BMP signaling: at the gate between activated melanocyte stem cells and differentiation

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Through recurrent bouts synchronous with the hair cycle, quiescent melanocyte stem cells (McSCs) become activated to generate proliferative progeny that differentiate into pigment-producing melanocytes. The signaling factors orchestrating these events remain incompletely understood. Here, we use single-cell RNA sequencing with comparative gene expression analysis to elucidate the transcriptional dynamics of McSCs through quiescence, activation, and melanocyte maturation. Unearthing converging signs of increased WNT and BMP signaling along this progression, we endeavored to understand how these pathways are integrated. Employing conditional lineage-specific genetic ablation studies in mice, we found that loss of BMP signaling in the lineage leads to hair graying due to a block in melanocyte maturation. We show that interestingly, BMP signaling functions downstream from activated McSCs and maintains WNT effector, transcription factor LEF1. Employing pseudotime analysis, genetics, and chromatin landscaping, we show that following WNT-mediated activation of McSCs, BMP and WNT pathways collaborate to trigger the commitment of proliferative progeny by fueling LEF1- and MITF-dependent differentiation. Our findings shed light upon the signaling interplay and timing of cues that orchestrate melanocyte lineage progression in the hair follicle and underscore a key role for BMP signaling in driving complete differentiation.

Keywords: BMP signaling; LEF1; MITF; melanocyte stem cells; pigment production; proliferative progeny; quiescence; stem cell activation

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while a related family member, TGF-β, controls McSC quiescence [Nishimura et al. 2010]. At anagen onset, WNT/β-catenin signaling stimulates the activation of both HFSCs and McSCs at the bulge base [hair germ [HG]] [Rabbani et al. 2011]. An additional twist comes from cross-communication, where WNT-activated HFSCs produce endothelins that further stimulate McSC expansion [Rabbani et al. 2011; Chang et al. 2013; Takeo et al. 2016].

Once activated, HFSCs within the HG generate short-lived proliferative progeny that envelope the dermal papilla (DP) to form a hair bulb from which the hair emerges. In contrast, HFSCs within the bulge generate the downward growing outer root sheath (ORS), which then distances the bulge from the proliferative hair bulb and its DP-signaling center, thereby restoring quiescence to the bulge and upper ORS [Hsu et al. 2014; Yang et al. 2017]. The McSC lineage seems to follow this trend [Fig. 1A]. As McSCs near the HG become activated [aMcSCs], they give rise to committed, proliferative progeny [McCPs] that coexist with the proliferative HFSC progeny in the hair bulb. There, they mature into melanin-producing melanocytes concomitantly with the differentiating hair cells at the heart of the hair bulb. Melanocyte activity continues until catagen ensues, when the differentiated McSC progeny undergo apoptosis alongside HFSC progeny.

Functional studies in mice have provided valuable insights into the final maturation steps in melanocyte differentiation, which require both WNT/β-catenin [Rabbani et al. 2011] and SCF/c-Kit signaling [Botchkareva et al. 2001; Liao et al. 2017]. However, our temporal knowledge of the signals governing the intermediate steps between McSCs and mature melanocytes remains limited, as does an understanding of the global changes in gene expression that occur throughout McSC lineage progression. Thus far, our transcriptional insights rest on single-cell qPCR or bulk RNA sequencing of McSCs [Osawa et al. 2005; Moon et al. 2017; Joshi et al. 2019; Zhang et al. 2020] or single-cell comparisons between McSCs and melanoma [Sun et al. 2019].

Understanding how the melanocyte lineage is orchestrated to yield normal hair pigmentation is a fundamental prerequisite to elucidating how these mechanisms are
disrupted in age-related and premature hair graying, as well as in melanocyte-related diseases such as vitiligo (Mort et al. 2015). It is also at the crux of understanding how melanoma cells can exhibit features of both stemness and differentiation as they progress through malignancy.

Here, we endeavored to bridge this gap by purifying McSCs as they transition through their discrete stages of the lineage and elucidating their transcriptional features. We then harnessed bioinformatics, genetics, and molecular approaches to dissect the signals that orchestrate these transitional steps. In doing so, we unearth and elucidate a surprising role of BMP signaling downstream from McSCs and at the intersection between early McCPs and mature melanocytes. We further uncover evidence that BMP collaborates with WNT signaling to promote full differentiation.

Results

Identifying and isolating melanocyte lineage populations throughout the hair cycle

To label all melanocyte lineage populations throughout the hair cycle, we used dopachrome tautomerase-eGFP (Dct-eGFP) mice (Gong et al. 2003). We confirmed that eGFP+ patterns overlapped with endogenous DCT, expressed in both the McSCs in the bulge and differentiating progeny in the hair bulb [Fig. 1B]. To distinguish the committed melanocyte populations in the hair bulb from the McSCs in the bulge region, we mated Dct-eGFP and Lef1-RFP transgenic mice [Rendl et al. 2005]. Within the hair bulb, Lef1-RFP was active in both melanocytes and the DP, but only melanocytes coexpressed Dct-eGFP and Lef1-RFP [Fig. 1B]. To determine the timing of McSC activation (proliferation), we administered 24 h pulses of 5-ethynyl-2-deoxyuridine (EdU) from late telogen through anagen (Supplemental Fig. S1A). As assessed by immunofluorescence (IFM) microscopy and quantification, McSC proliferative activity peaked at P21 (anagen I-II) (Supplemental Fig. S1B, C). At this stage, HG expansion and DP engulfment was minimal (Supplemental Fig. S1D). By P23 (anagen III), HGs had developed into hair bulbs, which became distanced from the bulge, allowing us to evaluate the McCP compartment. Of total DCT+ P23 hair bulb progeny in the hair bulb, namely McCPs and melanocytes.

We next devised fluorescence-activated cell sorting (FACS) strategies to isolate qMcSCs, aMcSCs, and differentiating McSC progeny [Fig. 1C; Supplemental Fig. S2A]. McSCs are quiescent throughout second telogen, and because hair bulbs are not present at this stage, it was possible to enrich for McSCs by first excluding endothelial [CD31+], immune [CD45+], fibroblast, platelet, and adipocyte [CD140a+], epidermal/HF infundibulum [Sca-1+] and bulge HFSC [CD34+] populations, and then selecting for eGFP+ integrin α6medium-low cells. By shifting to anagen I-II, we could enrich for aMcSCs with a similar strategy. Finally, by harvesting cells at (full) anagen VI and using additional markers RFP and c-Kit [CD117], we could enrich for McSC progeny in the hair bulb, namely McCPs and melanocytes.

Single-cell transcriptional profiling dissects the melanocyte lineage into six stages

For initial validation, we used bulk RNA sequencing to profile our qMcSC and differentiating hair bulb McSC progeny populations and found robust expression of lineage markers Dct, Sox10, and c-Kit, relative to HFSCs [Supplemental Fig. S2B; Supplemental Table S1]. Differential gene expression analysis with DESeq2 [Love et al. 2014] revealed major transcriptional differences between qMcSCs and differentiating McSC progeny (2385 genes up in qMcSCs; 2819 up in progeny, Padj ≤ 0.001, log2fold ≥ 1) [Supplemental Fig. S2C, Supplemental Table S1]. While our independent replicates showed strong correlations and reproducibility [Supplemental Fig. S2D], we anticipated heterogeneity from the presence of intermediate melanocyte lineage steps. We therefore performed single-cell RNA sequencing [scRNA-seq] on each population. Single-cell mRNAs were subjected to SmartSeq2, which provides higher depth of coverage than Drop-seq or 10X Genomics [Picelli et al. 2013]. Dctlow and Krt15+ cells were determined to be nonmelanocyte lineage cells and were eliminated from further consideration, as were cells displaying transcript reads from <1250 genes (see the Materials and Methods). In all cases the retained cells had an average of 1 × 105 reads/cell, with few reads from the full data sets mapping to the mitochondrial genome, reflecting the high quality of our data [Supplemental Fig. S3A, B]. Telogen-phase qMcSCs from independent replicates exhibited strongly overlapping profiles [Supplemental Fig. S3C, middle], reflecting minimal batch effects. Even for anagen, where slight variations in hair cycle stage can impact lineage profiles, hair bulb McSC progeny from independent mice were still more closely related to each other than to any other population. qMcSCs displayed the least complex transcriptome [Supplemental Fig. S3C, right], consistent with the findings of Osawa et al. [2005].

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Principal component analysis (PCA) with uniform manifold approximation and projection (UMAP) [Becht et al. 2019] coupled with k-nearest neighbors (k-NN) based community detection [Blondel et al. 2008] showed that our three FACS populations segregated into three distinct clusters, within which cells exhibited robust expression of Dct and Mitf, encoding the master melanocyte transcription factor [Fig. 1D; Goding and Arnheiter 2019]. Telogen and anagen I-II McSC populations [C1 and C2] shared significantly more similarities than to the differentiating McSC progeny hair bulb population [C3]. To understand what major transcriptional features define “stemness” versus “differentiation,” we combined clusters C1 and C2 and performed differential gene expression analysis against C3. Selected differentially expressed genes are shown in the heat map in Figure 1E (Materials and
Methods; Supplemental Table S2). Based on global transcriptional differences, we then assigned each cell a “stemness” and “differentiation” score as a measure of their stage in lineage progression [Materials and Methods, Supplemental Fig. S4A]. Cells ubiquitously expressed certain pan-lineage genes, including Sox10 (Supplemental Fig. S4B). However, C1 and C2 showed selective up-regulation of genes including Pax3, Sbno2, and Bcl2 [Nishimura et al. 2005], while C3 cells exhibited elevated transcripts involved in pigment production, including melanin production enzymes [Tyrp1 and Tyr] and melanosome-related proteins [i.e., Mlana and Oca2]. C3 was also enriched for transcripts involved in well-known melanocyte differentiation pathways, including c-Kit/SCF [Kitl] (Botchkareva et al. 2001; Liao et al. 2017) and WNT signaling [Lef1] (Fig. 1E; Rabbani et al. 2011).

We also detected transcriptional heterogeneity within each major cluster, which increased as cells passed from quiescence through activation to differentiation (Fig. 2A). Six distinct subpopulations were identified, three of which [C1, C2a, and C2b] showed positive scores for “stemness” (Fig. 2A,B). Cells within subclusters were further distinguished by their differential expression of proliferation genes, which we used to similarly generate a global score (Fig. 2B,C; Supplemental Fig. S4C,D). From these comparisons, cluster C2a was less quiescent than C1, while clusters C2b and C3a were particularly enriched for cell cycling transcripts [i.e., Mki67 and Ccna2]. Based upon their stemness, proliferation, and differentiation scores, as well as the hair cycle stage at which they were isolated, we assigned these four clusters as qMcSCs [C1], aMcSCs [C2a], a transitional state from aMcSCs to early McCPs [C2b], hair bulb McCPs [C3a], and differentiated melanocytes [C3b,c].

Although some cell cycle transcripts remained elevated in the more differentiated C3b and C3c subclusters, overall transcripts for key cell cycle genes were substantially lower than in the C3a subcluster. In contrast, C3c expressed the highest level of melanogenesis transcripts, marking this subcluster as having the most mature melanocytes (Fig. 2C,D). Overall, the six discrete melanocyte lineage cell subclusters could be demarcated into proliferative and nonproliferative populations, and then further delineated according to melanogenesis and melanosome gene expression patterns. Differential gene expression analysis and gene ontology (GO) analysis [PANTHER] (Thomas et al. 2006; Mi et al. 2019) underscored these distinctive features (Fig. 2E; Supplemental Tables S3–S6). Notably, many expected terms were enriched for each population, for example, Notch [Moriyama et al. 2006] and TGF-βR signaling [Nishimura et al. 2010] for qMcSCs in the C1 subcluster.

Uncovering signaling pathway integration during melanocyte lineage progression

We next turned our attention to signaling pathways and genes likely to be involved in lineage commitment and progression, focusing our attention on cohorts of genes that robustly change expression during transitional stages where less is understood. Expression of genes annotated under select subcluster GO terms (Fig. 2E) are illustrated in the heat map in Figure 3A [see also Supplemental Table S7].

WNT-signaling is known to promote McSC proliferation and drive their differentiation into pigment-producing melanocytes [Rabbani et al. 2011]. Our ability to subdivide the McSC lineage into six hierarchical subclusters now further revealed that genes implicated in WNT signaling and response were particularly elevated at the juncture between the proliferative anagen I-II C2b cells and the proliferative McCP C3a cells in the anagen VI hair bulb [Fig. 3A,B; Supplemental Fig. S5A; Supplemental Table S8]. This marked transition was further reflected by the significantly enriched [FDR = 1.7 × 10−4] “canonical WNT signaling pathway” in subcluster C3a, as compared with C3c (Fig. 2E).

Of further intrigue was a dramatic elevation of bone morphogenetic protein (BMP) pathway response genes occurring at the same juncture [Fig. 3A,B; Supplemental Fig. S5B; Supplemental Table S8]. Analogously, the “positive regulation of BMP signaling pathway” GO term was enriched in the C3a McCPs [FDR = 1.5 × 10−6] (Fig. 2E). BMPs have been reported to elicit ligand and context-specific effects on melanin production in cultured cells [Bilodeau et al. 2001; Jin et al. 2001; Yaar et al. 2006; Kawakami et al. 2008; Park et al. 2009; Singh et al. 2012; Yang et al. 2014]. However, their role in the native tissue microenvironment, as well as their cooperation with other niche signaling components throughout differentiation, is not well understood. We therefore sought deeper understanding of the role of BMP signaling in melanocyte lineage progression, while also exploring possible interplay with WNT.

Probing deeper into WNT pathway activation throughout the lineage, we found that in addition to the elevated levels of genes such as Lef1, Wif1, Me1, and Wnts5a [Wnts5a expression was consistent with Laron et al. 2018] within the McCP/melanocyte populations [Fig. 3C], nuclear LEF1, an established proxy for active WNT signaling, correlated with these patterns, showing strongest labeling within differentiating progeny in the hair bulb [Fig. 3D]. Similarly, we found robust up-regulation of BMP target genes Id1/2 and Mxs1/2 and ligands BMP2/7 in McCPs/melanocytes [Fig. 3E], consistent with their BMP receptor expression [Supplemental Fig. S5C].

BMP receptor signaling triggers the phosphorylation and activation of SMAD1/5/9 proteins, which form a complex with SMAD4 that translocates to the nucleus to regulate gene expression [Wang et al. 2014]. While the bulge niche is rich in BMPs that signal to pSMAD1/5/9+ HFSCs [Kobielak et al. 2007; Genader et al. 2014], surprisingly, neither quiescent nor activated McSCs displayed nuclear pSMAD1/5/9. In contrast, McCPs in anagen III HF showed nuclear pSMAD1/5/9+, and this was maintained as McCPs/melanocytes continued to differentiate [Fig. 3F]. Taken together, our comparative transcriptional analyses and IMF data were consistent with the hypothesis that BMP signaling might be acting downstream to promote differentiation, and that it might act in concert with WNT signaling to do so.

1716 GENES & DEVELOPMENT
To probe deeper into how WNT and BMP signaling might intersect, we employed loss of function mouse models for each pathway. We first demonstrated LEF1's importance in melanocyte differentiation in the HF by showing that the remaining HF s in Lef1-null mice (C57/BL6J) (van Genderen et al. 1994; Kratochwil et al. 1996) had DCT⁺ cells in the hair bulb, but that hairs were less pigmented (Fig. 4A; Supplemental Fig. S6A). Using TyrCreER Ctnnb1[lox/lox] R26-stop-fl-FP mice (Braut et al. 2001; Srinivas et al. 2001; Bosenberg et al. 2006) and the strategy illustrated in Figure 4B, we then conditionally ablated LEF1’s WNT signaling cofactor, β-catenin (Ctnnb1) in the melanocyte lineage to recapitulate a model similar to that used by Rabbani et al. [2011] to demonstrate a role…
for WNT in McSC proliferation and differentiation. As expected, after one hair cycle, second telogen Ctnnb1 cKO mice had reduced pigmentation relative to control (Fig. 4C). Having substantiated and extended the findings of prior studies (van Genderen et al. 1994; Kratochwil et al. 1996; Rabbani et al. 2011), we then turned to assessing the status of BMP signaling in the absence of WNT signaling. For this purpose, we waxed Ctnnb1 cKO mice to induce hair growth and then analyzed tissue 1 wk later when HFs were in anagen (Fig. 4B). Interestingly, Ctnnb1-null hair bulb McSC progeny still showed nuclear pSMAD1/5/9, indicating that BMP

Figure 3. WNT and BMP signaling increase during melanocyte differentiation. (A) Heat map of gene expression for select genes from GO terms in Figure 2E, including WNT and BMP signaling pathways. See also Supplemental Table S7. (B) Violin plots of canonical WNT and BMP target gene pathway scores in each subcluster. See also Supplemental Table S8. (C) Expression plots of WNT pathway genes. (D) IMF images showing LEF1 protein expression in the melanocyte lineage throughout hair cycling. Nuclear LEF1 (red) indicated by yellow arrows. (E) Expression plots of BMP pathway genes. (F) IMF images of melanocyte lineage populations in HFs throughout the hair cycle. pSMAD1/5/9 is absent in McSCs but becomes nuclear in McCPs/melanocytes (Mcs), indicated by yellow arrows. All scale bars, 25 μm.
signaling is still active in the absence of functional WNT signaling (Fig. 4D).

Given this result, we then proceeded to functionally interrogate the role of BMP signaling in melanocyte lineage progression by performing conditional melanocyte lineage-specific ablation of Bmpr1a (Alk3) to extinguish downstream signaling through this pathway [Mishina et al. 2002]. At P21, TyrCreER Bmpr1a^{fl/flox} R26fl-stop-YFP mice were treated with tamoxifen, and anagen skins were analyzed by IMF 1 wk after waxing (Fig. 4B).
Strikingly, by second telogen, Bmpr1a cKO mice exhibited gray hair, whereas mice expressing at least one Bmpr1a allele (TyrCreER<sup>+</sup>/<sup>-</sup>, TyrCreER<sup>+</sup> Bmpr1a<sup>−/−</sup> or TyrCreER<sup>+</sup> Bmpr1a<sup>−/+</sup>) showed no difference in coat color (Fig. 4E). By optical density, we found significantly reduced pigmentation in Bmpr1a cKO hair bulbs [Fig. 4F].

Adult hair graying phenotypes can be caused by defects in the McSCs themselves or at any point during differentiation [Nishimura 2011]. To test whether McSC integrity was compromised by loss of Bmpr1a, we performed waxing at second telogen to induce hair cycle entry and McSC activation, and then measured the number of McSCs 1 wk later. We found that McSC numbers were comparable when BMP signaling was abrogated [Fig. 4G], and following activation, Bmpr1a-null McSCs returned to quiescence normally, as measured by proliferation during a 4-h EdU pulse (Fig. 4H).

To further challenge the stem cell compartment upon loss of BMPR1a, we performed serial waxing, depleting second telogen mice a total of five times every 3 wk. Even after this heightened demand on McSC activity, waxing did not exacerbate the hair graying phenotype (Supplemental Fig. S6B). Indeed, upon analyzing anagen skin 1 wk after the fifth round of waxing, we found no significant difference in number of McSCs between control and Bmpr1a cKO mice (Fig. 4I).

Despite efficient Bmpr1a targeting and loss of both canonical [pSMAD1/5/9] and noncanonical [pP38] BMP signaling, DCT<sup>+</sup> McSC progeny were still present in hair bulbs of Bmpr1a cKO mice (Fig. 4J; Supplemental Fig. S6C,D). However, hair bulbs remained poorly pigmented 12 d into anagen, indicating that the defect was not merely due to delayed maturation [Supplemental Fig. S6E]. Taken together, these findings suggested that McSCs were intact and could trigger melanocyte lineage progression in the absence of BMP signaling, but that melanin production in the lineage was impaired. This is in striking contrast to WNT signaling, which impacted both McSC and hair bulb compartments during anagen [Rabbani et al. 2011].

Given these results, and the fact that most prior studies into the roles of BMP signaling had been performed with cultured cells, we initially turned to CRISPR/Cas9 targeting of Bmpr1a in the mouse melanocyte cell line, Melan-a2 [Sviderskaya et al. 1995]. Following sequence confirmation of successful targeting, however, we realized that BMPR1a expression could also be knocked down by lentiviral delivery of a short hairpin RNA targeting in vivo resulted in loss of nuclear pSMAD1/5/9 (Supplemental Fig. S6F). While Bmpr1a targeting in vivo resulted in loss of nuclear pSMAD1/5/9, there are other BMP/TGF-β receptor superfamily members that if coexpressed can heterodimerize in multiple ways and yield responses that differ depending upon ligand, receptor combinations, and intra- and extracellular regulator molecules [for review, see Wang et al. 2014]. While this complexity precluded our ability to interrogate the functional relevance of BMP signaling in vitro, it may begin to explain the context-dependent behaviors previously described of the BMP pathway in the melanocyte lineage [Bilodeau et al. 2001; Jin et al. 2001; Sharov et al. 2005; Yaar et al. 2006; Kawakami et al. 2008; Park et al. 2009; Han et al. 2012; Singh et al. 2012; Yang et al. 2014].

Returning to studying the importance of pSMAD1/5/9 signaling in vivo, we administered tamoxifen to Bmpr1a cKO and stage-matched control mice at P21, allowed them to undergo a full round of hair cycling, and then waxed to ensure synchronous induction of anagen (Fig. 4B). To capture hair bulb McSC progeny at the point when BMP signaling had become activated and differentiation had begun, we harvested skin 1 wk later when HF s were in anagen IV [Supplemental Fig. S6G]. YFP<sup>+</sup> cells were then isolated by FACS (gated against CD31, CD45, CD140a, Sca-1 and CD34) [Supplemental Fig. S7A] and subjected to scRNA-seq.

Similar to our temporal data sets, cells from Bmpr1a cKO and control mice showed a similar depth of sequencing, and independent replicates exhibited similar clustering [Supplemental Fig. S7B]. Approximately 5000 genes per cell were detected, with few reads from mitochondrial transcripts [Supplemental Fig. S7C]. As before, Dec1<sub>low</sub> and Krt15<sup>+</sup> cells were excluded from analysis. Unsupervised clustering revealed that YFP<sup>+</sup> Bmpr1a heterozygous McSC progeny showed transcriptomes comparable with YFP<sup>+</sup> wild-type, but Bmpr1a-null cells formed a unique cohort (Fig. 5A; Supplemental Fig. S7B).

Differential gene expression analysis revealed marked changes caused by Bmpr1a ablation in the melanocyte lineage [Fig. 5A,B]. As expected, mRNAs encoded by direct BMP target genes, such as Id2 and Id3, were dramatically down-regulated [Fig. 5B,C]. Melanogenesis transcripts such as Oca2, Rab38, and Gpr143 were also markedly diminished. Conversely, transcripts up-regulated in Bmpr1a-null cells included genes that have been implicated in stemness and/or migration; e.g., Vim, Cspg4, Adgrg1, and Enpp2 [Fig. 5A,B; Supplemental Fig. S7D]. To test whether these trends persisted longer into anagen, we FACS-purified YFP<sup>+</sup> cells 12 d postwaxing and performed RT-qPCR, with which we obtained consistent results [Fig. 5D]. Altogether, 80 genes were significantly [q ≥ 0.05] down-regulated and 346 genes were up-regulated upon Bmpr1a ablation [Supplemental Table S9].

To identify the juncture along the lineage at which loss of BMP signaling exerts its effects, we integrated our Bmpr1a-null and Bmpr1a control data sets into our six subclusters of wild-type melanocyte lineage cells and again performed UMAP-based dimensionality reduction with community-based clustering [Fig. 5E]. This analysis suggested that Bmpr1a-null cells were most similar to the early McCP cluster C2b, with approximately one-third of null cells found in the same community cluster [Fig. 5E,F]. We then implemented pseudotime analysis using Monocle3 [Cao et al. 2019], an unbiased method to map differentiation in which all biologically significant genes are considered. This algorithm ordered each cell along the McSC lineage trajectory according to its individual transcriptome. In concordance with our collection strategies and differential gene expression analyses between subclusters, C1 qMcSCs were placed at the start of the inferred trajectory, followed by aMcSCs, then McCPs/melanocytes, with the highly melanogenic cluster.
BMP signaling: between McSCs and differentiation

C3c forming the termination point. Bmpr1a-null cells fell between the anagen I-II C2b early McCP cells and the stage-matched control cells and the anagen VI hair bulb cells, including C3a McCPs (Fig. 5E,F; Supplemental S7E).

Considering how the Bmpr1a-null cells compared with the six wild-type subclusters, we found that in contrast to McCP subcluster C3a, which highlighted melanin biosynthesis as well as WNT and BMP signaling, the Bmpr1a-null cluster displayed an enhanced stemness and a diminished differentiation [pigmentation] signature, with reduced expression of genes involved in WNT and BMP signaling (Fig. 5G). These differences were not merely attributable to differences in anagen stage, as consider-ably more stemness transcripts were sustained while fewer differentiation genes were elevated in the Bmpr1a-null versus stage-matched control samples. This unique-ness of signature was further reinforced by implementing the "stemness" and "differentiation" scoring parameters established for our wild-type McSC and differentiating McSC progeny populations (Fig. 5H). Overall, the Bmpr1a-null hair bulb McSC progeny most closely associated with subcluster C2b [Fig. 5E,F; Supplemental Table S10]. Thus, although the level of pigmentation genes was greater in Bmpr1a-null versus wild-type C2b cells, these data indicated that without BMP signaling, the melanocyte lineage was arrested just downstream from the activated McSCs and early McCPs.
In agreement with their similarities to activated McSCs, DCT+ Bmpr1a-null McCPs in the hair bulb displayed higher EdU incorporation relative to their control counterparts (Fig. 6A). Moreover, overall numbers of DCT+ cells were increased in Bmpr1a cKO hair bulbs, indicating that the reduced pigmentation seen in these cells was not attributable to McCP depletion (Fig. 6B). In this regard, loss of BMP signaling stood in striking contrast to loss of WNT/β-catenin signaling in the melanocyte lineage, where reduced proliferation in the hair bulb was reported (Rabbani et al. 2011).

Given the increased number of McSC progeny within Bmpr1a cKO hair bulbs, we probed deeper into the underlying defect in hair pigmentation. Confirming our transcriptomic data, early melanogenic enzymes in the lineage, including DCT, tyrosinase (TYR), and tyrosinase-related protein 1 (TYRP1, reduced at the transcript level), were still present in Bmpr1a-null cells (Fig. 6A,C). Interestingly, however, even though PMEL, a structural component of melanosomes, was expressed, it exhibited an abnormal perinuclear localization in Bmpr1a-null cells, suggestive of a defect in melanosome maturation (Fig. 6D). This notion was further strengthened by the reduced expression of melanosome organization genes such as Oca2, Rab38, and Gpr143, and with diminished expression of Mc1r, which encodes a cell surface receptor.
that instructs melanocytes to make black/brown or yellow pigment, depending on ligand binding [Fig. 5B; Supplemental Table S9].

Melanosomes undergo four distinct maturation phases that are well-defined at the ultrastructural level [Marks and Seabra 2001]. Therefore, we analyzed melanosomes by electron microscopy [EM] in Bmpr1a-null and control cells. Area densities of total and mature melanosomes were similar between conditions. However, Bmpr1a-null cells in the hair bulb displayed a significantly higher density of immature melanosomes [Fig. 6E].

**Molecular underpinnings of the block in differentiation caused by loss of BMP signaling**

Our collective findings pointed to a newfound physiologic role of BMPR1a signaling in melanosome biogenesis and maturation and traced its loss to a block downstream from McSC activation/early McCPs and upstream of melanocyte differentiation [Fig. 7A]. Seeking insights into how loss of BMP signaling arrests lineage progression, we were intrigued to see a trending down-regulation of Lef1 transcripts in Bmpr1a-null cells, which we confirmed by IF [Fig. 7B]. In cultured melanocytes, stimulation with BMP2 and BMP4, ligands in anagen skin, confirmed by IMF (Fig. 7B). In cultured melanocytes, stimulation with BMP2 and BMP4, ligands in anagen skin, induced up-regulation of Lef1 mRNA and LEF1; this response was abrogated upon treatment with BMP receptor (ALK) inhibitor [Supplemental Fig. S8A,B].

Pursuing further insights into the molecular underpinnings of BMP’s action on the melanocyte lineage, we turned toward the master regulator of the melanocyte lineage, microphthalmia-associated transcription factor [MITF]. MITF drives a myriad of lineage processes including survival, proliferation, and differentiation [Goding and Arnheiter 2019]. According to the MITF rheostat model in melanoma, low MITF promotes stemness, intermediate MITF induces proliferation, and high MITF drives differentiation/melanogenesis [Carreira et al. 2006; Goding and Arnheiter 2019]. Interestingly, nuclear MITF was markedly diminished in the absence of BMP signaling [Fig. 7C], although not at the mRNA level [Supplemental Fig. S8C]. Moreover, we also found diminished nuclear MITF in Lef1- and Ctnnb1-null hair bulb McSC progeny [Fig. 7D,E]. In cultured melanocytes, we did not observe changes in MITF upon BMP stimulation, even though LEF1 was up-regulated [Supplemental Fig. S8B–D], further underscoring the importance of studying the role of BMP signaling in a physiologic context.

Given the numerous differentiation and pigmentation defects that converge on perturbations of MITF [Goding and Arnheiter 2019], our finding that nuclear MITF was diminished upon loss of BMP signaling in vivo could largely explain the block in melanocyte differentiation and the gray hair phenotype in our in Bmpr1a cKO mice. Moreover, LEF1 is known to interact with and regulate MITF [Dorsky et al. 2000; Takeda et al. 2000; Saito et al. 2002, Yasumoto et al. 2002], a feature intriguing in that LEF1 was also diminished upon loss of BMP signaling.

**Evidence for cooperation between BMP-regulated LEF1 and MITF in melanocyte maturation**

Our collective in vivo findings thus far led us to posit that BMP signaling might function to achieve the levels of nuclear MITF and LEF1 required to promote transcription of melanocyte maturation genes. Searching for indications of such interplay, we scanned for transcription factor motifs within the putative promoter regions of genes that were differentially expressed between control and Bmpr1a-null cells. We binned genes as either containing a MITF motif only, a LEF1 motif only, or both MITF and LEF1 motifs. In a second analyses, we binned these genes as containing either a MITF motif only, a SMAD motif only, or both SMAD and MITF motifs.

Of these comparisons, one had clear statistical significance: Putative promoters harboring both MITF and LEF1 sites exhibited a greater reduction in gene expression than MITF sites alone when BMP signaling was abrogated [Fig. 7F]. These effects were not seen with genes harboring MITF and SMAD sites nor with genes whose expression increased in Bmpr1a-null cells [Fig 7F,G]. Notably, the down-regulated genes whose promoters contained both LEF1 and MITF motifs encoded proteins specifically involved in transport, transmembrane transport, and melanosome organization, including Mc1r, Oca2, Trpm1, Rab38, and Shroom2 [Supplemental Table S11].

Given these results, we wondered whether further evidence for the cooperation of these factors could be found at the chromatin level. We hypothesized that chromatin openness would change as McSCs committed to melanocyte differentiation, and that sites that gain accessibility in differentiation would show transcription factor binding motifs critical for this process. To test this possibility, we performed assay for transposase-accessible chromatin with high-throughput sequencing (ATAC-seq) [Buenrostro et al. 2013] on FACS-purified qMcSCs in the bulge and on differentiating McCP/melanocyte progeny in the anagen hair bulb. In this way, we could evaluate and compare chromatin openness associated with stem and differentiating states.

Analyses of our ATAC-seq showed high reproducibility across biological replicates [Supplemental Fig. S8E]. Similar percentages of ATAC peaks were within 50 kb of canonical transcriptional start sites [TSSs] [28.2% for peaks increased by log2 fold change ≥ 1 in qMcSCs; 26.4% for McSC progeny], and peaks were similarly distributed throughout promoters, gene bodies, and intergenic regions [Fig. 8A]. Differential peak analysis indicated that qMcSCs and their differentiating progeny exhibit strikingly distinct chromatin accessibility profiles [log2 fold change ≥ 1, P ≤ 0.05] [Fig. 8B, Supplemental Table S12].

Interestingly, a number of these lineage sensitive peaks appeared to be responsive to the status of BMP signaling, as peaks associated with differentiation resided within the promoter/enhancer regions of genes down-regulated upon Bmpr1a ablation [Fig. 8C,D]. Notably, Lef1 showed chromatin remodeling as McSCs became activated and progressed along the lineage. Moreover, many of
the ATAC peaks within the *Lef1* locus that were more open in differentiating progeny contained SMAD motifs (Fig. 8E; Supplemental Fig. S8F). These analyses offered new insights into why *Lef1* transcripts were downregulated upon loss of BMP signaling and provided support for direct transcriptional regulation of *Lef1* by BMP signaling.

To ask what transcription factors may bind these differentially accessible regions, we performed an unbiased motif analysis. Strikingly, LEF/TCF and MITF family factors were the top most enriched factors in peaks that increased in differentiation (Fig. 8F; full list in Supplemental Table S13). Conversely, peaks increased in qMcSCs showed enrichment of SP/KLF and AP1 family motifs, as well as...
Figure 8. Chromatin accessibility is dynamic during lineage progression and reveals cooperation between MITF, LEF1, and SMAD. (A) Location of ATAC peaks expressed as percentages. (B) Heat map of differential peaks in qMcSCs versus their differentiating hair bulb progeny (log2 fold change ≥ 1, \(P \leq 0.05\)). See also Supplemental Table S12. (C) Number of peaks increased in either qMcSCs or their differentiating progeny within ±5 or 50 kb of the TSS of genes whose expression is diminished upon loss of BMPR1a signaling by scRNA-seq. (D) ATAC tracks showing differential peaks of select differentiation genes. (E) ATAC track of Lef1 gene showing chromatin opening in differentiating progeny in regions with SMAD motifs, indicated by dashed red boxes. Merged (mean averaged) of three biological replicates are shown for (D) and (E). (F) Plot expressing enrichment of select motifs in differential ATAC peaks between each cell type. See also Supplemental Table S13. Motifs with −log10(P-value) >300 are expressed as a maximum of 300 here. (G) Motif analysis within peaks subset by the following parameters: peaks enriched in differentiating progeny relative to McSCs and 50 kb ± TSS of genes down-regulated in Bmpr1a-null cells. [*] MITF family, [**] LEF/TCF family. See Supplemental Table S14. (H) Motif analysis within differentiating McSC progeny peaks within 50 kb ± TSS of genes whose expression is diminished in Bmpr1a-null cells. Table highlights LEF1, MITF, and SMAD(2, 3, and/or 4) motifs. Genes in bold specify those encoding proteins involved in melanogenesis. See also Supplemental Table S15.

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**Table 1**

| Motifs         | Oca2, Strf4, Nry, Padd2 |
|----------------|-------------------------|
| LEF1 only      | Scl26a2, Enpep, Ogdhl   |
| LEF1 & MITF    | Shroom2, Gpr143, Iig1, Me1, Scl26a2, Nav2, X6d4,.Adam10, Adam1, Gps1, Pht1, Pabgp, Gnas3, Pdgf15, Smad1, Hnf4b, Hnf4a4b, Tubb3, Sic7a1, Eno1, Egif6, Fam33b, Pim2|
| LEF1 & SMAD    | Mc1r, Rab38, Tprf,Scl26a4, Kif20a, Gpr13b, Epha4, Edn1, Cced, St63, Fcch, Ptnr, Mtas1, Wnt1, Akap9, Ibg3, Abp5051, Pren4, Tac1, Apo19l, Dach1, Scm1, Sec141f1, Smad1, 493052s3070k |
| MITF only      | Fmtn4a                  |
| MITF & SMAD    | Nnt1, Tmem47, Tfx1, Insl1, Pnor1 |
| SMAD only      | Scl45a2, Tprf, Iig3, Gnn, Calm1, Pilpem2, Pem1, Tmem133d, Dhhm2, D7en0443a, B3gl2e1, Aseh1, Skld |
those expected to play a role in SC biology, such as PAX3 and RBPJ [Notch] [Moriyama et al. 2006]. Motifs such as SOX10 were present in both qMcSC and McCP peaks [Fig. 8F].

Focusing on the entire cohort of genes whose expression was diminished upon loss of BMPR1a, we analyzed differentiation-enriched ATAC peaks that were ±50 kb [putative promoter and enhancer regions] of the TSS of these genes [Fig. 8C]. Again, we found significant enrichment of motifs for the binding of LEF/TCF [TCF7L2, TCF7, and LEF1] and MITF [Fig. 8G; Supplemental Table S14]. Consistent with our single-cell analysis of stemness factors [Fig. 1E], AP1 family motifs dominated the motifs in peaks increased in qMcSCs [Supplemental Table S14]. Finally, to assess how SMAD-mediated regulation might impact chromatin dynamics, we analyzed differentiating progeny for their promoter/enhancer chromatin peaks that were associated with genes whose expression was markedly changed upon BMPR1a loss. In agreement with our predictive motif scanning [Fig. 7F,G], our analysis suggested that many of the key genes involved in melanin biosynthesis, transport, and trafficking [i.e., Mcl1, Rab38, and Trpm1] may require cooperation of SMAD as well as MITF and LEF1 in controlling their lineage-specific activation [Fig. 8H; Supplemental Table S15].

Discussion

BMP signaling at the helm of melanocyte maturation

By devising methods to purify and transcriptionally profile individual cells from their native tissue context and at discrete steps along the melanocyte lineage, we made major inroads into illuminating the signals involved in this progression. Taking an unbiased approach, we harnessed the power of scRNA-seq to map the landscape of global gene expression changes that occur as McSCs transition from quiescence to activation to differentiation. In all, we discovered six discrete transcriptional profiles and uncovered signaling pathways and a plethora of transcriptional changes that occur as McSC progress through differentiation.

McSCs and HFSCs must coordinate their behaviors. A priori, because BMP signaling functions to maintain HFSCs in a quiescent state [Kobiela et al. 2007; Gendner et al. 2014; Adam et al. 2018], this pathway might have been predicted to also govern McSC quiescence. However, qMcSCs appear to be largely refractory to BMP signaling at the helm of melanocyte maturation. While quiescent HFSCs exist as a single layer flanking the BMP-rich inner bulge, qMcSCs reside near the base of the bulge, where BMP signaling is lower [Hsu et al. 2011; Yang et al. 2017]. Intriguingly, TGF-β signaling is elevated at the bulge base [Oshimori and Fuchs 2012], which could explain the paradoxical preferential sensitivity of McSCs to this related signaling pathway [Nishimura et al. 2010].

In contrast to the proliferative effects of WNT signaling on McSCs [Rabban et al. 2011], BMP signaling appeared to function after McCP specification. At this stage, cells were proliferative, but showed considerable signs of differentiation. Interestingly upon TyrCreER-mediated ablation of Ctnnb1 in the lineage, nuclear pSMAD1/5/9 was still seen McSC hair bulb progeny, suggesting that BMP signaling is not dependent on WNT signaling. In contrast, LEF1 was diminished in Bmpr1a hair bulb McSC progeny, indicative of a reliance of BMP signaling in WNT-mediated melanocyte maturation.

The reduction in MITF in Bmpr1a-null cells further positioned BMP signaling at the helm of melanocyte maturation. Indeed, loss of BMP signaling skewed the hair bulb population toward proliferative and less differentiated McCPs. In this regard, it is notable that in normal skin, BMPs are elevated in the apical region of the dermal papilla; i.e., the region of the hair bulb where proliferative precursors of both the hair shaft and the melanocytes reside [Yang et al. 2017]. BMP signaling also directly regulates a cohort of hair-specific genes in the proliferative progenitors of the hair shaft [Gendner et al. 2014; Adam et al. 2018].

Our new findings here suggest that BMP signaling plays a similar role in the melanocyte lineage by governing the expression of later stage pigmentation genes involved in melanosome maturation and transport. For melanocyte maturation, however, at least some of BMP’s effects may be indirect, operating through BMP-mediated activation of Lef1, followed by LEF1-mediated activation of Mitf and/or stabilization of MITF [Finan et al. 2009; Nishimura et al. 2010]. BMP signaling is not dependent on WNT signaling. In contrast, LEF1 was diminished in Bmpr1a hair bulb McSC progeny, indicative of a reliance of BMP signaling in WNT-mediated melanocyte maturation.

BMP signaling at the crossroads of MITF and LEF1

In placing BMP signaling upstream of Lef1 expression as well as both nuclear LEF1 and MITF, we could also begin to understand the profound effects of BMP signaling on pigment production, as well as a failure to progress appreciably past the immature McCP stage in the hair bulb. While the “rheostat model” alone might explain why some MITF targets were still expressed in Bmpr1a-null cells [i.e., low levels induce expression of genes involved in survival and proliferation] and others are diminished [i.e., high levels induce genes involved in melanosome biogenesis and transport], our data also hinted at the
importance of cooperation with the transcription factor LEF1 to completely achieve differentiation. Previous studies have provided some insights into how this interplay might occur, including a physical interaction of the two factors [Yasumoto et al. 2002] or stabilization of MITF protein through enhanced WNT signaling [Ploper et al. 2015].

The temporal nature of our gene expression data set made it a powerful tool with which to elucidate the precise coordination of niche pathways and transcription factor responses that interact to achieve quiescence, activation, proliferation, and differentiation in the melanocyte lineage. Most interesting was the marked sensitivity to BMP signaling that occurred in genes whose putative promoters harbor both LEF1- and MITF-binding motifs. When coupled with the specific genes implicated, these findings support a model whereby BMP signaling acts to activate Lef1, and LEF1 then promotes nuclear accumulation of and cooperates with MITF to govern a cohort of melanocyte maturation genes. Importantly, our chromatin landscaping analyses of McSCs in the bulge and their differentiating progeny in the hair bulb provided compelling evidence in support of this model. First, they revealed that a number of regions within the Lef1 gene locus that show enhanced chromatin-accessibility upon McSC lineage progression harbor SMAD motifs, corroborating our finding that Lef1 mRNAs are down-regulated upon Bmpr1a ablation, and that Lef1 is likely a direct BMP target gene. Second, they revealed that among the cohort of genes whose expression diminishes upon Bmpr1a ablation, many show ATAC peaks whose accessibility increases in differentiating wild-type McCPS/melanocytes and that have binding motifs for SMAD, LEF1 and/or MITF. Together, these data provide new insights into how BMP signals in this lineage.

**BMP signaling and melanoma: Are there lessons to be learned?**

BMP signaling has been correlated with melanoma invasiveness and migration [Hsu et al. 2005; Rothhammer et al. 2005; Braig and Bosserhoff 2013; Sinnberg et al. 2018; Venkatesan et al. 2018; Gramann et al. 2019]. Indeed, in zebrafish models of melanoma, signaling through BMP family member GDF6 has been shown to suppress differentiation through down-regulation of MITF and promote malignant progression [Venkatesan et al. 2018; Gramann et al. 2019]. Our findings provide compelling evidence that at least in the mammalian context, blocking BMP signaling suspends the melanocyte lineage in a committed progenitor-like state, where cells are proliferative but unable to fully differentiate. In this way, BMP signaling might have both oncogenic and tumor suppressive roles in melanoma, reminiscent of studies in the lineage where both promelanogenic/differentiation and anti-melanogenic/differentiation effects have been observed [Bilodeau et al. 2001; Jin et al. 2001; Sharov et al. 2005; Yaar et al. 2006; Kawakami et al. 2008; Park et al. 2009; Han et al. 2012; Singh et al. 2012; Yang et al. 2014].

Indeed, related family member TGF-β can enhance proliferation of tumor-initiating cells when signaling is silenced, while simultaneously playing a powerful counter role in promoting invasion and metastasis [Guasch et al. 2007; Oshimori et al. 2015; David et al. 2016; David and Massagué 2018, Su et al. 2020]. It is tempting to speculate that in melanoma, the juncture between the proliferative progeny of McSCs and differentiating melanocytes might become plastic, allowing cells to flip in their reliance upon BMP signaling, depending on the stage of disease progression. Future studies in mammalian models of melanoma may illuminate whether such mechanisms might be at work in disease.

**Materials and methods**

Contact for reagent and resource sharing

Additional requests for resources or information should be directed to and will be fulfilled by Elaine Fuchs (fuchslb@rockefeller.edu).

**Experimental models and subject details**

**Mice and procedures** Mice were maintained in the Association for Assessment and Accreditation of Laboratory Animal Care (AAALAC)-accredited animal facility [Comparative Bioscience Center] of The Rockefeller University. All procedures were performed under Institutional Animal Care and Use Committee (IACUC)-approved protocols. Genotyping was performed using the primers listed in Supplemental Table S16.

**Dct-eGFP mice** [Tg(Dct-EGFP)NY2Gsat/Mmucd, RRID: MMRRC_032849-UCD] were obtained from Mutant Mouse Resource and Research Center (MMRRC) at University of California at Davis, an National Institutes of Health-funded strain repository, and was donated to the MMRRC by Nathaniel Heintz, Ph.D., of The Rockefeller University [Gene Expression Nervous System Atlas (GENSAT)] [Gong et al. 2003]. Lef1-RFP mice were generated by Rendl et al. [2005]. Dct-eGFP [mixed C57BL6/J] [Jax 006644] males were crossed to CD1 females [CRL, strain 022] [to increase litter size and improve maternal care relative to C57BL6/J females] to generate mice used for qMcSC and aMcSC studies [Dct-eGFP]. Background strain was kept as consistent as possible to generate Dct-eGFP Lef1-RFP mice, where Dct-eGFP [mixed C57BL6/J] males were mated to Lef1-RFP [CD1 background] females. For all RNA and ATAC-seq experiments, male mice were used to minimize hair cycle variation and to acquire more cells and tissue per experiment.

To assess McSC proliferation [activation], two 100-µL doses of 5 mg/mL 5-ethyl-2′-deoxyuridine (EdU) were administered by intraperitoneal injection. Mice were euthanized and skin was collected for IMF analysis 24 h after the first injection. To assess McSC and hair bulb progeny proliferation 1 wk after waxing, mice were given one 100- to 150-µL intraperitoneal injection of EdU 4 h prior to sacrifice and skin collection.

**TyrCreER** [Bosenberg et al. 2006] were purchased from The Jackson Laboratories [JAX 012328] and crossed to our in house colonies of Bmpr1a floxed [Mishina et al. 2002] R26YFP floxed [JAX 006148; Srinivas et al. 2001] [mixed C57BL6/J background]. Fifty microliters of 2% tamoxifen in corn oil [Sigma] was administered by intraperitoneal injection for seven consecutive days starting at P21 [weaning age]. In second telogen [P60-75], the back skin was waxed under isoflurane anesthesia, and buprenorphine was given for pain management. Male mice were singly housed for 1 wk after the procedure until the time of euthanasia. Male mice were used for sequencing studies and microscopy
quantiﬁcations; male and female mice were used for IMF analysis [sex-matched for comparison], and female mice were used for RT-qPCR. The Lef1 KO strain [C57BL/6 background] was obtained from Dr. Rudolf Grosschedl [Kratochwil et al. 1996]. Ctnnb1 floxed [C57BL/6] were obtained from the Jackson Laboratories [JAX 004152] [Brault et al. 2001]. Photographs of mice were acquired with iPhone 6, 7 Plus, and 11 Pro Max cameras and brightened in Fiji/ImageJ as necessary.

Method details

Fluorescence-activated cell sorting To obtain single-cell suspensions for FACS, male back skin was dissected and subjected to chemical and mechanical digestion. Telogen [PS8-P80 for qMcSCs] and anagen I-II [P21 for aMcSCs] skin was dissected and scraped with a dull scalpel to remove excess fat prior to incubation in trypsin EDTA with the dermis side down for 45–60 min at 37°C. The hairy side of the skin was scraped with a dull scalpel against the direction of hair growth to release cells in the HFs. Anagen skin [P9–P10 or 1 wk or 12 d postwaxing in second telogen] was digested dermis side down in 0.25% collagenase [Sigma] in HBSS [Gibco] for 25–30 min at 37°C; DNase [Roche] and 5 mM MgCl₂ were sometimes added to minimize clumping. Cells were then scraped from the dermal side to release cells to the hair bulb. The resulting cell suspensions were quenched with FACS buffer [5% FBS in PBS], filtered, spun down, and washed before incubating with primary antibodies for 25 min on ice. Cells were washed with FACS buffer again before incubating in secondary antibody ~10 min on ice. For bulk RNA-seq, cells were sorted with a 70-µm nozzle directly into TRIzol [Thermo Fisher]. For single-cell RNA-seq, cells were sorted with a 100-µm nozzle into 96-well plates, gating more conservatively to minimize the capture of contaminating nonmelanocyte lineage cells.

The following antibodies were used for labeling the resulting single-cell suspensions: PerCP-Cy5.5- or APC-Cy7-Sca-1 [1:1000], e-Fluor660-CD34, BV421-CD49f or PE-Cy7-CD49f, PE-Cy7-CD117 [1:100], Biotin-CD140a, Biotin-CD31, and Biotin-CD45 [1:150]. For biotin-conjugated antibodies, cells were washed and incubated for 10 min in PE-Streptavidin [1:150] or APC-Cy7-Streptavidin [1:500]. DAPI was used for live/dead cell exclusion. Flow cytometry plots were generated using FlowJo to illustrate the strategies used to isolate each cell population. Manual compensation was performed for presentation.

Immunofluorescence and microscopy For IMF analysis of tissue, Immunofluorescence and microscopy 
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The cut site region was PCR amplified [primers in Supplemental Table S16], and cutting efficiency was tested by T7 endonuclease assay [New England Biolabs]. To generate the Bmpr1a knockout line, the parental clonal cell line was transfected and again sorted as single cells into 96-well plates. Gene knockout was confirmed by MiSeq sequencing [primers in Supplemental Table S16] and the genome.net Cas analyzer tool.

**RT-qPCR** Cultured cells were washed with PBS and harvested in TRI Reagent LS [Sigma]. FACs-purified cells were sorted directly into TRI Reagent LS. Total RNA was purified using the Direct-zol RNA MicroPrep kit [Zymo Research] according to the kit protocol. cDNA was made with the SuperScript VILO cDNA synthesis kit [Thermo Fisher] and diluted before RT-qPCR using SYBR Green PCR master mix [Thermo Fisher]. Ppib2 was used as a housekeeping gene. Primer sequences are listed in Supplemental Table S16.

**Immunoblotting** To harvest, cultured cells were washed on the plate in cold PBS, lysed in RIPA buffer [Millipore] supplemented with protease and phosphatase inhibitors [Roche], and collected by scraping. Cells were lysed for 30 min on ice and then centrifuged to collect the supernatant. Protein concentration was determined by BCA assay [Pierce] against a bovine serum albumin standard curve. Up to 50 μg of protein was run on NuPAGE 4%–12% Bis-Tris gels [Invitrogen] for 1 h at 200 V in NuPAGE MOPS SDS running buffer [Invitrogen]. Protein was transferred overnight onto PVDF membrane at in NuPage transfer buffer [Invitrogen] with methanol at 30 V at 4°C. Membranes were blocked in Odyssey PBS blocking buffer for at least 1 h at room temperature before incubating with primary antibodies overnight at 4°C in Odyssey buffer with Tween-20. Membranes were washed several times in 0.1% Tween-20 in PBS before incubating with fluorescent secondary antibody.

The following primary antibodies and dilutions were used: phospho-Smad1 (Ser463/465)/Smad5 (Ser463/465)/Smad9 (Ser463/467) D8B10 [1:1000], Cell Signaling; SMAID [1:500], Invitrogen; LEF1 [1:1000], Cell Signaling; MITF [1:1000], Cell Signaling, and α-tubulin [1:10000], Invitrogen.

Fluorescent secondary antibodies were used at 1:10000 (donkey anti-rabbit 800 and donkey anti-mouse 680 [LI-COR]). Membranes were imaged with an Odyssey CLx machine [LI-COR].

**Bulk RNA-seq** Cells were FACs purified into TRI Reagent LS, and total RNA was purified using the Direct-zol RNA MicroPrep kit [Zymo Research]. For McSCs, cells from 4 male mice at P60 were pooled for replicate 1, and cells from four male mice at P62 were pooled for replicate 2. For differentiating McSC progeny in the hair bulb, cells from male P10 mice were also pooled: three mice for replicate 1, two mice for replicate 2, and four mice for replicate 3. The Weill Cornell Medical College Genomic Core facility performed quality control analysis and sequencing. Briefly, RNA quality was determined by Agilent 2100 Bioanalyzer, and cDNA libraries were prepared using the Illumina TrueSeq mRNA sample preparation kit [nonstranded, poly-A selection] and sequenced on an Illumina HiSeq 4000 instrument.

**Single-cell cDNA synthesis and library preparation** Single-cell RNA-seq libraries were prepared using a slightly modified version of the Smart-seq2 protocol described by Picelli et al. [2014]. Cells were sorted into 96-well plates containing lysis buffer and flash frozen in liquid nitrogen and stored at -80°C until further processing. External RNA [ERCC] spike-ins were implemented to assess technical noise associated with library preparation and sequencing and to facilitate comparative analyses [Brennecke et al. 2013]. Plates were thawed on ice, then lysed at 72°C for 3 min. The RNA was subject to reverse transcription (4U/μl Maxima H transcriptase), template switching reaction, and whole transcriptome amplification, and cDNA was purified with AMPure XP beads. To test for low quality libraries or empty wells, RT-qPCR for GAPDH was performed before proceeding. Illumina sequencing libraries were then prepared using the Nextera XT DNA library preparation kit [Illumina]. After barcoding, samples were pooled and cleaned up with 0.9X by volume AMPure XP beads [Beckman Coulter]. Pooled libraries were sequenced on an Illumina NextSeq 500 instrument using a 75-bp paired-end read mid-output setting.

**ATAC library preparation** Cells were FACs purified using an 85-μm nozzle for gentler sorting to improve viability. McSCs were isolated as described in Supplemental Figure S2A from Dct-eGFP+/CD1 second telogen skin at P60 (replicate 1, cells from three male mice pooled) and P80 [replicates 2 and 3, cells from four male mice pooled for each], differentiating progeny were isolated from Dct-eGFP+ [C57BL6/J background with some CD1] anagen skin at P9 [replicate 1, cells from one male mouse; replicates 2 and 3, cells from two male mice pooled for each] using the following FACS strategy: dump (CD31, CD45, CD140a−, Sca-1−, CD34−, eGFPhigh) and CD117+ cells. Library preparation was performed as described [Buenrostro et al. 2013] with some modifications. Briefly, after sorting, cells were washed with PBS, pelleted, and resuspended in cold lysis buffer (10 mM Tris-HCl at pH 7.4, 10 mM NaCl, 3 mM MgCl2, 0.1% Igepal CA-630). Lysis buffer was removed by centrifugation, and samples were incubated in the transposition reaction for 30 min at 37°C [Illumina Nextera DNA preparation kit, TDE1 enzyme, 10 μL]. The reaction was terminated by adding Tagmentation Clean Up buffer (300 mM EDTA, 900 mM NaCl) and purified with the Qiagen MinElute PCR purification kit. The DNA was then PCR amplified with barcode identifiers for 12–18 cycles, and the products taken at three cycling times were tested by D1000 tape station [Agilent]. Optimal samples were then pooled and bead purified before sequencing on NextSeq High Output 75 single read (40×40-bp paired-end).

**Schematics** Some graphics from SMART Servier Medical Art [https://smart.servier.com] were used to make to make schematics. Other illustrations were made in Adobe Illustrator and Microsoft PowerPoint.

**Quantification and statistical analysis**

**Statistical analyses** Statistical tests for microscopy quantifications and RT-qPCR were performed with GraphPad Prism 7. Column data was first subjected to D’Agostino and Pearson normality testing. For data without normal distribution, two-tailed Mann–Whitney test was used to compare two groups, and Kruskal–Wallis test with Dunn's with multiple comparisons was used to compare between more than two groups. For RT-qPCR, where n number was too low for normality testing, unpaired t-test or ordinary one-way ANOVA with Dunnett’s multiple comparison tests were performed.

**Bulk RNA-seq analysis** Sequence and transcript coordinates for mouse mm10 UCSC genome and gene models were retrieved from the Bioconductor Bsgenome.Mmuscus.ucsc.mm10 (version 1.4.0) and TxDb.Mmuscus.ucsc.mm10.knownGene (version 3.4.0) Bioconductor libraries, respectively. Read length was 51 bp. Transcript expressions were calculated using the Salmon quantification software [version 0.8.2] [Patro et al. 2017].
and gene expression levels as TPMs and counts were obtained using Tximport [version 1.8.0] (Soneson et al. 2015). Normalization and rlog transformation of raw read counts in genes were performed using DESeq2 [version 1.20.0] (Love et al. 2014). Variability between samples was assessed with hierarchical clustering and heat maps of between sample distances implemented in the Pheatmap R package [1.0.10], as described by Love et al. (2014), and sample similarity is expressed as Euclidean distance as well as Pearson correlation. Published HFSC RNA-seq data from Ge et al. (2017) was obtained from Gene Expression Omnibus (GEO) accession number GSE89928 (GSM2656733 CL_BuA and GSM2656734 CL_BuB).

Single-cell RNA-seq analysis Sequence and transcript coordinates for mouse release M23 (GRCm38.p6) genome and gene models were downloaded from GENCODE [https://www.gencodegenes.org/mouse/release_M23.html]. Adaptors were first trimmed from reads using Skewer (version 0.2.2). Sequencing reads for scRNA-seq libraries were aligned to the mouse reference genome, combined with sequences for ERCC spike-ins as artificial chromosomes, using STAR [version 2.5.2a] (Dobin et al. 2013) with default parameters for paired-end reads. Transcript expressions were calculated using the Salmon quantification software [version 0.14.1] (Patro et al. 2017) and gene expression levels as TPMs and counts were obtained using Tximport [version 1.12.3] (Soneson et al. 2015). TPMs were transformed to log2[TPM + 1]. For downstream analyses, cells with <1250 genes detected per cell and genes expressed in <5% of the cell population were removed. We also excluded cells expressing lower levels of the lineage marker Dct [log2[TPM] ≤ 6] and those expressing the HF lineage gene Krt15 [log2[TPM] > 6]. After filtering, the number of cells in our data set were as follows: qMcSCs n = 104 cells [n = 3 mice], aMcSCs n = 193 cells [n = 3 mice], hair bulb McSC progeny n = 308 cells [n = 2 mice], Bmpr1a wild-type control n = 205 cells [n = 1 mouse], Bmpr1a heterozygous n = 304 cells [n = 2 mice], and Bmpr1a-null n = 421 cells [n = 2 mice].

Analyses and visualization of data were conducted in a Python environment built on the Numpy, SciPy, matplotlib, scikit-learn package, and Pandas libraries (Pedregosa et al. 2011). Batch effect variation between sequencing runs and biological replicates was assessed by examining the Euclidean distances in PCA space between replicates within each specific time point data set (corresponding to one of qMcSCs, aMcSCs, or various stages of melanocyte differentiation) versus across time points, which showed closer relationships between replicates compared with variation between biologically distinct time points. Furthermore, PCA analysis of the only data set that showed some evidence of replicate-specific clustering patterns (P10 anagen VI) captured biologically relevant differences in gene expression separating replicates, including variation in M/G1 cell cycle phase, melanosome organization, and BMP signaling pathway genes. This strongly suggests the differences between P10 replicates is likely to represent true biological variation between animals in this highly dynamic cell population, rather than technical batch effects.

To distinguish true biological variability in gene expression from technical noise, we used a statistical model for identifying highly variable genes compared with ERCC spike-ins as described by Brennecke et al. (2013). Briefly, we used a custom script based on the methodology described by Brennecke et al. (2013, running in R version 3.6.1), to identify those genes with higher level of variation [at least 10% above the technical variation] and false discovery rate (FDR) value less than 0.1. To identify cell clusters and visualize the data, we first centered and scaled the highly variable gene data set and performed PCA on the list of highly variable genes. A jack-straw approach for the first 100 components [based on Seurat v3.1.1] (Stuart et al. 2019) was implemented in R to calculate a statistical significance for each principal component (PC), and components with P < 0.05 were retained for future analysis. Significant PCs were used as input for nonlinear dimensionality reduction, performed using UMAP implemented in scikit-learn. To identify clusters we used a graph-based clustering approach based on building a k-nearest neighbors (kNN) graph and clustering with the Louvain algorithm [with k set to one-fifth of the data set size, and resolution parameter of 1 × 10−5]. Euclidean distance in PCA space served as input for both UMAP generation and Louvain clustering. During second level analysis, we used the same strategy for dimensionality reduction, clustering and visualization, with variable genes and PCs determined separately for each of the main cell clusters.

To construct pseudotime trajectories we used the same strategy for dimensionality reduction and clustering of each data set as before, constructed a principal graph based on the community analysis, and computed pseudotime values for each cell using Monocle 3 [version 0.2.0] (Trapnell et al. 2014; Cao et al. 2019). To identify genes that change as a function of pseudotime we used the “graph_test()” function of Monocle 3, to implement Moran’s I test. Genes were considered to be significantly changing with pseudotime with a q-value < 1.0 × 10−4.

Differential gene expression was used to identify genes specific to each cluster. Briefly, we used raw count matrices for the biologically variable genes within each data set and applied them to the DESeq2 package [version 1.24.0] (Love et al. 2014) using R. We used a negative binomial fit to model differential gene expression, factored the data set based on the Louvain cluster assignments, and used a threshold of 0.75 to construct Wald tests of significance. Genes were considered to be differentially expressed if log2[fold change] ≥1 and P-adjusted value [q] ≤0.05. Lowly expressed differential genes [baseMean expression ≤5] were discarded from visualization and further analysis. The expression levels of specific genes of interest were visualized as log2[TPM + 1] values on the corresponding UMAP representation of the data. The “stemness” and “differentiation” signature gene sets used to generate cumulative “stemness” or “differentiation” scores are the top 100 up-regulated or down-regulated differentially expressed genes between McSCs and differentiating hair bulb McSC progeny, and largely encompassed genes previously implicated in their biology. The “AddModuleScore” function from Seurat version 3.1.1 was used to calculate the average expression levels of each gene set on a single-cell level, subtracted from the aggreagated expression of control feature sets, as was originally described in Tirosch et al. (2016). Resulting stemness or differentiation score values for each cell are color-coded on corresponding UMAP visualizations of the data, or on violin plots binned by subcluster cell identity.

Gene ontology (GO) analyses of differentially expressed genes were performed using PANTHER tools overrepresentation test [version 14.0], and annotated using the GO biological process complete list (Thomas et al. 2006, Mi et al. 2019). Significant associations were calculated using the Fisher’s exact test, and multiple comparisons were corrected for using the false discovery rate (FDR). GO terms were arranged hierarchically, and parent terms with greater than twofold enrichment and FDR < 0.05 were selected for presentation. GO terms explicitly linked to tissues other than skin/HSFs were excluded from presentation in the main text, but are included in Supplemental Tables. Where indicated, GO analyses were also annotated using DAVID Bioinformatics Resources 6.8 “Functional Annotation,” “BP Direct” lists (Huang et al. 2009a,b).

To visualize changes in gene expression across data sets, heat maps were constructed in a Python environment built on the
expression was used. The cell cycle an amalgamated list of cell cycle stage specific gene ex-
cific gene lists were obtained from Macosko et al. (2015). To gen-
lists are available in Supplemental Table S8. Cell cycle stage spe-
tissues, and be expressed in our data sets. Complete target gene
the signaling pathway in at least two mesenchymal or epithelial
to be included, target genes had to be validated as direct targets of
BMP or ca-
se, for example for BMP or canonical WNT signaling pathways,
maps using the clustermap() function with the previously gener-
the farthest point algorithm. The matplotlib extension Sea-
tween genes calculated based on pairwise correlation distances
matrix was scaled and centered, and hierarchical linkages be-
dotime heat maps by-passed this filtering step. Each expression
analysis was based solely on biologically variable genes, the pseu-
the cluster cells) were retained for visualization. As pseudotime
was used to filter the biologically variable genes associated with
GO 2 through the Gene Ontology Consortium. Each GO gene list
change in the

Bmpr1a-null cluster relative to the control cluster. Motifs had to
be in at least one TSS to be considered positive for that
genes were then grouped based on the presence or absence
sequences under these

motifs existed in the TSS. The difference in fold change between two
groups were tested by Wilcoxon test with the null hypothesis that
there is no difference in the distribution between the test
groups. The results were plotted with R



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Numpy, SciPy, matplotlib, scikit-learn package and pandas li-
braries [Pedregosa et al. 2011]. For specific GO annotations, the
corresponding Mus musculus gene lists were obtained from Ami-
GO 2 through the Gene Ontology Consortium. Each GO gene list
was used to filter the biologically variable genes associated with
the data set to generate normalized expression matrices, and
genes characteristic for clusters of cells [expressed in >50% of
the cluster cells] were retained for visualization. As pseudotime
analysis was based solely on biologically variable genes, the pseu-
dotime heat maps by-passed this filtering step. Each expression
matrix was scaled and centered, and hierarchical linkages be-
tween genes calculated based on pairwise correlation distances
using the farthest point algorithm. The matplotlib extension Sea-
born visualization library [version 0.9.1] was used to create heat
maps using the clustermap() function with the previously gener-
ated gene-based hierarchical linkages fed in.

To generate comprehensive gene sets based on GO term analy-
yses, for example for BMP or canonical WNT signaling pathways,
the corresponding Mus musculus gene lists were obtained from Ami-
GO 2 through the Gene Ontology Consortium. BMP or ca-
nonical WNT target gene lists were culled from mouse literature.
To be included, target genes had to be validated as direct targets of
the signaling pathway in at least two mesenchymal or epithelial
tissues, and be expressed in our data sets. Complete target gene
lists are available in Supplemental Table S8. Cell cycle stage spe-
cific gene lists were obtained from Macosko et al. [2015]. To gen-
erate a global proliferation gene set score across all stages of the
cell cycle an amalgamated list of cell cycle stage specific gene ex-
pression was used. The “AddModuleScore” function from Seurat
v3.1.1 was used to calculate the average expression levels of each
gene set on a single-cell level, subtracted by the aggregated ex-
pression of control feature sets, as was originally described in Tir-
osh et al. [2016]. Resulting gene set scores for each cell are color-
oded on corresponding UMAP visualizations of the data, or on
violin plots binned by subcluster cell identity.

Promoter analysis HOMER genome-wide motif predictions were
used to scan for LEF1, MITF, or SMAD motifs within 2.5 kb up-
stream of the transcription start site (TSS) of the genes that
were significantly [pad ≤ 0.05, baseMean ≥ 5] changed in the
Bmpr1a-null cluster relative to the control cluster. Motifs had to
be found in at least one TSS to be considered positive for that
group. Genes were then grouped based on the presence or absence
of motifs in the TSS. The difference in fold change between two
groups were tested by Wilcoxon test with the null hypothesis that
there is no difference in the distribution between the test
groups. The results were plotted with R



ATAC-seq analysis ATAC-seq reads are aligned to the mm10 ge-
nome from the Bsgenome.Mmuseus.UCSC.mm10 Bioconduc-
tor package [version 1.4.0] using Rsubread’s align method in
paired-end mode with fragments between 1 and 5000 base-pairs
considered properly paired [Liao et al. 2013]. Normalized, frag-
ment signal bigWigs were created using the rtracklayer package.
Peak calls were made with MACS2 software in BAMPE mode
[Zhang et al. 2008; Feng et al. 2012] and sequences under these
peaks used with the motifMatcher Bioconductor package and
JASPAR motif database [Fornes et al. 2020] to identify known
motifs within ATAC-seq peaks. Differential ATAC-seq signal
was identified using the DESeq2 package [Ross-Innes et al.
2012; Love et al. 2014] and enrichment for motifs identified using
the Limma packages geneSetTest function [Ritchie et al. 2015].
Integrative Genomics Viewer (IGV) was used for visualization.

Data and software availability

The data sets and code generated during this study are available at
GEO [GSE147299], https://www.ncbi.nlm.nih.gov/geo/query/acc
cgi?acc=GSE147299.

Competing interest statement

E.F. is a scientific advisory board member of L’Oreal and of Arse-
nal Biosciences.

HOMER genome-wide motif predictions were used to scan for LEF1, MITF, or SMAD motifs within 2.5 kb upstream of the transcription start site (TSS) of the genes that were significantly [pad ≤ 0.05, baseMean ≥ 5] changed in the Bmpr1a-null cluster relative to the control cluster. Motifs had to be found in at least one TSS to be considered positive for that group. Genes were then grouped based on the presence or absence of motifs in the TSS. The difference in fold change between two groups were tested by Wilcoxon test with the null hypothesis that there is no difference in the distribution between the test groups. The results were plotted with R [3.6.1] and ggplot2.

ATAC-seq analysis ATAC-seq reads are aligned to the mm10 genome from the Bsgenome.Mmuseus.UCSC.mm10 Bioconductor package [version 1.4.0] using Rsubread’s align method in paired-end mode with fragments between 1 and 5000 base-pairs considered properly paired [Liao et al. 2013]. Normalized, fragment signal bigWigs were created using the rtracklayer package. Peak calls were made with MACS2 software in BAMPE mode [Zhang et al. 2008; Feng et al. 2012] and sequences under these peaks used with the motifMatcher Bioconductor package and JASPAR motif database [Fornes et al. 2020] to identify known motifs within ATAC-seq peaks. Differential ATAC-seq signal was identified using the DESeq2 package [Ross-Innes et al. 2012; Love et al. 2014] and enrichment for motifs identified using the Limma packages geneSetTest function [Ritchie et al. 2015]. Integrative Genomics Viewer (IGV) was used for visualization.
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