytes (PMEL[33]) harboring BRAF[V600E]. ID1 overexpression led to a marked

reduction in TCF12 occupancy on chromatin and loss of TCF12 motif enrichment relative to

control (Fig. 4c, 4d). This marked decrease in TCF12 occupancy focally affected neuronal genes

and melanocyte development genes. In an unbiased analysis, gene ontology analysis of peaks

within genes with annotated TCF12 motifs highlighted an enrichment of neuronal genes, including

cell body, synapse regulation, and somatodendritic compartment regulation in PMEL cells

(Extended Data Fig. 12e). Specifically, ID1 overexpression led to loss of TCF12 occupancy at

known neuronal genes, and tumor suppressor genes, DLGAP4 and UNC5A. High DLGAP4

expression is a known positive prognostic biomarker in melanoma and regulates synaptic signaling

[34-36]. Further, UNC5A is a netrin receptor that controls neurite outgrowth in the developing nervous

system but loss of UNC5A has established roles in malignant transforming mechanisms such as

anchorage independent growth and escape of apoptosis in tissues outside the nervous system[37, 38].

In addition to known markers of the neural lineage, TCF12 occupancy was lost at melanocyte

markers, such as PKNOX2 and PRKCZ. PKNOX2 is an identified tumor suppressor gene in solid

tumors, such as gastric cancer, with known roles in neural crest differentiation[39, 40]. Further,

PKNOX2 has been shown to be uniquely enriched in human melanocytes relative to other skin

cells and functions in melanin production[41]. Similarly, PRKCZ, a member of the protein kinase C

family, drives melanin production in melanocytes[42]. As such, loss of PRKCZ highlights that CPZs

may represent unpigmented clinically covert precursor lesions.

To explore how modulation of the ID1/TCF12 axis affected cell viability in primary

melanocytes and transformed melanoma cells, we measured cell proliferation in human PMEL and

A375 melanoma cells. Overexpression of ID1 in PMEL cells led to a significant increase in cell

proliferation relative to Clover control, likely due to the loss of development and cell cycle

regulators (Fig. 4e, 4f; $p<0.01$).   Highlighting the importance of TCF12 loss in driving the

malignant program, we observed a significant loss of cell proliferation and viability in A375

human melanoma cells following TCF12 overexpression relative to Clover control (Fig. 4g,

$p<0.001$). These data suggest restoring TCF12 activity can reverse malignant potential in both

dysplastic melanocytes and transformed malignant cells via ID1 or TCF12 overexpression,

respectively. Finally, we overexpressed TCF12 in zebrafish using the MiniCoopR system and

found a significant decrease in the number of CPZs formed and tumor onset compared to control

(Fig. 4h, 4i). In total, these data support the hypothesis that ID1 binds to and inhibits TCF12,

preventing it from activating downstream targets thereby enabling morphological and

transcriptional changes to form cancer precursor zones.

Discussion

Our studies support a model in which a cancerized field of morphologically normal melanocytes

expand, become dysplastic, and reactivate a neural crest progenitor program, culminating in

tumorigenesis. The data presented here add new insights to our understanding of melanomagenesis

by visualizing the earliest stages in initiation, revealing that a group of melanocytes undergo

independent, transcriptionally driven cell state changes before one cell becomes malignant. These

findings are in line with cell of origin studies that show the transformation of interfollicular and

differentiated melanocytes in the mouse tail[43]. Gene expression analysis and immunofluorescence

staining show that cancer precursor zones in zebrafish resemble atypical melanocytic proliferative

lesions that constitute presumed precursor fields that often surround melanomas in patient samples.

The strong correlation between species suggests that melanoma initiates through transcriptionally

driven cell state changes, with ID1 inhibition of TCF12 and IRF3 as drivers of this process. We

demonstrate that ID1 represses downstream targets of TCF12, which are key to the maintenance

differentiated melanocytes and cell cycle control.

The field cancerization theory posits that pro-tumorigenic mutations increase the malignant

transformation potential of cells. Potential precursor zones of frank malignancy have been

identified in several different organs, including Barrett's esophagus, ductal carcinoma in situ,

prostatic intraepithelial neoplasia, the uterine cervix, and dysplastic lesions in melanoma[44-47]. The

distinct transcriptional and chromatin states we identified during melanoma initiation are

reminiscent of the cancer attractor state theory, which posits that cancer cells sit in a low energy

basin of attraction that drives and reinforces epigenetic change[10, 11, 48]. The tight, distinct clustering

of *crestin+* melanocytes in our single cell RNA-seq analysis supports the cancer attractor state

theory. The finding that a cancer precursor zone encompasses the initiating cell of melanoma

suggests that other cancers may also initiate from within a surrounding group of precursor cells

and drive towards a cancer attractor state.

Author Contributions

A.M.M., M.H.C., and L.I.Z. conceived of the study; L.I.Z. and G.F.M. supervised the research;

A.M.M. and M.H.C. designed and performed experiments with the assistance of J.B., E.W., I.F.G.,

E.L., C.B., and P.A.; H.R.N. performed ID1 overexpressing RNA-seq experiment and advised on

experimental design for all studies; J.K.M. assisted with histology and pathologic assessment of

fish and human samples; M.P. and S.Y. performed computational analysis for all sequencing

experiments; C.G.L. provided human samples and pathologic analysis.

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Methods

Generation of BRAF;p53 zebrafish

mitfa:BRAF^{V600E};p53$^{-/-}$;mitfa$^{-/-}$;crestin:EGFP;mcr:Empty;mitfa:mCherry;tyr$^{-/-}$ zebrafish (referred

to as *BRAF;p53*) were generated by injecting *mcr:Empty* and *mitfa:mCherry* at 25 ng/μL, 5

mg/mL Cas9 protein (PNA Bio CP02), a gRNA targeting *tyr* (GGACTGGAGGACTTCTGGGG)

at 50 ng/μL, and Tol2 mRNA at 20 ng/μl into the single cell stage of *mitfa:BRAF^{V600E};p53$^{-/-}$;mitfa$^-$*

$^{/-}$;crestin:EGFP embryos[49]. These fish were originally developed to be used in combination with

MiniCoopR (mcr), which is a transgenic tool that permits selective mis-expression of any gene in

melanocytes, thereby allowing us to assess the ability of each gene to promote or inhibit melanoma

initiation[16]. Injection of the MiniCoopR vector mosaically rescues melanocyte development via an

mitfa minigene, allowing melanomas to form. The MiniCoopR vector also contains the open

reading frame of any candidate gene driven specifically in melanocytes by the *mitfa* promoter,

which enables us to assess its oncogenic potential. For the morphological and molecular

characterization experiments, a MiniCoopR vector containing only the *mitfa* minigene was

injected to rescue melanocyte formation (mcr:Empty).

mitfa:BRAF^{V600E};p53$^{-/-}$;mitfa$^{-/-}$;crestin:EGFP   zebrafish   were   generated   by   crossing

mitfa:BRAF^{V600E};p53$^{-/-}$;crestin:EGFP[3] zebrafish with *mitfa:BRAF^{V600E};p53$^{-/-}$;mitfa$^{-/-}$* zebrafish.

mitfa:mCherry was cloned by Gateway reaction using the zebrafish *mitfa* promoter and the Tol2

mCherry middle entry plasmids. The *tyrosinase* gRNA was synthesized using SP6 in vitro

transcription[17, 50]. In brief, the *tyr* oligo template (CCTCCATACGATTTAGGTGACACT

ATAGGACTGGAGGACTTCTGGGGGTTTTAGAGCTAGAAATAGCAAG) and the constant

oligonucleotide (AAAAGCACCGACTCGGTGCCACTTTTTCAAGTTGATAACGGACTAGC

CTTATTTTAACTTGCTATTTCTAGCTCTAAAAC) were annealed and filled in with T4 DNA

polymerase (New England BioLabs, M0203S). The product was PCR amplified, gel purified, and

transcribed using MEGAscript T7/SP6 (ThermoFisher Scientific, AM1333). gRNAs were cleaned

up using Direct-zol RNA Miniprep kit (Zymo Research, R2051).

To generate the *casper;mcr:NRAS^{Q61R}* or *casper;mcr:BRAF^{V600E}* zebrafish, *mcr:NRAS^{Q61R}* or

mcr:BRAF^{V600E} was injected into *casper;crestin:EGFP* embryos along with Tol2 mRNA and

mitfa:mCherry at a concentration of 25 ng/µL.

This study was performed in strict accordance with the recommendations in the Guide for the Care

and Use of Laboratory Animals of the National Institutes of Health. The animal research protocol

was approved by the Institutional Animal Care and Use Committee of Boston Children's Hospital.

All zebrafish used in this study were maintained and euthanized under the guidelines of the

Institutional Animal Care and Use Committee of Boston Children's Hospital.

Histology

Fish were euthanized and fixed in 4% paraformaldehyde overnight at 4°C. Paraffin embedding,

sectioning, Hematoxylin and Eosin (H&E) staining were performed according to standard

techniques by the Brigham & Women's Hospital Pathology Core. Immunohistochemistry was

performed with 5 µm thick formalin-fixed, paraffin-embedded tissue sections using the Leica

Bond III automated staining platform and the Leica Biosystems Refine Detection Kit. Antibody

mCherry from Thermo Fisher, catalog number M11217, clone 16D7, was run at 1:200 dilution

with EDTA antigen retrieval and Goat Anti-Rat IgG secondary antibody from Vector Labs catalog

number PI-9401-.5. Antibody GFP from Abcam, catalog number 6556, polyclonal, was run at

1:800 dilution with EDTA antigen retrieval. Antibody PCNA from Cell Signaling Technology,

catalog number 2586, clone PC(10), was run at 1:16000 dilution with citrate antigen retrieval.

Antibody Phospho ERK (p44/42 MAPK), clone D13.14.4E, from Cell Signaling Technologies,

catalog number 4370, was run at 1:150 dilution with citrate antigen retrieval.

All human tissue samples were derived from the Pathology Archives of the Brigham and Women's

Hospital with full institutional review board approval. Patient consent for experiments was not

required because de-identified pathological specimens of human samples are discarded material

by our institution, and thus the studies were exempt. Immunofluorescence studies were performed

on paraffin-embedded sections of formalin-fixed tissue. Tissue sections cut at 5 µm intervals were

deparaffinized, rehydrated, and heated with Target Antigen Retrieval Solution (Dako, Agilent

Technologies) in a pressure cooker. Sections were blocked with 10% animal serum for 30 minutes

before incubation with primary antibodies. Samples were incubated overnight at 4°C with the

following primary antibodies: mouse IgG1 anti-IRF3 (1:100; SCBT, sc-33641), mouse IgG2a anti-

ID1 (1:500; SCBT, sc133104), mouse IgG2b anti-MART1 (1:1, Biolegend, 917902). Sections

were then treated with 0.1% Sudan Black (Abcam) for 10 minutes to remove autofluorescence.

The following secondary antibodies were used: goat anti-mouse IgG2b AF488 (1:2000;

Invitrogen, A-21141), goat anti-mouse IgG2a AF594 (1:2000; Invitrogen, A-21135), goat anti-

mouse IgG1 AF647 (1:2000; Invitrogen, A-21235). Slides were mounted with Floromount G with

NucBlue.

In Situ Hybridization (RNAscope)

RNAscope Multiplex Fluorescent Assay[28] (Biotechne) was performed on formalin-fixed paraffin

embedded cancerized field, cancer precursor zone, patch, and tumor sections. Melanoma stages

were assigned by H&E and mCherry or GFP expression (IHC methods described above). Protocol

was followed according to manufacturer's instructions except protease 3 was used instead of

protease 4. Each sample was hybridized with zebrafish id1 (#517531, C1 probe, red) and zebrafish

crestin (#534061, C2 probe, far red) alongside a DAPI stain to mark nuclei. RNAscope was

provided by the Neurobiology Imaging Facility (NIF) at Harvard Medical School. Stained slides

were imaged with a 40x objective on a Nikon Eclipse Ti-2 spinning disk confocal microscope. All

images were acquired using NIS-Elements (Nikon) and minimally processed using Imaris.

Imaging

Zebrafish were anesthetized with 4% MS-222 (Pentair, TRS1) and imaged on a Nikon SMZ18

Stereomicroscope or a Nikon C2si Laser Scanning Confocal using a 10x objective. Histologic

sections were imaged on a Nikon Eclipse Ti-2 Spinning Disk Confocal using a 40x or 100x

objective. Maximum intensity projections of Z stacks or three-dimensional re-constructions are

presented here. Images were minimally processed using Photoshop, FIJI, or Imaris. Multiple tiled

images of adult zebrafish and histology were stitched together by using the automated Photomerge

function in Photoshop. Given the variability in the rescue specific to each animal, number of

zones/patches per fish were weighted for *mitfa:mCherry* rescue.

FACS Cell Isolation

mitfa:BRAF^{V600E};p53$^{-/-}$;mitfa$^{-/-}$;crestin:EGFP;mcr:Empty;mitfa:mCherry;tyr$^{-/-}$ zebrafish from the

same cohort were categorized by stage: control cancerized field, cancer precursor zone, small

crestin patch, medium *crestin* patch, and large *crestin* patch/tumor. The zones and patches were

manually dissected and individually chopped for 1-2 minutes. The finely chopped tissue was

suspended in 3 mLs of TrypLE Express (ThermoFisher Scientific, 12605028) and incubated at

37°C shaking at 300 rpm for 30 minutes. Samples were filtered through a 40 µm filter and washed

with 5 mL of FACS buffer, consisting of DPBS (ThermoFisher Scientific, 14190144), 10x

Penicillin Streptomycin (ThermoFisher Scientific, 15140122), and 2% heat-inactivated FBS

(ThermoFisher Scientific, A3840001). Samples were centrifuged at 500 rcf for 5 minutes and

resuspended in 150 µL FACS buffer. Samples were stained with SYTOX blue (ThermoFisher,

S34857) immediately before sorting for live (SYTOX blue negative), mCherry and/or EGFP

positive cells on a FACSAria II (BD Biosciences). FACS cell isolation was used for bulk and

single cell RNA-seq, ATAC-seq, and transplantation experiments.

432
433
scRNA-seq
Cells isolated from the skin of 6 control fish and 4 fish with small *crestin* patches were FACS

sorted for *mitfa:mCherry*, as described above. 1360 melanocytes were combined with 29,000

mCherry negative skin cells in the control and 6224 melanocytes were combined with 24,000

mCherry negative skin cells in the *crestin* patch sample to yield 2 samples each with 30,000 cells.

These samples were run through the inDrops platform by the Single Cell Core at Harvard Medical

School, as previously described[53, 54]. Libraries were run on an Illumina HiSeq 4000 with paired

end 150bp. inDrops single-cell RNA-seq analysis follows the instruction as described in

https://github.com/indrops/indrops. inDrops Library v3 requires manual demultiplex raw bcl into

different samples. Zebrafish Bowtie transcriptome index was built based on Ensembl GRCz11

genome sequence and gene annotation[55]. The output data matrix contains the raw count of each

gene for each cell barcode. Analysis was performed using the Seurat and monocle 3 packages in

R[56-58].

RNA-seq

RNA was extracted from 5,000 sorted melanocytes from 3-4 control or ID1-overexpressing fish

per stage using RNeasy Micro Kit (Qiagen, 74004). Ultralow input RNA-seq was performed using

the SMART-Seq v4 Ultra Low Input RNA kit for Sequencing (Clontech, 634888) and Nextera XT

DNA Library Preparation Guide (Illumina, FC-131-1024). Libraries were run on an Illumina

HiSeq 4000 with paired end 150bp. Quality control of RNA-Seq datasets was performed by

FastQC and Cutadapt to remove adaptor sequences and low-quality regions[59, 60]. The high-quality

reads were aligned to Ensembl build GRCz11 of zebrafish genome using Tophat 2.0.11 without

novel splicing form calls[61]. Transcript abundance and differential expression were calculated with

Cufflinks 2.2.1[62]. Differential expression analysis was performed using DESeq2 in R[46]. Pathway

analysis was performed using Metascape and graphed using GraphPad Prism 7 (GraphPad

Software)[63].

ATAC-seq

5,000 sorted melanocytes from 3-4 control or ID1-overexpressing fish at each stage were lysed

and subjected to "tagmentation" reaction and library construction as previously described[64].

Libraries were run on an Illumina HiSeq 4000 with paired end 150bp. All zebrafish ATAC-Seq

datasets were aligned to build version Ensembl build GRCz11of the zebrafish genome using

Bowtie2 (version 2.2.1) with the following parameters: --end-to-end, -N0, -L20[55]. We used the

MACS2 (version 2.1.0) peak finding algorithm to identify regions of ATAC-Seq peaks and derive

the normalized tracks, with the following parameter --nomodel --shift -100 --extsize 200 --

SPMR[65]. A q-value threshold of enrichment of 0.05 was used for all datasets. The motifs enriched

in ATAC-seq peaks of interest were analyzed using the findMotifsGenome program in the

HOMER package[66]. The corresponding zebrafish genome sequences are used as background

sequences in motif search. The top known HOMER motifs and de novo motifs with q-value less

than *** are calculated. HOMER motif results are presented as a heatmap of the percent of targets

containing the indicated motif minus background. For example, 41.8% of new or increased target

peaks in large *crestin+* samples contain a RUNX motif, as compared to 20.7% of background

peaks. Therefore, the heatmap indicates 21.1% of target peaks are enriched for a RUNX motif.

Transplantation

FACS   isolated   *mitfa:mCherry+/crestin:EGFP-*   cancer   precursor   zone   cells   or

mitfa:mCherry+/crestin:EGFP+ cells from small *crestin* patches or tumors were enrobed in 3 µL

Matrigel (Corning, 356234) and injected at concentrations of 5,000, 3,000, 1,000, or 500 cells

under the skin of *casper* recipient fish irradiated sub-lethally with 30 Gy split over 2 days. Isolated

cells from each donor went into 3-7 recipient fish. Fish were imaged as described above and limit

dilution calculations and curves were generated using the Extreme Limiting Dilution Analysis

software (http://bioinf.wehi.edu.au/software/elda/)[49] on data collected 14 days post-transplant.

IP-MS

The fusion (no stop) human ID1 or Clover open reading frames were cloned into the pcDNA3.2

V5 destination vector (Invitrogen, 12489019) to create V5 tagged ID1 or Clover constructs.

Constructs   were   transiently   transfected   into   A375   human   melanoma   cells   using

Lipofectamine®3000 (Invitrogen, L3000001) in 10 cm^2 plates with three independent replicates.

48 hours after transfection, cytoplasmic and nuclear fractions were isolated using the NE-PER™

Nuclear and Cytoplasmic Extraction Kit (ThermoFisher, 78833) and lysed per protocol. Anti-V5

(Clone V5-10, Sigma, V8012) was conjugated to protein G beads (ThermoFisher, 10004D). IP'd

proteins were eluted and submitted for mass spectrometry using the Taplin Mass Spectrometry

Facility at Harvard University. Proteins were included in analysis if all replicates had greater than

3 peptides.

ChIP-Seq

PMEL/hTERT/CDK4(R24C)/p53DD/BRAF(V600E) cells (courtesy of D. Fisher lab[33]) were

transiently transfected with V5-Clover or V5-ID1, as detailed above, in 15 cm^2 plates with three

independent replicates. Cells were crosslinked in 11% formaldehyde, lysed, and sheared as

described[67]. Solubilized chromatin was immunoprecipitated with 10 µg TCF12 antibody (SCBT,

sc357). Antibody-chromatin complexes were purified and libraries were prepared as described[67].

All ChIP-Seq datasets were aligned to Ensembl build version GRCh38 of the human genome using

Bowtie2 (version 2.2.1) with the following parameters: --end-to-end, -N0, -L2086[55]. MACS2

(version 2.1.0) peak finding algorithm was used to identify regions of ChIP-Seq peaks, with a q-

value threshold of enrichment of 0.05 for all datasets[65]. ChipSeeker is utilized to annotate ChIP-

seq peaks to neighboring genes according to Ensembl gene annotation[68]. The parameters are

defined as proximal promoter: 500 bp upstream – 50 bp downstream of TSS; distal promoter: 2 k

bp upstream – 500 bp downstream of TSS; enhancer: 100 k bp from TSS. The genome-wide

occupancy profile figures were generated by deeptools2 with peaks centered[69]. HOMER analysis

was performed to confirm transcription factor binding under peaks with hg38 genome utilized as

a random set of background peaks for motif enrichment[66]. To count TCF12 motifs enriched in

called peaks, mapped TCF12 motifs across the hg38 genome from ENCODE enrichment was

used[70]. Reactome pathway analysis was completed with Gene Ontology[71, 72].

Cell Viability

The fusion human TCF12 open reading frame was cloned into pcDNA3.2 V5 destination vector

to create V5 tagged TCF12. A375 or PMEL cells were seeded in opaque-walled 96-well plates.

5,000 cells were transiently transfected with V5-ID1 (A375), V5-TCF12 (PMEL) or a V5-clover

control (A375 and PMEL) for 48 hours. Cell viability was then measured using CellTiter-Glo

(Promega #G7570) and luminescence was read on Synergy Neo plate reader. Experiments were

performed in technical triplicate for each biological replicate (n=3).

Statistics

A two-way ANOVA with Tukey's multiple comparisons test was used to compare the weighted

number of cancer precursor zones or *crestin* patches and the immunofluorescence quantitation.

The calculations were performed using GraphPad Prism 7 (GraphPad Software).

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FIGURE LEGENDS
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Figure 1 | Cancer precursor zone formation precedes neural crest reactivation in zebrafish.

a, Melanoma initiation in zebrafish occurs in stages, beginning with the formation of a

cancerized field of rescued melanocytes, which expand to form a cancer precursor zone. Neural

crest reactivation occurs in cells within the cancer precursor zone and tumor formation follows,

with all tumor cells becoming *crestin:EGFP+*. Scale bar: 1 mm. **b,** Swimmers plot of 13

zebrafish during melanoma initiation show variable onset of CPZ and neural reactivation (#

represents fish time course in Fig 1a; * represents fish time course in Extended Data Fig. 1c). **c,**

Transverse sections of zebrafish skin stained for mCherry and EGFP at the cancerized field,

cancer precursor zone, neural crest reactivation, and tumor formation stages. Scale bars, 100 μm.

d, Limit dilution curves at 14 days post-transplant (dpt). Confidence interval estimate is

displayed as 1/(engraftment cell frequency).

Figure 2 | Single cell expression atlas of melanoma initiation. a, UMAP of cancerized fields

(935 cells) or small *crestin:EGFP+* patches with surrounding cancer precursor zone (1251 cells).

b, UMAP of Pseudotime analysis with cancerized field melanocytes set as the trunk. **c,**

Melanocyte and neural crest gene expression in cancerized field control melanocyte (mel),

cancer precursor zone (CPZ), and *crestin+* melanoma cell subsets. **d,** Pseudotime analysis of

genes identified as stage specific in scRNA-seq. **e**, Differentially expressed genes categorized

into signaling (red), immune (blue), and neural crest (green). **f**, Open chromatin peaks in

promoter or enhancer regions of stage-specific genes.

Figure 3 | Stage-specific transcription factors induce cancer precursor zone formation

through ID1 and neural crest reactivation via IRF3. a, Images showing *mitfa:mCherry* and

crestin:EGFP fluorescence in zebrafish overexpressing ID1, IRF3, compared to an empty

control. White dashed line outlines the cancer precursor zones or patches. Gut autofluorescence

can be seen outside of the dashed lines. **b**, Violin plot showing the weighted number of cancer

precursor zones or *crestin+* patches per fish. Each dot represents one fish; n = 35-65. ns = not

significant; **$p \leq 0.01$ **c,** Stacked bar graph representing the number of tumors in ID1 and IRF3

overexpressing zebrafish compared to empty control (n=15 fish per cohort). **d,** Kaplan-Meier

curve of melanoma incidence. ID1 or IRF3 overexpression does not accelerate melanoma onset

compared to an empty control. n.s. = not significant. **e,** Immunofluorescent staining of

melanocytes (MART1), ID1, and IRF3 in human skin samples. Scale bars, 10μm. **f,** Percent of

melanocytes that are positive for cytoplasmic and/or nuclear ID1 and nuclear IRF3. Each dot

represents one patient (n = 5-9). **$p \leq 0.01$; ****$p \leq 0.0001$.

Data Figure 4 | ID1 sequesters TCF12 from chromatin to drive cancer precursor zone

formation through loss of development and signaling regulatory genes. a, Volcano plot

analysis of IP-MS data comparing proteins bound to ID1 vs Clover pull down. **b**, Heatmap of

TCF12 targets identified by their decreased expression in the context of ID1 overexpression and

the presence of a TCF12 motif in the promoter/enhancer region. **c,** Heatmaps depicting ChIP-seq

peak structure following Clover (left, gray) or ID1 (right, purple) overexpression in

PMEL(BRAF^{V600E}) cells. A kb window is centered on the peak center highlighting peaks lost

after ID1 overexpression. **d,** Quantification of peaks with TCF12 motifs following Clover

(n=805) or ID1 (n=336) overexpression. TCF12 motif is only enriched in Clover overexpressing

cells (p = 1e-17). **e,** IGV tracks of TCF12 motif containing peaks in clover (gray) or ID1 (purple)

overexpressing cells. Purple bar indicates peak called by MACS. **f,g.** Bar graph representing

viable cells relative to non-transfected controls following Clover or ID1 overexpression in

PMEL(BRAF^{V600E}) (**f**) or TCF12 overexpression in A375 metastatic melanoma cells (**g**). ID1

overexpression increases proliferation in primary melanocytes while TCF12 decreases

proliferation in transformed malignant cells. ***, p<0.01; ****, p<0.001. **h,** Violin plot showing

the weighted number of cancer precursor zones following TCF12 overexpression relative to

empty control. Each dot represents one fish; n = 12. ns ****$p \leq 0.001$. **i,** Kaplan-Meier curve of

melanoma incidence. Melanoma incidence is delayed in TCF12 overexpressing fish that develop

CPZs. *, p<0.05.

EXTENDED DATA FIGURE LEGENDS

Extended Data Figure 1 | Cancer precursor zone formation precedes neural crest

reactivation in zebrafish. a, Melanoma initiates in a stage-specific manner, beginning with the

formation of an *mitfa^{high} crestin:EGFP-* cancer precursor zone. The neural crest progenitor state

is reactivated from within the cancer precursor zone and tumor formation occurs following this

reactivation. **b,** Melanoma initiation stages in a zebrafish with a dorsal tumor. Scale bar: 1mm**. c**

, Images of melanoma initiation for zebrafish marked * in Fig. 1a. Scale bar: 1mm. **c,** Violin plot

of CPZ and patch onset. CPZ have a median onset of 11 wpf (circle, left) while patch occur at

15wpf (square, right). Median is shown in solid black line. Each dot is a fish; n=25. **d-e,** Cancer

precursor zone encompassing *crestin* reactivation is seen in *mcr:BRAF^{V600E}* (**d**) and

mcr:NRAS^{Q61R} (**e**).

Extended Data Figure 2 | Neural crest reactivation is not dependent on anatomic position,

size, or *mitfa:mCherry* intensity. a, CPZ anatomic location across the zebrafish body. CPZ

incidence does not correlate with anatomic location. Size of the dot represents the count of CPZs

at that location, n=64. **b, c.** Violin plots of cancer precursor zone area (**b**) or mean mitfa:mCherry

intensity within that area (**c**). Size or mitfa:mCherry intensity does not predict which CPZs will

reactivate neural crest (square; mean area: 103.2±30.0; mean intensity: 0.49±0.26) compared to

CPZs that stall in the CPZ stage (circle; mean area: 0.60±0.28; mean intensity: 103.0±23.1).

Each point is a fish, n=20.

Extended Data Figure 3 | Cancer precursor zones are proliferative regions of atypical

melanocytes that precede neural crest reactivation. a, Confocal images depicting

morphological changes; cancerized field melanocytes are dendritic and regularly spaced,

whereas patches with neural crest reactivation are made of atypical non-dendritic melanocytes. **b**,

PCNA staining reveals that cancer precursor zones and patches with neural crest reactivation are

highly proliferative compared to infrequently proliferative cancerized field melanocytes. Scale

bars, 100 μm. **c**, H&E staining of cancer precursor zones show that nuclei are dysplastic. Scale

bars, 100μm.

Extended Data Figure 4 | Malignant potential of melanoma initiation stages. a, Schematic of

experiment. Precursor lesion (*mitfa:mCherry+/crestin-*), small *crestin* patch

(*mitfa:mCherry+/crestin+*), or tumor cells (*mitfa:mCherry+/crestin+*) were isolated via FACS

and subcutaneously (Sub-Q) transplanted into irradiated (IR) recipients at various densities. **b**,

Confidence interval table for 1/(engraftment cell frequency) in transplant studies. **c**, Time course

imaging of precursor lesion transplantation shows that *crestin* is reactivated upon engraftment.

Extended Data Figure 5 | Single cell transcriptome of melanoma initiation. a, Dot plot

showing the expression of cell-type specific genes used to identify clusters. The dot plot is

broken up by sample, with the control cancerized field (ctrl) in red and the *crestin* (cre) sample in

blue. **b**, UMAP visualization of cell clusters marked by sample. The control sample contains

cells isolated from a cancerized field and the *crestin* sample contains cells isolated from a small

crestin+ patch with a surrounding cancer precursor zone. **c,** UMAP of single cells (blue) with

excluded doublets (red) from the scRNA-seq. **d,** UMAP of the assigned cell identities from the

Pseudotime presented in Fig. 2b.

Extended Data Figure 6 | Differential gene expression across melanoma stages. a-c, Volcano

plots showing differentially expressed genes in CPZ (**a**), patch (**b**), and tumor (**c**) relative to

cancerized field control. Dashed lines represent a 2 $\log_2$FC (over cancerized field) and a 10e-3 q-

value as assigned by DeSeq2. Colored regions indicate genes that meet the logFC and q-value

threshold. **d**, PCA plot of cancerized field, cancer precursor zones, and patches. CPZ

melanocytes (yellow) represent a transitory population from cancerized field (red) to patch

(green and blue) and tumor (purple).

Extended Data Figure 7 | Stage-specific gene expression during melanoma initiation in

zebrafish and humans. a, Gene expression of stage-specific differentially expressed genes from

FACS-sorted melanocytes during melanoma initiation. **b**, Stage-specific expression signatures in

intermediate lesions, thin melanoma, and thick melanoma samples taken from human patients

from Shain et al. Cancer Cell, 2018. **c**, Stage-specific gene expression in human melanoma

samples from microarray analysis performed in Smith et al. Cancer Biol Ther 2005. Samples

were taken from normal skin (NS), benign nevi (BN), atypical nevi (AN), melanoma in situ

(MIS), vertical growth phase (VGP), metastatic growth phase (MGP), and lymph node

metastasis (LN).

Extended Data Figure 8 | Differential accessible chromatin regions across melanoma stages.

a-c, Volcano plots showing differentially accessible regions in CPZ (**a**), patch (**b**), and tumor (**c**)

relative to cancerized field control. Dashed lines represent a 2 $\log_2$FC (over cancerized field) and

a 10e-3 q-value as assigned by DeSeq2. Colored regions indicate peaks that meet the logFC and

q-value threshold.

Extended Data Figure 9 | Enriched binding motifs during melanoma initiation. a-g, ATAC-

seq peak tracks showing open chromatin in enhancers and/or promoters for the cancer precursor

zone genes *id1* (**a**), *il6st* (**b**), and *tfap2a* (**c**), as well as the *crestin* patch genes *ctnnb2* (**d**), *irf3* (**e**),

sox10 (**f**), and *crestin* (**g**). Black box highlights the peak that is displayed in Figure 2d.

Extended Data Figure 10 | MAPK pathway is not transcriptionally active in cancerized

field melanocytes. a, MAPK pathway gene expression. CPZ = cancer precursor zone. **b**, ATAC-

seq peak tracks demonstrating that chromatin around MAPK pathway genes does not open until

the formation of a cancer precursor zone. **c**, mCherry and phospho-ERK in cancerized field

(left), early *mitfa* $^{\text{med}}$ cancer precursor zone (middle), and *mitfa* $^{\text{high}}$ cancer precursor zone (right).

Scale bars, 100 μm. **d**, Expression levels of BRAF$^{\text{V600E}}$ are unchanged between FACS-isolated

cancerized field (CF) and cancer precursor zone (CPZ) melanocytes, n=3-8. ****$p \leq 0.0001$.

Extended Data Figure 11 | ID1 involved in early-stage melanoma initiation and IRF3

involved in late-stage melanoma initiation. a,b. RNAScope in situ hybridization for id1 (red)

and DAPI nuclear stain in fixed and sectioned cancerized field (**a**) or CPZ (**b**) melanocytes.

Dashed line outlines the skin surface. Zebrafish id1 is not expressed in cancerized field

melanocytes but is expressed abundantly in CPZ melanocytes. **c**, Heatmap of IRF3 pathway

members increasing expression at later stages of initiation. Cancerized field (CF), cancer

precursor zone (CPZ), as well as small (S), medium (M), and large (L) *crestin* cells were

compared. **d**, Mean intensity of ID1in individual melanocytes in normal skin (N), atypical

melanocyte zones (AML), melanoma in situ (MIS), and primary melanoma (PM) samples,

n=109-132 melanocytes across 5-9 patient samples. $**p \leq 0.01$; $***p \leq 0.001$; $****p \leq 0.0001$.

Extended Data Figure 12 | ID1 inhibits TCF12, preventing binding to downstream targets

and enabling cancer precursor zone formation. a, Heatmap of the Log2FC of ID1-bound

proteins over control. **b,** Expression of known TCF12 targets identified by the TRANSFAC

Curated Transcription Factor Targets dataset. **c**, Expression of TCF12 targets decrease in cancer

precursor zone and further decrease when ID1 is overexpressed. **d,** Increasing ID1 expression

decreases open chromatin peaks in promoter or enhancer regions of *rhogb* and *parvg*. **e,** Dot plot

of enriched GO terms in genes with annotated TCF12 motif lost following ID1 overexpression.

Figure 1

Figure 2

Figure 3

Figure 4